Dear editor,

Please find enclosed our detailed responses to the minor revisions suggested by both reviewers. Most of the figures have been reworked to include labels, following suggestions by D. Lunt. New figures have been added as well (Fig. 1; Fig. 14; Fig. 16; Fig 17; Fig. 18; Fig. 20). We have moved the performance section in the appendixes as suggested by rev#1. We provide several new references and have reworked the introduction to fit with both reviewers suggestions. We benefited from the help of J.-L. Dufresne, who is now a co-author of the paper. The detailed responses can be found below.

Best regards,

Pierre Sepulchre on behalf of co-authors.

Anonymous Referee #1

Overall Comments:

I applaud the authors for this very impressive and detailed work. The paper covers a lot of ground and will appeal to a diverse audience including computational scientists and software engineers (technical and performance analysis), model developers (tuning decisions and model details), climate scientists (model validation against observations and similar earth system models), and finally, paleoclimate modellers and developers (deep time case study). I do have a number of suggestions, however, which are detailed below. Primarily, they are mostly minor, but important to improve the readability of this manuscript. Inconsistencies in organization and figures must be addressed before full publication.

We thank referee #1 for these comments. In the following we have replied to every comment and suggestion made, and modified the manuscript accordingly.

As the focus in main paper on scientific improvements, the authors may want to consider moving the performance-related discussion to the appendix, and climate state figures back into the main text. Another idea might be to keep the appendix material themed; something like (a) performance analysis and (b) modes of variability. Although both of these topics are very important, if the main paper covers the main model improvements and climate state analysis, it may help to tighten up the manuscript.

We agree. We have moved the performance section to the appendix and reorganized the whole section.

Lastly, although the authors note that the case study is not meant to be an evaluation of the IPSL model with the Cretaceous time period, I would like to see more discussion on how it compares to proxy data (or other published work).

Our Cretaceous simulation is only intended for illustrating the steps that were necessary to adapt IPSL-CM5A2 to the deep-time and to create deep-time boundary conditions. We refer the reader to the paper by Laugié et al. (2019) for a complete comparison of the Turonian climate simulated with IPSL-CM5A2 with reconstructions inferred from the proxy record. Laugié et al. (2019) report results from a slightly different Cretaceous simulation performed with
IPSCLM5A2. The Laugé simulation uses the same land-sea mask and boundary conditions than we do but prescribes an estimation of the tidally-driven mixing associated with the dissipation of internal wave energy for the M2 tidal component. In addition, our Cretaceous simulation uses different values in the parameterization of the dissipation time of atmospheric waves, which were originally prescribed for numerical stability at the start of the simulation but were not corrected. These values make our Cretaceous simulation about 1 to 1.5°C warmer than the Laugé simulation, which uses appropriate values in the parameterization.

**Specific Comments:**

**Introduction:** Somewhere in the introduction, the acronym IPSL should be defined.

Done.

Lines 20-66: The CESM system has been run for multi-millennia for CCSM3, and now, currently with CESM1 under the TraCE project. Although this is not continuous in that the model is stopped, adjusted, and restarted for separate timeslices, it is a multi-millennia simulation nonetheless. For CCSM3:

http://www.cesm.ucar.edu/working_groups/Paleo/documentation/sims-projects/trace-21ka/ and Feng He Phd dissertation (U. Wisconsin, 2011); for CESM1, contact Dr.He.

Thank you for this suggestion. We now mention this work, and also cite the recent Kutzbach et al. paper (PNAS, 2020), that uses acceleration techniques to mimic multi-millenial simulations with CCSM3. This section has also been totally reworked following reviewer #2 suggestions.

Line 39: Define “branched”. Even for climate modellers, this term can be a source of confusion.

Done. The sentence now reads: “Later, experimental designs with GCMs for deep-time paleoclimate modelling relied on simulations branched on each other (i.e. final-state of equilibrated simulations is used as initial state for new simulations with a different forcing) to make the ocean reach different equilibria in a reasonable amount of time […]”

Paragraph Line 67: For those not familiar with the IPSL model, a better overview of the original model is needed for the introductory context, i.e. a mention of all model components (are there not sea ice and river runoff in NEMO?). The Dufresne et al reference is not enough.

We now mention each component in this section. We describe later in the text the different components of both models. In particular, we have added a more thorough description of LIM2, that now reads : “LIM2 (Fichefet and Maqueda, 1997; Timmermann et al., 2005) is a sea-ice model with a single sea-ice category. Open water is represented through ice concentration. As stated in (Uotila et al., 2017) who provide a detailed comparison between LIM2 and LIM3 in the 170 NEMO, LIM2 implements the snow-ice formation by infiltration and freezing of seawater into snow when deep enough. The effect of subgrid-scale snow and ice thickness distributions is implicitly parameterised by enhancing the conduction of heat through the ice and by melting the ice laterally to account for thin ice melting. The surface albedo depends on the state of the surface (frozen or melting), snow depth and ice thickness following (Shine and Henderson-Sellers, 1985).”

We also added a table with the main characteristics of the model components.
From IPSL-CM5A to IPSL-CM5A2:
Section 2.1. It would be very useful to see a table summarizing this information including how each component has changed (or not changed), including resolutions.
We agree. The new version now has such a table.

Table 1: What is the significance of the bolded information? Please note in the table caption. Also, JOILOT-CURIE has not been mentioned yet. Adding the heading "supercomputer" to this column would help.
The bolded information highlights the configuration that has been selected. The table and its caption have been corrected.

Sections 2.2 and 2.3. These sections would work just as well as an appendix.
They are now part of the appendix.

Tuning
Table 2 needs units. How are the Trenberth estimates used here?
Table 2 is now corrected. Trenberth values were used as a baseline to evaluate our model results, but we mistakenly omitted them in the submitted manuscript. It is now corrected.

Lines 268 and 280: change “insure” to “ensure”.
Done.

Comparison with observations

Line 323. There is no figure showing spatial biases, so please add “not shown”.
Done.

Line 351. “every region...”
Corrected.

Line 377, Figures 7, 8, 9. It isn’t clear when you are referring to Figure 7 or 8, or Figure 7 or 9 in the text, thus making these sections hard to follow. Also, Line 377 refers to Fig 15 15, when I think the authors mean Fig 9. Please review text and figures for consistency. Also, please make IPSL-CMA titles in the figures consistent with one another. Figs 8/9 uses “LR” when others do not.
Corrected.

Figure 10. This is more than just a Hovmueller (only panels c and d).
Corrected.

Paragraph Line 396 and Figure 11. Consistently refer to figures here as in the manuscript, i.e. use the lettering a,b,c,d. Also, please use the dataset name in the figure
caption, rather than simply “data”. For surface temperature, I assume this is HadiSST? What is used for observation for the salinity, WOA2013? I don’t see this information in text either. The information is now given. Both temperature and salinity observation fields come from WOA2013.

Line 426-429 and Figure 15a. The figure caption needs to be improved. It is clear that 15b, the colors are the differences from the two models, but not so with 15a. If the purpose is to highlight the improvement in AMOC for 15a, then a figure showing the timeseries of AMOC, maximum below 500, between the two models, could be an alternative figure. We have corrected the figure and caption that were actually wrong. We now present the AMOC anomaly for 100 years of preindustrial run for both models (new figure below). We thank the reviewer for suggesting to add the time series of AMOC maximum in the two model versions. Nevertheless, there is no point in strictly comparing the time evolution of these two time series as there is no reason for internal variability in the two simulations to be phased. We thus prefer to stick with the comparison between spectra as shown on Fig. 17.

![Figure 15a](image1.png)

Section 4.2.4 and Figure 17.
Consistency between this figure and others in this manuscript regarding lettering or positional references, and labeling of models is needed. Also, the caption needs to be improved to include years and model names.
Corrected.
For the text, some discussion with observations is needed for context.
Done. We now refer to several publications to discuss the 20-yr variability.
Figure A5. What is the difference between the gray and black lines in the power spectra? Also, labelling of appendix figures needs to be consistent with the rest of the manuscript.

The grey line shows values of the HadISST dataset for comparison.

Figure A6. Consistency in labelling.
Corrected.
Deep time case study
Line 503. “...ran a 3000-year long...”Section 5.1 and Figure 18. I am not sure what I am looking at in Figure 18a. Are the white areas supposed to be the grey-shading for the refinement? (This could be a document format issue so, perhaps noting the refinement areas with a black box rather than colored shading would be better). Also, for Figure 18b, my recommendation would be to make this with the cretaceous continental outlines, not modern. I think the discussion of pole and refinement placement would be much clearer to readers who have never modified grids for deep time if the paleo outlines were used.
We have added the cretaceous continental outlines as recommended. The grey shading on fig 18a (now fig. 17a) depicts the fact that continental points have been used to refine the resolution over the different seas mentioned in the caption text.

Section 5.2. What did you use for other important forcings, such as solar input and methane? Should we assume that all else is based on pre-industrial forcings? Does this model prescribe or predict aerosol emissions? Dust, sea salt, carbon from fire, and other biogenic aerosols such as secondary organic aerosols (isoprene, etc), may also be extremely important climate forcings, even for deep time.

Thank you for pointing out these omissions. The solar constant was set at Cenomanian-Turonian value (1353.36 W.m²). Methane and prescribed aerosol emissions were set at preindustrial values, as this version of IPSL-CM5A2 does not predict aerosol emissions. We totally agree that aerosols emissions could be extremely important for Creataceous climate, but computing them (if possible at all) was out of the scope of this paper.

Line 580. I suspect the “?” was meant to stand in for something? Please fix.
Fixed, it was a missing reference (to Wilson et al. Palaeo-3, 2012) in our latex file.

Line 582 and Figure 19. Should this be Fig 19a? Given that 19b is discussed first, a and b should be swapped in Figure 19.
Corrected. Fig 19 (now fig. 18) has been modified.
Line 592.  Have the authors worked on applying the dynamic vegetation to the Cretaceous? Dynamic vegetation capabilities are highly valued in the deep time modelling community. As in IPSL-CM5A, dynamic vegetation was not activated in these simulations, but we are currently working on implementing this feature in the near future.

Line 607 and Figure 20.  Define T2M. Is it 2-meter air temperature, same as “t2m”? Please make consistent throughout the manuscript.
Yes. Corrected.

Also, I would not consider 20e a “trend” plot as the figure caption states. Although trend can certainly be implied, but the variable being plotted is a whole value as it varies through time, the unit being degC(or Joule).
We have replaced trend by time-series.

Figure 21.  Assuming this is annual mean. How many years (the final 100?) went into this plot?
The mean Cretaceous climate state illustrated on Fig. 21 is indeed the annual mean, and you are correct in inferring that the final 100 years were averaged to make this plot. The caption has been modified accordingly.

Line 635.  Fill in Supplemental Figure XX. Is this referring to figures A8 and A9?
Corrected.

Figure A8 and A9.  Please label consistently. Briefly note/explain why you chose JMA and JAS and northern/southern hemisphere winter.
We use JFM and JAS as Northern/Southern Hemisphere winter because the maximum mixed-layer depth in (deep) convection areas occurs in March and September for the Northern/Southern Hemispheres, respectively.

Conclusions
Line 645.  "This article aims..." I would recommend changing the title of this section to "Conclusions and Discussion" because of 1) the inclusion of new appendix material and 2) the relevance of the new material to the overall discussion on challenges for (deep time) paleoclimate modelling.
Unfortunately the title “Conclusions” is imposed in the GMD template (copernicus.cls in the Latex template).

Appendices
An appendix section describing the many acronyms in this paper would be helpful. Appendix organization needs to be fixed.
Fixed.
There are two A1’s. Fixed.
Also, only the first A1 includes text to introduce the appendix figures. A few sentences introducing each appendix section would greatly improve the readability of this appendices.

Corrected.

Review #2 by D. Lunt.

I was looking forward to reviewing this paper because (a) it is exactly the sort of paper that we hoped the existence of GMD would enable and encourage when journal first started, (b) I have been involved in a similar paper describing our own model (Valdes et al, 2017), and (c) it has a section on configuring the model for deep-time simulations, which is of great interest scientifically to me. I think that this paper will be very useful, providing a definitive benchmark for this model. I hope that the authors start a Special Issue on this model, and submit bug-fixes/developments as they occur, in follow-up papers. I have some questions/comments (none of which are major), some suggestions for additional work (which I would view as nice but optional), and some minor corrections.

We thank Dan Lunt for these comments and the suggestion of starting a special issue on the model. We will consider this option very seriously. In the following we answer the different comments and describe how we addressed them in the manuscript when necessary.

Questions/Comments:

•Introduction: as well as paleoclimate, there will be interest from the forcings/feedbacks community associated with longrun MIP (e.g. Rugenstein et al, 2019). Would be good to mention this. In general, I think you underplay the interest of this model to the non-paleo community.

Agreed, we have modified the introduction accordingly, and included this reference.

•Abstract needs to include the resolution of the model (horizontal and vertical, atmosphere and ocean).

The resolution has been added to the abstract.

•Line 20: Fully complex(e.g. CMIP6) models don’t only carry out snapshots, often transient runs (e.g. RCPs!). More generally, remember the audience for this paper is not just paleo scientists-you should assume an audience that is not familiar with paleoclimates, e.g. avoid terms such as “Quaternary”, “last interglacial” without explaining when they were.

Agreed, we have defined pre-Quaternary in the introduction.
•Line 64: I would remove the statistical approach from the transient section, and instead add another two types of approach here, which are:

(a)“multiple snapshots”. Within this you could include e.g. Singarayer and Valdes (2010), and Marzocchi et al (2015).

(b) statistical approaches based on multiple simulations, e.g. Lord et al (which you already cite), and Simon et al (2017), and Erb et al (2015).

We agree and have modified the text accordingly. We also include more references suggested by reviewer #1, so the entire section has been reorganised. Now it reads:

Through years, several strategies have been set up to overcome the issue of the computing cost of long integrations. They involve both the development of dedicated models and peculiar experimental designs. Regarding the former, different models have been used:

- Earth System Models of Intermediate Complexity (Claussen et al., 2002), with coarse spatial resolution and simplifications in the physics, allow very efficient computation times and have been used extensively to explore Quaternary paleoclimates and run transient simulations (Roberts et al., 2014; Caley et al., 2014). As an example, the iLOVECLIM model (Goosse et al., 2010; Roche, 2013) provides more than 2,500 simulated years per day (hereafter SYPD) on one computing core at a~5.6° spatial resolution (3 vertical levels) in the atmosphere and a full OGCM at 3°x3° (21 verti-35cal levels) with two tracers. Performance is still 850 SYPD with 20 tracers in the ocean (Bouttes et al., 2015b; Missiaen et al., 2019).

- Versions of fully-coupled GCMs have been maintained at very low (~5-7°) spatial resolutions, allowing long integrations in a reasonable amount of wallclock time. Examples include FOAM (Fast Ocean Atmosphere Model). FOAM couples the NCAR CCM2 (7.5°x4.5° and 18 vertical levels but an updated physics equivalent to the NCAR CCM3 model) atmosphere model and an ocean model similar to GFDL MOM. It reaches an average of 250 SYPD and has been routinely used for almost twenty years in numerous deep-time paleoclimate studies (Brown et al., 2009; Donnadieu et al., 2006; Ladant et al., 2014; Pohl et al., 2014; Poulson et al., 2001; Roberts et al., 2014). FAMOUS (FAst MetOffice/UK Universities Simulator), as part of the BRIDGE HadCM3 family of climate models (Valdes et al., 2017), is another example. It runs at a 7.5°longitude x 5°latitude resolution with 11 levels in the atmosphere and averages 450 SYPD, and that is frequently used for paleoclimate (e.g. (Gregoire et al., 2012; Roberts et al., 2014) and future climate studies (Bouttes et al., 2015a).

- Climate modelling groups have also maintained versions of their GCMs at low (~2.5°) resolution. Within the BRIDGE HadCM3 family, HadCM3BL (3.75°x2.5° and 19 levels in the atmosphere; 3.75°x2.5° and 20 vertical levels in the ocean) is used to perform long integrations at 85 SYPD (e.g. (Kennedy-Asser et al., 2020)). NCAR has also developed a low-resolution (T31 –i.e. approximately 3.75°longitude x 3.75°latitude, 26 vertical levels, 35 SYPD) version of CCSM3 GCM (Yeager et al., 2006) that provided “the first synchronously coupled atmosphere-ocean general circulation model simulation from the Last Glacial Maximum to the Bolling-Allerød (BA) warming” (Roberts et al., 2014) as well as transient Holocene simulations (Liu et al., 2014). CCSM3 has been used also to run a transient simulation of the last deglaciation within the TraCE program (He, 2011). NCAR has released a similar version for CCSM4 (Shields et al., 2012), at the same “low resolution”, with performance reaching 64 SYPD. This model has been used for multi-millennial fully-coupled paleoclimate simulations (Brierley and Fedorov, 2016; Burts et al., 2017). A similar strategy has been applied with the ECHAM5/MPI-OM
model (formerly known as COSMOS), which has been used at T31 resolution in the atmosphere for long paleoclimate simulations of the Holocene (Dallmeyer et al., 2017; Fischer and Jungclaus, 2011), and has been recently implemented with water isotopes (Wanner et al., 2016).

Experimental designs dedicated to multi-millenium simulations have also followed different strategies through the years:

- Early methods to circumvent the constraint of computation time have involved (i) using atmospheric-only models forced by prescribed sea-surface temperatures (e.g. (Roberts et al., 2014)), (ii) coupling the atmospheric component to mixed-layer ocean models with prescribed heat transport – i.e. slab ocean models- (Otto-Bliesner and Upchurch, 1997), and (iii) asynchronous coupling between atmospheric and ocean models (Bice et al., 2000). If efficient in terms of computation time, such strategies prevented any assessment of ocean dynamics and associated feedbacks on the climate system. Later, experimental designs with GCMs for deep-time paleoclimate modelling relied on simulations branched on each other (i.e. the final-state of equilibrated simulations is used as an initial state for new simulations with a different forcing) to make the ocean reach different equilibria in a reasonable amount of time (namely less than 1000 years, (Liu et al., 2009)).

- For transient experiments, i.e. with an evolving insolation, Lorenz and Lohmann (2004) proposed an accelerated insolation forcing; when the simulation calendar advances by one year, the calendar for the computation of the incoming insolation advances by 10 to 100 years. The method assumes that the fast components of the climate system (atmosphere, upper ocean) are in constant equilibrium with insolation. It is suitable only for periods of time with almost no change for the slow components (deep ocean, land ice). This method is easily implementable in any climate model (e.g., Crosta et al., 2018). This acceleration technique has been used multiple times with CCSM3, the latest example being the study of the last 140,000 years of climate evolution of Africa by Kutzbach et al. (2020). Others have applied a "multiple snapshots" strategy, involving several multi-centennial (200 to 500 years) simulations forced by varying orbital parameters, to explore the impact of insolation on the fast components of the climate system (Singarayer and Valdes, 2010; Marzocchi et al., 2015).

- Statistical approaches based on multiple simulations also allowed to explore transient climate changes. They involve either the use of statistical emulators, calibrated using several steady-state GCM simulations, with varying orbital configurations and atmospheric pCO2 concentrations (Lord et al., 2017), or linear approximations between single-forcing simulations (Erb et al., 2015; Simon et al., 2017).
• Line 96: It is not always clear from the description which components are identical to IPSL-CM5A, and which are “updated versions of IPSL-CM5A components”. We have now included a table specifying the differences between both versions.

• Line 116 indicates that “The differences mostly include bug corrections, various improvements in the physics/dynamics interface, and energy conservation. Changes also concern the Emanuel convection scheme, with bug corrections and changes in the upper bound of the deep convection loop.”. It would be good to maybe say a little bit more about these, and explain why these changes were made.
  The purpose of this sentence was to say that the changes between LMDZ5A and LMDZ5A2 were essentially computational and had almost no effect on the model results. The sentence was simplified and modified as follows:
  “The differences mainly concern bug fixes and various improvements in the physical/dynamic interface in preparation for the next generation of models. They have very limited effect on model results and translate into a slight improvement in energy conservation.”

• Figure 1 looks pretty, but would include more information if a standard cylindrical projection were used? Or maybe if views towards the Pacific were also added.
  Yes. We have decided to keep this projection and show both hemispheres for LMDZ and NEMO.
Section 2.1: I like the way that the actual tagged releases of each component are given for the new model and for the old model. This is very good practice. Is there also a version number for the overall model, for both new and old versions? And is there a unique update number for NEMO and OASIS as well (the later section on Code Availability indicates that there is)?

We now present a table with specific details on version/revision numbers.

Line 229: Not sure what “adjusted” means in “provided an adjusted global surface air temperature of 11.3°C”

We rephrased to “A first untuned 1,000-year spin-up of IPSL-CM5A2 forced by CMIP5 pre-industrial boundary conditions provided an adjusted global surface air temperature of 11.3°C.”

Line 230: “Such a cold bias had been depicted in the previous IPSL-CM5A (Dufresne et al., 2013)”. What was the magnitude of that bias?

Dufresne et al. (2013) discuss a bias of 1K in the historical run compared to observations.

Line 231: “We targeted to increase global-mean surface temperature by 2.2°C to reach 13.5°C in pre-industrial conditions with IPSL-CM5A2, expecting this value to translate into 15.5°C in simulations with present-day conditions.”. This needs a reference to some observational dataset, either to validate the 13.5 preindustrial choice, and/or the 15.5 present-day choice. Also mention the uncertainty in these observations.

We agree. We rephrased and it now reads:

A first untuned 1,000-year spin-up of IPSL-CM5A2 forced by CMIP5 pre-industrial boundary conditions provided an equilibrated global surface air temperature of 11.3°C. Such a cold bias had been depicted in IPSL-CM5A (Dufresne et al., 2013). We targeted to increase global-mean surface temperature by 2.2°C to reach 13.5°C in pre-industrial conditions with IPSL-CM5A2, expecting this value to translate into 14.5°C for a present-day simulation, a value consistent with observations (Dee et al., 2011).

Line 236: “Previous results obtained with LMDZ and IPSL-CM5A show that changing TOA balance by 1 W.m⁻² results in a change in temperature of 1 K.” I guess you mean “initial TOA imbalance”?

We actually mean TOA balance. We follow the protocol described in Hourdin et al, 2017: “A common practice to fulfill this constraint is to adjust the top-of-atmosphere or surface energy balance in atmosphere-only simulations exposed to observed sea surface temperatures (component tuning) and check if the temperature obtained in coupled models is realistic. This energy balance tuning is crucial since a change by 1 W m⁻² of the global energy balance typically produces a change of about 0.5–1.5 K in the global-mean surface temperature in coupled simulations depending on the sensitivity of the given model.”

Is this the TOA in a PI run, or historical, or....?

The TOA is that of the preindustrial run.
• Line 237: “It is also the typical value of the sensitivity of global temperature to greenhouse gas concentration” is rather opaque….do you mean that the overall net feedback parameter, lambda, in this model is ~1.0 W/m2 / K?
Agreed, we removed this sentence.

• Line 249: “when two of the bugs mentioned above”…there are several bugs mentioned above…which ones are you referring to?
The paragraph has been corrected to specify the two bugs.

• Line 257: “respectively 0.029 W.m-2 and 0.023 W.m−2”. Is the difference between these two primarily due to interannual variability, or gradual accumulation of energy in the ocean, or bugs in energy conservations, or something else?
This difference is most likely linked to the fact that the model is not fully energy-conservative.

• Line 258: What are the causes of the non-conservation of energy?
The non conservation of energy in the model has several causes. Conversion terms between potential energy and kinetic energy and dissipation of kinetic energy to heat are not fully considered (both in the atmosphere and in the ocean). The energy fluxes into the ocean associated with rain and snow, riverflow and icebergs are also approximated.

In very long simulations, small non-conservation of energy is not a problem, because the energy system is “open” to the outside –i.e. a small non-conservation may have very similar results to slightly increasing the solar constant –the model just finds a new equilibrium. However, the hydrological system is “closed”, and so non-conservation can result in substantial drifts in e.g. global mean salinity over very longtime periods. How well does the model conserve water, and are any correction factors included to maintain global mean water/salinity content? OK, I see this is addressed later (lines 272-280). Did you consider a salinity correction term? See e.g. the HadCM3 runs in DeepMIP –Lunt et al (submitted to Climate of the Past).

We have not yet considered a salinity conservation term because the non conservation is linked to the LIM2 sea ice model. IPSL-CM5A2 will likely be upgraded to include the fully water-conservative LIM3 sea ice model in the future but, in the short term, we expect IPSL-CM5A2 to be mostly used for simulating warm climates, thereby reducing any potential salinity drift.

• Line 268: “Such a difference is also small enough compared to the regional biases in sea surface temperature.” How do you define “small enough”? Would 0.5 degrees by small enough? What about 0.8, or 1.0?! Better to compare with some other models, and just say, e.g. “this is closer to observations than X% of CMIP5 models”, or(and)link it to the uncertainty in the observations –0.3 degrees is probably within error of the observations?
We agree this sentence is unclear, and can be misleading. We have simply removed it.
• Supp Info could be used for some more plots, e.g. the old model version of Figure 4(d,e), the old model version of Figure 5(c).6(a,d) etc etc. I expect you have already made these plots!
Yes, we agree it would be valuable to have these plots in the appendixes, but the paper is already very long. Instead we refer to the online atlases comparing both versions of the model.

• Label all figures, not only in the caption but above the figure panels as well. Some plots are model minus observations, some model minus model, some just model. Clearere labelling would help (I know I could get this info from the caption but this is much slower than labelling the panels!)
Most of the figures have been modified to include labels.

• Fugure 15a –presumably this is the new model (not stated in the caption). Would be good to see this also for the old model.
This figure has been modified and now presents the anomaly between IPSL-CM5A2 and IPSL-CM5A both for the AMOC and the barotropic streamfunction.

• Line 595: would be good to show evolution of TOA flux, and show a Gregory plot.
We now include both plots in the appendixes section.
Section 5.4. This section implies that the model worked “first-time”, “out-of-the-box” for the new Cretaceous paleogeography. I would be surprised (and impressed!) if this were the case. If not, then it would be good to explain what needed to be done to allow the model to run, e.g. smoothing topography/bathymetry etc.

This section is intended to review the issues that were hampering the use of earlier versions of the IPSL model for deep-time simulations and to present the major steps required to set up a deep-time paleoclimate simulation in IPSL-CM5A2. Most of the locks preventing the model to run indeed concerned the grid structure (e.g. locations of the poles) and the boundary conditions.

As most other models, some tweaks are necessary for IPSL-CM5A2 to run with very different configurations. For instance, numerical instabilities require that the timestep of NEMO is slightly increased. In the atmosphere, warmer climates due to increased atmospheric CO2 levels regularly make the convection scheme crash. We usually circumvent these instabilities by adding white noise to surface pressure values in the restart files or changing atmospheric wave dissipation time in the model. By increasingly using IPSL-CM5A2 in a deep-time configuration, we have observed that this latest method greatly reduces the number of crashes but that the model is not tuned for running with these dissipation values. As a result, the Cretaceous simulation that we report here (being one of the first ever made with the deep-time version of IPSL-CM5A2) is 1 to 1.5°C warmer than the similar Cretaceous simulation reported by Laugé et al. (2019), which has been performed later and runs with appropriate dissipation values (see also response to reviewer #1). However, polar smoothing applied to topography/bathymetry was not needed for the model to run.
Suggestions

•Line 282: Presumably there is quite a “kick” to the model going from your PREIND forcings to the 1850 forcings; presumably it might have been a good idea to hold the model steady at 1850 CMIP5 forcings for a few decades before starting the 1850-2005 simulation?

Actually, the model was run for thousands of years in preindustrial conditions before. Then the restarts of PREIND simulation have been used to initialize the historical simulation.

•HadCM3 runs at about 100 SYPD these days (see Valdes et al, 2017). It might be nice to compare some metrics with them, in a consistent style e.g. Geckler plots (see Figure 2 in Valdes et al, 2017).

We agree and we think it could indeed be the scope of a dedicated paper, comparing the performance of fast low resolution climate models. For information we still provide a parallel coordinate graph comparing IPSL-CM5A2 metrics to other CMIP5 models, and highlighting HadCM3 as well. The variables are sorted to display the results by increasing order of performance for IPSL-CM5A (black line). IPSL-CM5A2 (blue line) displays improvements in temperature fields, whereas precipitation is slightly worsened. HadCM3 (CMIP5 version) is displayed in red, showing better performances in wind fields and precipitation, and worse in temperature. We did not have the values for HadCM3BL, which would be more relevant to compare to IPSL-CM5A2.

•In addition to the Cretaceous simulation with the new model grid, it would be very informative to re-do a preindustrial control with this new grid, to see how much of a role the modern-specific tweaks play.

We agree and plan to do this simulation in the near future.
Minor corrections
Line 8: “and” missing.
corrected.
Line 27: “which” should be “whose”?
rephrased.
Line 150: spell out “phytoplankton” and “zooplankton”.
corrected.
Figure 2 caption: “The surface of the disks” should be “The surface area of the disks”.
corrected.
Figure 4: label panels (d) and (e) as CRF and temp anomalies (just so we don’t have to read the
caption!).
Figure 19b. Black continental outline is missing for some coastlines.
corrected.
Line 580. Reference missing.
Now included.

Figure 21: why suddenly a new map projection?
We provide new versions of this figure (now figure 20), consistent with the projections used earlier in the manuscript.
I was surprised that Jean-Louis Dufresne was not a co-author, given his contributions to the original model?
J-L Dufresne has been of great help contributing to the revised version of the manuscript, and is now a co-author.

All the following suggested references are now cited in the text.


Abstract. Based on the CMIP5-generation previous IPSL earth system model, we designed a new version, IPSL-CM5A2, aiming at running multi-millennial simulations typical of deep-time paleoclimates studies. Three priorities were followed during the set-up of the model: (1) improving the overall model computing performance, (2) overcoming a persistent cold bias depicted in the previous model generation, and (3) making the model able to handle the specific continental configurations of the geological past. Technical developments have been performed on separate components and on the coupling system to speed up the whole coupled model. These developments include the integration of hybrid parallelization MPI-OpenMP in LMDz atmospheric component, the use of a new library to perform parallel asynchronous input/output by using computing cores as “IO servers”, and the use of a parallel coupling library between the ocean and the atmospheric components. The model, which runs with an atmospheric resolution of 3.75 ° x 1.875 ° and 2 to 0.5 ° in the ocean, can now simulate ~100 years per day, opening new possibilities towards the production of multi-millennial simulations with a full earth system model. The tuning strategy employed to overcome a persistent cold bias is detailed. The confrontation of an historical simulation to climatological observations shows overall improved ocean meridional overturning circulation, marine productivity and latitudinal position of zonal wind patterns. We also present the numerous steps required to run the IPSL-CM5A2 for deep-time paleoclimates through a preliminary case-study for the Cretaceous. Namely, a specific work on the ocean model grid was required to run the model
for specific continental configurations in which continents are relocated according to past paleogeographic reconstructions. By briefly discussing the spin-up of such a simulation, we elaborate on the requirements and challenges awaiting paleoclimate modelling in the next years, namely finding the best trade-off between the level of description of the processes and the computing cost on supercomputers.

1 Introduction

Despite the rise of high-performance computing (HPC, all acronyms except model names, are defined in Appendix D), the ever-growing complexity and resolution of General Circulation Models (and subsequent Earth System Models –ESMs-) have restricted their use to centennial integrations for future climate projections, or short “snapshot” experiments for paleoclimates, with durations ranging from decades to a few thousands of years. Longer simulations of several thousands of years are yet often desirable. Milennial-length simulations are useful to project very long-term future climatic trends, but also to properly quantify equilibrium climate sensitivity of models (Rugenstein et al., 2020). In paleoclimate studies, such long simulations are mandatory to either (i) reach a fully equilibrated deep ocean when initialized from idealized thermohaline conditions (a typical procedure in deep-time, pre-Quaternary -i.e. older than 2.6 million-years-, paleoclimate simulations) or (ii) address multi-millennial transient climate evolution such as glacial-interglacial cycles of the Quaternary period (He, 2011). Through years, several strategies have been set up to overcome the issue of the computing cost of long integrations. They involve both the development of dedicated models and peculiar experimental designs. Regarding the former, different models have been used:

- Earth System Models of Intermediate Complexity (Claussen et al., 2002), with coarse spatial resolution and simplifications in the physics, allow very efficient computation times and have been used extensively to explore Quaternary paleoclimates and run transient simulations (Roberts et al., 2014; Caley et al., 2014). As an example, the iLOVECLIM model (Goosse et al., 2010; Roche, 2013) provides more than 2,500 simulated years per day (hereafter SYPD) on one computing core at a ~5.6 ° spatial resolution (3 vertical levels) in the atmosphere and a full OGCM at 3 °x 3 ° (21 vertical levels) with two tracers. Performance is still 850 SYPD with 20 tracers in the ocean (Bouttes et al., 2015b; Missiaen et al., 2019).

- Versions of fully-coupled GCMs have been maintained at very low (~5 – 7 °) spatial resolutions, allowing long integrations in a reasonable amount of wallclock time. Examples include FOAM (Fast Ocean Atmosphere Model). It couples the NCAR CCM2 (7.5 ° x 4.5 ° and 18 vertical levels but an updated physics equivalent to the NCAR CCM3 model) atmosphere model and an ocean model similar to GFDL MOM. It reaches an average of 250 SYPD and has been routinely used for almost twenty years in numerous deep-time paleoclimate studies (Brown et al., 2009; Donnadieu et al., 2006; Ladant et al., 2014; Pohl et al., 2014; Poulsen et al., 2001; Roberts et al., 2014). FAMOUS (FAst Met Office/UK Universities Simulator), as part of the BRIDGE HadCM3 family of climate models (Valdes et al., 2017), is another example. It runs at a 7.5 ° longitude x 5 ° latitude resolution with 11 levels in the atmosphere and averages 450 SYPD, and that
is frequently used for paleoclimate (e.g. Gregoire et al., 2012; Roberts et al., 2014) and future climate studies (Bouttes et al., 2015a).

- Climate modelling groups have also maintained versions of their GCMs at a low (~2–4 °) resolution. Within the BRIDGE HadCM3 family, HadCM3BL (3.75 ° x 2.5 ° and 19 levels in the atmosphere; 3.75 ° x 2.5 ° and 20 vertical levels in the ocean) is used to perform long integrations at 85SYPD (e.g. Kennedy-Asser et al., 2020). NCAR has also developed a low-resolution (T31 –i.e. approximately 3.75° longitude x 3.75° latitude, 26 vertical levels, 35 SYPD) version of CCSM3 GCM (Yeager et al., 2006) that provided “the first synchronously coupled atmosphere-ocean general circulation model simulation from the Last Glacial Maximum to the Bølling-Allerød (BA) warming” (Roberts et al., 2014) as well as transient Holocene simulations (Liu et al., 2014). CCSM3 has been used also to run a transient simulation of the last deglaciation within the TraCE program (He, 2011). NCAR has released a similar version for CCSM4 (Shields et al., 2012), at the same “low resolution”, with performance reaching 64 SYPD. This model has been used for multi-millennial fully-coupled paleoclimate simulations (Brierley and Fedorov, 2016; Burls et al., 2017). A similar strategy has been applied with the ECHAM5/MPI-OM model (formerly known as COSMOS), which has been used at T31 resolution in the atmosphere for long paleoclimate simulations of the Holocene (Dallmeyer et al., 2017; Fischer and Jungclaus, 2011), and has been recently implemented with water isotopes (Werner et al., 2016).

Experimental designs dedicated to multi-millenia simulations have also followed different strategies through the years:

- Early methods to circumvent the constraint of computation time have involved (i) using atmospheric-only models forced by prescribed sea-surface temperatures (e.g. Roberts et al. (2014)), (ii) coupling the atmospheric component to mixed-layer ocean models with prescribed heat transport –i.e. slab ocean models- (Otto-Bliesner and Upchurch, 1997), and (iii) asynchronous coupling between atmospheric and ocean models (Bice et al., 2000). If efficient in terms of computation time, such strategies prevented any assessment of ocean dynamics and associated feedbacks on the climate system. Later, experimental designs with GCMs for deep-time paleoclimate modelling relied on simulations branched on each other (i.e. the final-state of equilibrated simulations is used as an initial state for new simulations with a different forcing) to make the ocean reach different equilibria in a reasonable amount of time (namely less than 1000 years, Liu et al. (2009)).

- For transient experiments, i.e. with an evolving insolation, Lorenz and Lohmann (2004) proposed an accelerated insolation forcing; when the simulation calendar advances by one year, the calendar for the computation of the incoming insolation advances by 10 to 100 years. The method assumes that the fast components of the climate system (atmosphere, upper ocean) are in constant equilibrium with insolation. It is suitable only for periods of time with almost no change for the slow components (deep ocean, land ice). This method is easily implementable in any climate model (e.g., Crosta et al., 2018). This acceleration technique has been used multiple times with CCSM3, the latest example being the study of the last 140,000 years of climate evolution of Africa by Kutzbach et al. (2020). Others have applied a “multiple snapshots” strategy, involving several multi-centennial (200 to 500 years) simulations forced by varying orbital parameters, to explore the impact of insolation on the fast components of the climate system (Singarayer and Valdés, 2010; Marzocchi et al., 2015).
Statistical approaches based on multiple simulations also allowed to explore transient climate changes. They involve either the use of statistical emulators, calibrated using several steady-state GCM simulations, with varying orbital configurations and atmospheric pCO2 concentrations (Lord et al., 2017), or linear approximations between single-forcing simulations (Erb et al., 2015; Simon et al., 2017).

The original IPSL-CM5A-LR (IPSL stands for Institut Pierre Simon Laplace. We will omit the -LR suffix in the following.) model was developed and released in 2013 “to study the long-term response of the climate system to natural and anthropogenic forcings as part of the 5th Phase of the Coupled Model Intercomparison Project (CMIP5)” (Dufresne et al., 2013). This earth system model coupled the LMDZ5 atmospheric model (3.75 x 1.875° in longitude-latitude, 39 vertical levels) (Hourdin et al., 2013) and ORCHIDEE land surface model (Krinner et al., 2005) with the NEMO nucleus (Madec, 2015), that includes ocean and sea-ice dynamics (2 degrees resolution and 31 vertical levels in the ocean), together with the marine biogeochemical model PISCES (Aumont et al., 2015). IPSL-CM5A has been used for several paleoclimate studies (e.g., Kageyama et al., 2013; Roberts et al., 2014; Tan et al., 2017; Zhuang and Giardino, 2012), including studies benefiting from the explicit representation of marine biogeochemistry (Bopp et al., 2017; Ladant et al., 2018; Le Mézo et al., 2017), vegetation dynamics and land biosphere (Tan et al., 2017). Still, IPSL-CM5A computation time, which averages 10 SYPD, has hindered its use for multi-millennial experiments that are typical for Quaternary or “deep-time” paleoclimate studies in which a fully-equilibrated deep ocean is mandatory. In this paper, we present and evaluate IPSL-CM5A2, a CMIP5-class ESM, including interactive vegetation and carbon cycle, and designed for multi-thousand years experiments.

2 From IPSL-CM5A to IPSL-CM5A2

Apart from better computing performances, setting up IPSL-CM5A2 aimed at reducing two of the main biases of IPSL-CM5A. First, an important cold bias has been depicted in global surface air temperature (t2m) in IPSL-CM5A (Dufresne et al., 2013). This cold bias is associated with a position of the atmospheric eddy driven jets too much equatorward beyond the spread of other CMIP5 models, especially in the Northern Pacific and Southern Hemisphere (Barnes and Polvani, 2013). Moreover, the Atlantic Meridional Overturning Circulation (AMOC) has maximum values between 8 and 10 Sv in historical simulations with IPSL-CM5A, a value lower than the observation range and below the values obtained in the other CMIP5 models (Zhang and Wang, 2013).

IPSL-CM5A2 development was initiated in 2017 on the supercomputer CURIE, operated at the Très Grand Centre de Calcul (TGCC), Bruyères-le-Châtel, France, as a part of the GENCI French computing consortium. CURIE offered different fractions of x86-64 computing resources, including 5040 “thin” nodes (BullX B510). Each node was made of two 8-core 2.7 GHz Intel Sandy Bridge processors, providing a total of 80 640 cores for computation. CURIE was replaced by the supercomputer JOLIOT-CURIE in July 2018. On this latter machine, IPSL models are run on a partition of 1 656 nodes, each node made of two 24-core 2.7 GHz Intel Skylake processors, providing a total of 79 488 cores for computation. Here we depict how IPSL-CM5A2 differs from IPSL-CM5A in terms of components (2.1) and technical characteristics (2.2). Then we present the computing performances of the new model, first on CURIE compared to IPSL-CM5A as it was originally conducted, then on...
Table 1. Characteristics of IPSL-CM5 and IPSL-CM5A2 models components.

<table>
<thead>
<tr>
<th>Model Name</th>
<th>Grid type and Resolution</th>
<th>IPSL-CM5A</th>
<th>IPSL-CM5A2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Atmosphere</td>
<td>LMDZ5</td>
<td>Longitude-latitude 3.75 ° x 1.875 °</td>
<td>revision 2063 June 2014</td>
</tr>
<tr>
<td>Land Surface</td>
<td>ORCHIDEE</td>
<td>ORCHIDEE 1.9_5 late 2011 Specific branch revision 6277 July 2018</td>
<td></td>
</tr>
<tr>
<td>Coupler</td>
<td>OASIS</td>
<td>OASIS 3 revision 3258 OASIS3-MCT 2.0 revision 4775 (IPSL server)</td>
<td></td>
</tr>
<tr>
<td>Ocean</td>
<td>NEMO</td>
<td>Tripolar curvilinear ORCA2 2 ° and 0.5 ° in the tropics 31 vertical levels version 3.2 version 3.6 revision 6665</td>
<td></td>
</tr>
<tr>
<td>Sea-ice</td>
<td>LIM2</td>
<td>Same horizontal grid as NEMO 2-level Identical</td>
<td></td>
</tr>
<tr>
<td>Marine biogeochemistry</td>
<td>PISCES</td>
<td>Same as NEMO PISCES-v1 PISCES-v2</td>
<td></td>
</tr>
</tbody>
</table>

JOLIOT-CURIE, the TGCC machine that will be used in the next years by the IPSL climate model community and on which we conducted a set of scaling experiments (A2).

2.1 IPSL-CM5A2 components: evolution from IPSL-CM5A

IPSL-CM5A2 is built on updated versions of IPSL-CM5A components, and is configured at the same spatial resolutions, i.e. the regular 3.75 x 1.875° longitude-latitude grid and 39 vertical levels for LMDZ, and the ORCA2 curvilinear grid for NEMO, which is 2 degrees with a refinement to 0.5° in the tropics, and 31 vertical levels (Fig.1 and table 1). This choice results from an early evaluation of the performances presented in Appendix A. Here we target the significant differences between the two models and provide a brief summary of each component. We refer the reader to Dufresne et al. 2013 for more detailed description of LMDZ, NEMO and ORCHIDEE characteristics.

2.1.1 LMDZ

LMDZ is the atmospheric general circulation model developed at Laboratoire de Météorologie Dynamique, France. LMD models have been the atmospheric component of the IPSL coupled models since the earliest IPSL coupled model, that included LMD5.3 (Braconnot et al., 1997). LMDZ is based on the coupling of a dynamical core in which primitive equations of meteorology are discretized, and a set of physical parameterizations. Here, we use the version LMDZ5A, where A designates the so-called “standard physics” (SP) for the physical atmospheric parameterizations, also used in IPSL-CM5A. The choice of keeping this CMIP5-like physical settings and not benefiting from improvements in the representation of convective boundary layer, cumulus clouds and the diurnal cycle of continental convective rainfall included in the “New Physics” (NP) is driven both
Figure 1. NEMO (a) and LMDZ (b) grid cell areas, in km$^2$, showing the refinements of the ocean grid in the tropics and over the Mediterranean Sea as well as the refinement of the atmospheric grid to the pole.

by (i) the fact that no coupled version of IPSL including the new parameterizations was available when IPSL-CM5A2 project was initialized and (ii) computing cost. The NP parameterizations in LMDZ have been validated for higher vertical resolution (79 levels) and require higher-frequency time coupling between the physics and the dynamics. SP indeed requires a time step for the dynamics of 3 minutes and a coupling every 30 minutes with the physics, whereas dynamics is computed every 2.15 minute and coupled every 15 minutes in NP, leading to significant increase in computation time (see next section). In terms of revision nomenclature, IPSL-CM5A included LMDZ5, revision 2063, whereas IPSL-CM5A2 includes revision 3342. This represents a >2-year gap between both versions. The differences mainly concern bug fixes and various improvements in the physical/dynamic interface in preparation for the next generation of models. They have very limited effect on model results and translate into a slight improvement in energy conservation.
2.1.2 ORCHIDEE

ORCHIDEE (Krinner et al., 2005) is a dynamic global vegetation model (DGVM) that can run either in stand-alone mode, coupled to LMDZ, or included as the land surface component in the IPSL climate model. ORCHIDEE is made of three parts: (i) the “hydrology module” SECHIBA (Schématisation des Echanges Hydriques à l’Interface entre la Biosphère et l’Atmosphère) (De Rosnay and Polcher, 1998; Ducoudré et al., 1993), (ii) parameterizations regarding vegetation dynamics based on plant functional types (PFTs) and derived from the LPJ DGVM (Sitch et al., 2003), and (iii) a “carbon module” called STOMATE (Saclay Toulouse Orsay Model for the Analysis of Terrestrial Ecosystems), that simulates vegetation phenology and carbon dynamics. SECHIBA “describes exchanges of energy and water between the atmosphere and the biosphere, and the soil water budget” (Krinner et al., 2005) with a time step of 30 minutes. In IPSL-CM5A and IPSL-CM5A2, SECHIBA includes a two-layer soil hydrology scheme, in which maximum water storage capacity is set globally to 300 kg m$^{-2}$ over a two-meter soil depth (Guimberteau et al., 2014). Water is redistributed between the two layers through a downward flux parameterized the early ideas of Choisnel (Choisnel et al., 1995; Ducharne et al., 1998). Rain falling from the canopy feeds the upper layer that loses water both by root extraction and soil evaporation, whereas water storage in the bottom layer decreases only as a function root extraction (Guimberteau et al., 2014). When total soil moisture storage reaches the maximum water storage, the excess water amount is converted to runoff. STOMATE links the fast hydrologic and biophysical processes to the slow processes of vegetation dynamics from LPJ. It includes formulations for photosynthesis, carbon allocation, leaf phenology, and respiration (autotrophic and heterotrophic). Vegetation is represented through 13 plant functional types (PFTs), one including bare soil. Each PFT is characterized by specific parameters controlling their dynamics, i.e. climatic control of vegetation establishment, elimination, light competition, fire occurrence and tree mortality. IPSL-CM5A2 includes the version 3930 of the ORCHIDEE model, whereas IPSL-CM5A was based on an older tagged version (ORCHIDEE 1.9.5, for the record). From the technical point of view, these two versions differ in that the IPSL-CM5A2 version of ORCHIDEE benefits from XIOS for output diagnostics and from hybrid (MPI-OpenMP) parallelization, as for LMDZ. Amongst several bug corrections, continental evaporation computation has been corrected to ensure a full closure of the water budget.

2.1.3 NEMO-LIM2-PISCES

NEMO has been upgraded from v3.2 to v3.6. River runoffs are now added through a non-zero depth, and have a specific temperature and salinity. Although a new version of the sea-ice model (LIM3, (Rouset et al., 2015)) has been developed and is included in the 6th version of IPSL model (IPSL-CM6A, Boucher et al. in preparation), the coupling with the ocean model on the ORCA2 mesh was not ready and we chose to keep the LIM2 configuration for IPSL-CM5A2. LIM2 (Fichefet and Maqueda, 1997; Timmermann et al., 2005) is a sea-ice model with a single sea-ice category. Open water is represented through ice concentration. As stated in (Uotila et al., 2017) who provide a detailed comparison between LIM2 and LIM3 in the NEMO, LIM2 implements the snow-ice formation by infiltration and freezing of seawater into snow when deep enough. The effect of subgrid-scale snow and ice thickness distributions is implicitly parameterised by enhancing the conduction of heat through the ice and by melting the ice laterally to account for thin ice melting. The surface albedo depends on the state of the
surface (frozen or melting), snow depth and ice thickness following (Shine and Henderson-Sellers, 1985). The biogeochemical component of NEMO is PISCES (Pelagic Interaction Scheme for Carbon and Ecosystem Studies) and has been upgraded from PISCES-v1 (Aumont and Bopp, 2006) to PISCES-v2 (Aumont et al., 2015) in IPSL-CM5A2. PISCES simulates the ocean carbon cycle and includes a simple representation of the lower trophic level ecosystem, with four plankton functional types (two for phytoplankton and two for zooplankton) and the cycles of the main oceanic nutrients (N,P,Si,Fe). Without any change for the general model structure (with 24 state variables), the transition from PISCES-v1 to PISCES-v2 does include a number of developments, e.g. on the iron cycle, on zooplankton grazing, on dissolved and particulate organic matter cycling (see Aumont et al. (2015) for a detailed description).

2.2 Technical developments

Technical developments have been performed on both individual components and the coupling system to accelerate the entire coupled model.

2.2.1 OASIS3-MCT coupling library and interpolation scheme for ocean-atmosphere exchanges

The ocean-sea ice-atmosphere coupling is close to the one used in IPSL-CM5A (Dufresne et al., 2013), except the coupling system has been switched from OASIS3.3 to OASIS3-MCT (for Model Coupling Toolkit, Valcke, 2013), that constitutes a parallel coupling library used in both NEMO and LMDz components. It ensures fully parallel interpolation and exchange of the coupling fields. As in IPSL-CM5A, 24 coupling fields are exchanged, 7 from the ocean to the atmosphere and 17 from the atmosphere to the ocean (Valcke, 2013).

2.2.2 XIOS library

The use of XIOS as input-output library allows performing parallel asynchronous input/output by using computing cores as “IO servers”. These servers are dedicated to the reading and the writing of the data and permit model computation to continue while data is written or read.

2.2.3 Integration of hybrid parallelization MPI-OpenMP in LMDz atmospheric component

In LMDz, longitudinal filtering is applied on the dynamical equations for latitude higher than 60° in both hemispheres (Hourdin et al., 2013). Thus the choice of MPI domain decomposition is optimal only along latitudinal bands. This decomposition has been initiated in the LMDz version of IPSL-CM5A with a minimum of three latitude bands per MPI process, reduced to two in IPSL-CM5A2. In addition, the use of multi-core processors in HPC led us to add a shared memory parallel processing (OpenMP) in LMDz. The use of this shared memory parallelism on vertical levels in the dynamics allows to increase the number of cores used on the supercomputer and consequently to reduce the elapsed time of the simulation (see Appendix A).
3 Tuning IPSL-CM5A2

IPSL-CM5A2 is evaluated through two sets of simulations. First, we describe the setup of a steady-state pre-industrial simulation that involved several experiments with tuning steps and bug fixes. This setup is used to discuss the tuning strategy, the model large-scale characteristics, ocean spinup, energy and freshwater budgets, as well as multi-centennial ocean variability. Second, a transient “historical” simulation, forced by boundary condition between years 1850 and 2005, has been designed to evaluate the model with respect to observations (section 4).

3.1 Tuning: target and strategy

We focus here on the general tuning strategy adopted by defining the target and how we reached it, rather than giving a comprehensive description of the numerous simulations that have been designed to obtain the final pre-industrial simulation. As depicted in Hourdin et al. (2017), several choices need to be made when tuning a model, namely defining a relevant target and associated metrics, defining in what mode the model is to be tuned (e.g. atmospheric only, fully coupled, etc.), and choosing which parameters will be used to reach the target. As most of modeling groups (Hourdin et al., 2017) our choice was to tune IPSL-CM5A2 through pre-industrial runs, using alternatively atmospheric and fully-coupled experiments. A first untuned 1,000-year spin-up of IPSL-CM5A2 forced by CMIP5 pre-industrial boundary conditions provided an equilibrated global surface air temperature of 11.3°C. Such a cold bias had been depicted in IPSL-CM5A (Dufresne et al., 2013). We targeted to increase global-mean surface temperature by 2.2°C to reach 13.5°C in pre-industrial conditions with IPSL-CM5A2, expecting this value to translate into 14.5°C for a present-day simulation, a value consistent with observations (Dee et al., 2011).

Regarding the free parameters, the choice was to use a parameter which controls the conversion of cloud water to rainfall to alter the cloud radiative forcing (CRF), and thereby the TOA net radiation ($Q_{TOA}$, counted positive downward) and the global-mean surface temperature. Using clouds parameters for tuning, the most uncertain parameters that affect the most radiation, is also well shared practice among modeling groups. Previous results obtained with LMDZ and IPSL-CM5A show that changing TOA balance by 1 W.m$^{-2}$ results in a change in temperature of 1 °C. It is also the typical value of the sensitivity of global temperature to greenhouse gas concentration. Our +2.2°C target thus translates into an increase of $Q_{TOA}$ by 2.2 W.m$^{-2}$. In LMDZ, the conversion of cloud water to rainfall is computed following Sundqvist (1978), as

$$\frac{dq_{lw}}{dt} = \frac{q_{lw}}{\tau_{conv}} \left[ 1 - e^{-\left( \frac{q_{lw}}{c_{lw}} \right)^2} \right]$$

(1)

where $q_{lw}$ is the mixing ratio, $c_{lw}$ is the in-cloud water threshold for autoconversion, set at 0.418 g.kg$^{-1}$ in IPSL-CM5A.

For $q_{lw}$ $\gg$ $c_{lw}$ the time constant for auto-conversion is $\tau_{conv}$ (here set at 1800 s) while it goes to infinity (no auto-conversion) for $q_{lw}$ $\ll$ $c_{lw}$.
Decreasing $c_{lw}$ is expected to lower cloud water content and reduce the CRF. First, we carried out atmospheric-only simulations with varying $c_{lw}$ values to define the sensitivity of $Q_{TOA}$ to $c_{lw}$. The resulting linear relationship provided a value of 0.325 g.kg$^{-1}$ for $c_{lw}$ to obtain an increase of 2.2 W.m$^{-2}$ in $Q_{TOA}$. Thus we ran a second pre-industrial fully-coupled simulation of 500 years with this tuning. This experiment was interrupted when two bugs (The first was an erroneous computation of the weights necessary to exchange fluxes between the ocean and the atmosphere introduced in the IPSL-CM5A2 versions of OASIS and ORCHIDEE, and the second was a water conservation issue in the land surface scheme) were identified and fixed. A second set of tuning was required as the bug correction altered the global 2-meter temperature. After several adjustments, the target of +2.2 W.m$^{-2}$ at TOA was expected to be obtained by setting $c_{lw}$ at 0.343 g.kg$^{-1}$. A ultimate fully-coupled experiment (called here PREIND) including this tuning was then branched on the previous simulation and run for 2,800 years.

3.2 Energetic and freshwater balances, equilibrium and drifts

Table 2. Global yearly averages of the radiative budget terms (in W.m$^{-2}$) for PREIND last 100 years and historical experiment for the 1980–1999 period. Estimated values come from Trenberth et al. (2009), based on the ERBE period (1985–1989) or on the CERES period (2000–2004), and retrieved from Voldoire et al. (2013).

<table>
<thead>
<tr>
<th></th>
<th>PREIND</th>
<th>HISTORICAL</th>
<th>ESTIMATES</th>
</tr>
</thead>
<tbody>
<tr>
<td>Net solar radiation at TOA</td>
<td>239.9</td>
<td>242.7</td>
<td>224-244</td>
</tr>
<tr>
<td>Outgoing LW radiation at TOA</td>
<td>239.8</td>
<td>242.1</td>
<td>233-253</td>
</tr>
<tr>
<td>$Q_{TOA}$</td>
<td>0.029</td>
<td>0.537</td>
<td>-</td>
</tr>
<tr>
<td>Net solar radiation at surface</td>
<td>171.3</td>
<td>172.7</td>
<td>155-170</td>
</tr>
<tr>
<td>$Q_{surf}$</td>
<td>0.023</td>
<td>0.58</td>
<td>-</td>
</tr>
<tr>
<td>Net LW radiation at surface</td>
<td>-67.29</td>
<td>-66.73</td>
<td>-(48-72)</td>
</tr>
<tr>
<td>Incoming SW at surface</td>
<td>197.7</td>
<td>198.7</td>
<td>-</td>
</tr>
<tr>
<td>Outgoing SW at surface</td>
<td>26.38</td>
<td>25.98</td>
<td>16-45</td>
</tr>
<tr>
<td>Incoming LW at surface</td>
<td>326.5</td>
<td>332.6</td>
<td>324-345</td>
</tr>
<tr>
<td>Outgoing LW at surface</td>
<td>393.8</td>
<td>399.4</td>
<td>390-396</td>
</tr>
<tr>
<td>Sensible heat flux</td>
<td>-22.98</td>
<td>-22.65</td>
<td>-(15-24)</td>
</tr>
<tr>
<td>Latent heat flux</td>
<td>-81</td>
<td>-82.76</td>
<td>-(78-90)</td>
</tr>
</tbody>
</table>

As expected, the imposed tuning induces a decrease in CRF in our preindustrial run (-18.97 W.m$^{-2}$ before tuning; -16.80 W.m$^{-2}$ after tuning, see section 4.1 for more details about CRF). Table 2 shows the terms of the energetic balance averaged over the last decade of simulation. After 2800 years of integration, preindustrial $Q_{TOA}$ and $Q_{surf}$ are stabilized to values close to zero (respectively 0.029 W.m$^{-2}$ and 0.023 W.m$^{-2}$), showing an acceptable energetic imbalance and limited spurious energy excess or deficit. Energy conservation is not fully ensured neither in the NEMO nor in the LMDZ component of the model, but the global numerical leak is small. The non-conservative term, expressed as the difference between $Q_{TOA}$ and the ocean heat content (OHC) change $\frac{d\text{OHC}}{dt}$ (Hobbs et al., 2016), is small (0.094 W.m$^{-2}$ over the last 1000 years) compared to IPSL-
CM5A (0.18 W.m\(^{-2}\)), which might be an improvement linked to corrections of energy conservation between IPSL-CM5A and IPSL-CM5A2 versions of LMDZ (see section 2.1). Figure 2a indicates that this imbalance is constant after \(~300\) years of preindustrial simulation, and that OHC stabilizes after \(~2000\) years of simulation. This ensures that the model drift is small, therefore validating the use of IPSL-CM5A2 for multi-millennial simulations.

It takes about 500 years for sea surface temperatures and 2-meter air to stabilize after PREIND initialization (Fig. 2b). They converge to 17.5 °C and 13.2 °C, when globally averaged, respectively. We thus missed the 13.5 °C target by 0.3 °C which is reasonable when considering the very simple procedure used. Such a difference is also small enough compared to the regional biases in sea surface temperature. An additional 1,500 years is required for the global average ocean temperature to stabilize, as deeper ocean temperatures still cool by \(~0.1\) °C to 0.2 °C during this period (Fig. 2c).

Over the entire PREIND experiment, sea-surface height (SSH) depicts a constant positive drift of 0.19 m / century, revealing that the freshwater budget is not fully closed (not shown). This issue already existed in previous versions of the model, and is partly linked to the absence of a proper freshwater conservation between ocean and sea-ice components of NEMO. As a consequence, a linear decrease in global ocean salinity is simulated, corresponding to a steady global ocean freshening of \(2 \times 10^{-3}\) psu / century. It appears that this drift varies with global sea-ice volume, as shown by abrupt 4xCO2 simulations that are characterized by a drift in SSH reduced to 0.01 m / century and a corresponding weaker freshening of the global ocean (4.7 \(\times 10^{-4}\) psu / century, not shown). Such a drift might be problematic in the future if IPSL-CM5A2 is to be used for very long (transient-like) integrations. Improvements on this concern are expected from the future replacement of LIM2 by the more recent LIM3 version (Rousset et al., 2015), for which conservation is ensured.

### 4 Comparison of IPSL-CM5A and IPSL-CM5A2 with observations

An historical experiment was performed starting from the last year of the 2800-year PREIND spin-up. This transient simulation was integrated for 155 years between 1850 and 2005. Boundary conditions are described in (Dufresne et al., 2013). Yearly evolution of long-lived greenhouse gases concentrations and total solar irradiance (TSI) correspond to the historical requirements of the CMIP5 project, with notably reduced TSI to consider the volcanic radiative forcing based on an updated version of Sato et al. (1993). Tropospheric aerosols, tropospheric and stratospheric ozone were prepared as described in (Szopa et al., 2013) and land use changes processed from crop and pasture datasets developed by Hurtt et al. (2011), as described in (Dufresne et al., 2013). Unless otherwise expressed, comparison between the historical run and observations is made by using a mean seasonal cycle for the period 1980-2005. The comparison between simulated and observed cloud radiative forcing is made using the Earth’s Radiant Energy System (CERES) Energy Balanced and Filled Fluxes (EBAF) gridded dataset over the period 2000-2004 (Loeb et al., 2009).
Figure 2. Time-series of globally averaged variables over the 2,800 years of the PREIND experiment. a) annual-mean heat content anomaly \((10^{24} \text{ J})\) implied by \(Q_{\text{TOA}}\) (orange line, cumulative values) and OHC (blue line). As in (Hobbs et al., 2016), dashed black lines show the heat content implied by a \(Q_{\text{TOA}}\) of \(\pm 0.5 \text{ W.m}^{-2}\), the observed late-twentieth-century \(Q_{\text{TOA}}\) (Roemmich et al., 2015). b) Sea-surface temperature (orange) and 2-meter air temperature (blue). c) 1000-m (orange) and 500-m (blue) ocean potential temperature.
Figure 3. Zonally-averaged long-wave (a), shortwave (b) and total (c) cloud radiative forcing (in W.m$^{-2}$) from IPSL-CM5A2 and IPSL-CM5A historical simulations, compared to the Clouds and the Earth’s Radiant Energy System (CERES) gridded dataset (Loeb et al., 2009). (d) Yearly-averaged total CRF anomaly between IPSL-CM5A2 and CERES data. As CRF is a negative value, positive (negative) anomaly indicates an underestimation (overestimation) of the absolute value of the CRF in the model. (e) Yearly-averaged difference in near-surface air temperature (°C) between IPSL-CM5A2 and ERA-Interim (Dee et al., 2011). Grey contours indicate surface elevation.

4.1 Atmosphere mean state and identified biases

4.1.1 Cloud radiative forcing

LMDZ5A (and hereby IPSL-CM5A) has been shown to underestimate low and mid-level cloud cover and overestimate high-level clouds at mid and high latitudes (Hourdin et al., 2013), biases shared with many climate models (Zhang et al., 2005). LMDZ5A cloud representation has been since strongly improved with the new set of physical parameterizations and associated increase in vertical resolution (Hourdin et al., 2020) that is currently being used for CMIP6 project. In IPSL-CM5A2, the biases of the standard parameterization translate into CRF, with too reflective clouds and therefore too negative shortwave CRF at mid and high latitudes (-15 to -25 W.m$^{-2}$), and in some convective regions of the tropics an underestimation of outgoing longwave trapping and therefore too positive LW CRF (+5 to +15 W.m$^{-2}$), when compared to the CERES dataset. The tuning depicted earlier leads to a less negative total CRF at all latitudes compared to IPSL-CM5A (Fig. 3).

The too negative CRF in the southern hemisphere mid-latitudes is only slightly corrected, as less negative shortwave (SW) CRF
and is partly compensated by less positive longwave (LW) CRF (Fig. 3abc). In the southern tropics, the tuning mostly worsens the model biases in SW and LW CRF, in relation with a feedback loop in the continental tropical areas (namely the Amazon), where convection is reduced. Tuning still improves total CRF between 15°N and 40°N, driven by the better estimation of SW CRF over the oceans. The positive bias of CRF over the eastern boundaries of Atlantic and Pacific oceans remain unchanged (Fig. 3d), and is partly related to a bad representation of low cloud over these regions (Hourdin et al., 2015). These biases translate into surface air temperature biases (Fig. 3e). Regions with positive bias of CRF (e.g. the eastern boundaries of the Pacific and the south Atlantic) are associated with up to 2.5 °C warm biases. Conversely, regions with negative bias of CRF, like the northern and southern Atlantic and several mountain ranges (the Andes, the Rockies and the Tibetan Plateau) show cold biases (down to -6 °C for the Tibetan Plateau).

4.1.2 Temperature and general circulation

The tuning of IPSL-CM5A2 induces a low latitude warming whose amplitude varies with height, from +1°C at 850 hPa to +2.5 °C at 200 hPa, when compared to IPSL-CM5A, as expected from the adjustment of the moist adiabatic lapse rate (Fig. 4a). As a consequence of the increased the latitudinal temperature gradient between 200 and 300 hPa, the zonal jets expand towards high latitudes, especially in the Northern Hemisphere (Fig. 4b). This is an improvement, as IPSL-CM5A had been shown to underestimate the jets latitudinal extension (Barnes and Polvani, 2013). Despite this improvement, IPSL-CM5A2 still underestimates the jet extension in the Southern Hemisphere, as shown by a negative anomaly in zonal wind in the high latitudes and an overestimation of the zonal winds in the low latitudes. At 850 hPa and surface, IPSL-CM5A2 provides a better representation of the zonal winds in the high latitudes: the negative anomaly with ERA-Interim is reduced along the Antarctic, showing that the westerlies have moved southward. The 850 hPa temperature anomalies also show that tuning has been effective in IPSL-CM5A2 to reduce the cold bias of the model, but also that the warm bias over the eastern tropical Pacific has strengthened. In the tropics, IPSL-CM5A2 depicts biases similar to IPSL-CM5A, i.e. over the Pacific Ocean, surface easterlies bad seasonal cycle lead to an overall yearly misposition, in relation with the double Intertropical Convergence Zone (ITCZ), and over the Atlantic, easterlies strength is systematically underestimated (not shown). IPSL-CM5A2 also shows overestimated winter westerlies in the northern mid-latitudes, that translate into an anomalous wind stress over the northern Atlantic.

Sea-level pressure (slp) and surface wind patterns simulated by IPSL-CM5A2 are compared to ERA-Interim reanalysis in Fig. 5. The major circulation structures, such as the subtropical Highs of the Pacific and Atlantic, the December-January-February (DJF) Atlantic mid-latitude Low, the June-July-August (JJA) easterlies reversal in the monsoon regions, are well-captured by the model. DJF westerlies are overestimated over the Atlantic. Conversely, the model reproduces well the slp patterns at low latitudes in JJA. Simulated slp values for JJA are also underestimated between 40°S and 60°S and overestimated along the Antarctic coastline, in line with the zonal wind bias depicted earlier. As a consequence, westerlies velocity maxima are restricted to 45°S instead of 50°S in the reanalysis (Fig. 5cf). The southern eastern Pacific is an exception to this general pattern, as sea level pressures are overestimated between 45°S and the Antarctic coastline over this region, associated with very low surface winds velocity values. Over western Africa, the monsoonal surface winds intensity and position are well-simulated,
Figure 4. Zonally-averaged annual anomalies between: a) IPSL-CM5A2 and IPSL-CM5A temperature, b) IPSL-CM5A2 and IPSL-CM5A zonal wind, c) IPSL-CM5A2 and ERA-Interim (Dee et al., 2011) zonal wind. In each plot, contour lines indicate absolute values for IPSL-CM5A2, IPSL-CM5A2 and ERA-Interim, respectively.
in line with the good positioning of the Sahara Thermal Low (Fig. 5de). Both winter and summer Asian monsoon surface wind patterns are correctly positioned as well, but their intensity is underestimated. Overall, the model overestimates the strength of surface easterlies, both in winter and summer. This is particularly visible over the Sahara region in DJF (Fig. 5ab) and the northern part of South America in JJA (Fig. 5de).

### 4.1.3 Tropical precipitation

Regarding precipitation, IPSL-CM5A2 and IPSL-CM5 essentially share the same qualities and biases, i.e. an underestimation of rainfall in the mid-latitudes (20°-40°) of both hemispheres, an overestimation of rainfall in the southern tropics, characterized on the yearly zonal average by a peak at ~10 °S corresponding to a “double ITCZ” (Fig. 6e). Dry biases (-0.5 to -2 mm.d⁻¹) are marked over the western edges of the northern and southern Atlantic at mid-latitudes, as well as the northwestern Pacific in JJA. The regional anomaly between IPSL-CM5A2 and GPCP data also show that the tropical biases vary with longitude: namely rainfall is overestimated over the equatorial western Indian, equatorial Africa and south equatorial Atlantic (+1 to +3 mm.d⁻¹) in DJF (Fig. 6ac), whereas there is a strong underestimation of tropical rainfall over South America, specifically over the Amazon basin (down to ~ 5 mm.d⁻¹).
Figure 6. Seasonally-averaged precipitation (mm/day) for IPSL-CM5A2 (a,c) and anomaly between IPSL-CM5A2 and GPCP data (b, d). Top is DJF, bottom is JJA. e) Zonal average of absolute annually-averaged precipitation (mm/day) for IPSL-CM5A2, IPSL-CM5A and GPCP data.

South America (Fig. 7). The dry bias over South America and every region of the Amazon basin is a long-standing feature of the IPSL coupled models that likely has multiple origins, as precipitation over this region depends both on moisture advection from the Atlantic (and thus from the capability of the model to reproduce correct atmospheric dynamics and SSTs) and continent recycling of water through evapotranspiration. Dufresne et al. (2013) showed that soil depth had to be increased from 2 to 4 meters in IPSL-CM5A to favor seasonal water retention in the soil, allow vegetation to grow, and thereby reduce a too strong dry bias over the Amazon. We kept this tuning in IPSL-CM5A2. Still, the seasonal 850 hPa wind divergence shows that a strong dipole between restricted convective zones and expanded subsiding zones is simulated over tropical South America (Fig. B2). This pattern induces restricted areas with strong (>8 mm.d⁻¹) convective precipitation (northwestern Amazon in JJA, south-central Amazon in DJF) connected to larger zones where rainfall is limited by subsidence. The increase in the dry bias over South America between IPSL-CM5A and IPSL-CM5A2 results from a feedback loop linked to vegetation change. The tuning of the clouds produces excessive warming (Fig. 3), especially over the northwestern Amazon basin, that combines with DJF subsidence to strongly reduce rainfall amounts and thereby tropical rainforest cover (not shown). In line with earlier deforestation experiments carried out with the IPSL model (Davin and de Noblet-Ducoudré, 2010), such a decrease in tree cover has a strong feedback on temperature, as it leads to a decrease of latent heat flux and increase in sensible heat fluxes (overpassing the albedo increase, that tends to cool surface air temperature). As an illustration, the 2-meter temperature difference between IPSL-CM5A2 and IPSL-CM5A reaches + 3.5°C over northwestern Amazonia, with IPSL-CM5A2 annual mean two-meter temperature reaching 29.9 °C (not shown). Vegetation decrease also further reduces water recycling through lower evapotranspiration as confirmed by a supplemental historical experiment with IPSL-CM5A2 initialized with observed
values of vegetation rather than the results from PREIND that shows that prescribing rainforest over the Amazon basin cools two-meter temperature by 1 to 3 °C and increase annual rainfall amount by 2 mm.d$^{-1}$ (not shown). The seasonal cycles of precipitation computed over the northern and southern parts of the Amazon basin show the simulated rainfall collapse during the respective hemispheric winters, while the year-round underestimation of rainfall in the western part of the basin seems to related to a mislocation of rainfall, that occur on top of the Andes instead of the foothills.

**Africa (Fig. 8).** Conversely, precipitation rates are overestimated over equatorial Africa. The divergence of 850 hPa wind shows that the model overestimates convergence of moist air masses south of the equator in DJF (Fig. B2). Moreover the +0.5 to +2 °C anomalies in the eastern tropical Atlantic SST likely favor enhanced evaporation and moisture advection over the continent. Over central Africa, rainfall seasonal cycle is bimodal with two humid seasons, but the maxima lag by one month when compared to data. More importantly, rainfall amounts are largely overestimated by $\sim +66\%$ during the humid seasons. Over the western African monsoon region, IPSL-CM5A and IPSL-CM5A2 both lag the start of the rainy season, leading to an overall underestimation of rainfall before the monsoon peak. However this rainfall maximum occurs in August as for observations,
Figure 8. Rainfall patterns over Africa. The map shows the precipitation annual mean (mm/day) simulated in IPSL-CM5A2 historical experiment. Inserts on the sides of the map show the seasonal cycle over the Congo basin and the west African monsoon region for IPSL-CM5A, IPSL-CM5A2 and GPCP data.

and the subsequent decrease in simulated correctly.

Indian monsoon and Indonesia (Fig. 9). Compared to IPSL-CM5A, the Indian monsoon has been improved in IPSL-CM5A2 (Fig. 9ab). The 1-month lag that characterized IPSL-CM5A is still present, but the amplitude of the seasonal cycle of precipitation, although still underestimated, has been enhanced, peaking at 7.3 mm.d$^{-1}$ in August (5.6 mm.d$^{-1}$ in August in IPSL-CM5A), closer to observations (9.3 mm.d$^{-1}$ in July in GPCP data). This improvement is linked both to higher large-scale and convective precipitation during summer, consecutive of the enhanced hydrological cycle linked to the tuning-induced warming. However, the artificially high precipitation rates on the Himalayan foothills have not been corrected. They are made mostly of large-scale precipitation that are overestimated due to a too strong cooling over the highest altitude grid points.

Fig. 9c shows Hovmöller diagrams for GPCP dataset compared to IPSL-CM5A2. Although at the first order, the seasonal cycle of precipitation is correctly represented over the “Maritime Continent”, IPSL-CM5A2 depicts a strong humid bias over the
region (up to +5 mm.d$^{-1}$ during the humid season). This bias has been described earlier with LMDZ pre-industrial simulations run either in atmospheric-only or fully-coupled mode (Toh et al., 2018; Sarr et al., 2019) and it was shown that LMDZ systematically overestimates rainfall over both land and sea surfaces of the Maritime Continent, together with western Indian Ocean and Bengal Bay. Given the complexity of the Indonesian region, with a myriad of islands that the model spatial resolution cannot capture, it is no surprise that rainfall are misrepresented in this model (Qian, 2008).
4.2 Ocean mean state and identified biases

The annual SST averaged between 50°S-50°N simulated by IPSL-CM5A2 for the last decades of the historical run is 22.3 °C (1 °C warmer than IPSL-CM5A), a value in the higher range of CMIP5 ESMs and consistent with observations (Fig. B1). The warm bias over eastern subtropical Pacific and Atlantic oceans is more marked in IPSL-CM5A2 than in IPSL-CM5A as a response to CRF tuning (Fig. 10ab). This bias has already been reported in several coupled models using NEMO, namely CNRM-ESM1 (Séférian et al., 2016), CNRM-CM5.1 (Voldoire et al., 2013), IPSL-CM4 (Marti et al., 2010) and IPSL-CM5A (Dufresne et al., 2013). As stated in Voldoire et al. (2013), ocean-only experiments (Griffies et al., 2009) have suggested that such a warm bias was likely linked to “poorly resolved coastal upwellings and underestimated associated westward mass transport due to the coarse model grid resolution”. The 0.5-2 °C warm bias over the Antarctic circumpolar current (ACC) is comparable to IPSL-CM5A and can be related to the poor simulation of the CRF over this region (Hyder et al., 2018) and the associated mispositioning of mid-latitude westerlies depicted above. Conversely, the tuning has reduced the cold bias in the Pacific and Atlantic gyres. Surface salinity biases are rather similar between both models. The surface salinity anomaly follows the E-P pattern over the tropical oceans (not shown), showing the influence of the double ITCZ issue on the tropical ocean freshwater balance. Both models depict too fresh surface waters in the northern Atlantic and locations of deep water
formation (Fig. 10cd). The ocean also depicts temperature biases at depth both in the Atlantic and Pacific oceans (Fig. 11). Zonally-averaged potential temperature anomalies show that the warm bias in the northern Atlantic between 1000 and 3500 m has been amplified, as well as the warm bias in the northern Pacific, that concerns the entire water column in IPSL-CM5A2. The strong (> + 2.5 °C) overestimation of temperatures in the subsurface waters of the southern hemisphere is almost identical in both versions of the model. The cold (< – 1 °C) bias in the high latitudes of the southern hemisphere has been slightly reduced, but is still present due to the lack of strong enough overturning in this region. Sea-ice extent (Fig. 12) has been improved in the North Atlantic sector, but remains overestimated. The Nordic Seas are covered by sea ice in winter in IPSL-CM5A, while IPSL-CM5A2 shows overestimated sea ice mostly in the Barents Sea. In the Labrador Sea, both IPSL-CM5A and IPSL-CM5A2 have overestimated sea ice, likely preventing water mass sinking and convection in this region. In contrast, sea ice is underestimated in both versions for the North Pacific sector, i.e. in the Bering and Okhotsk seas.

### 4.2.1 Ocean Heat Transport and Meridional Overturning Circulation

IPSL-CM5A2 ocean meridional heat transport is higher (in absolute value) than the one simulated with IPSL-CM5A in both hemispheres (Fig. 13). This increase in the Northern Hemisphere is mainly related to the increase in the Atlantic meridional heat transport in the mid-latitudes (not shown). When compared to observations, the two model versions have asymmetric biases, with the northward transport being underestimated in the Northern Hemisphere and the southward transport overestimated in the Southern Hemisphere low latitudes (between the equator and 35°S). Southward heat transport is too strong between the equator and 40°S and is in better agreement with data than IPSL-CM5A between 40°S and 60°S. The Atlantic Meridional Overturning Circulation (AMOC) is more vigorous in IPSL-CM5A2, with a maximum in depth (below 500 m) and latitude (from 30°S to 60°N) in preindustrial conditions moving from 10.3 ± 1.2 Sv (Escudier et al., 2013) to 12.02 ± 1.1 Sv (taken for the last 1000 years of simulation, Fig. 14a). The last 20 years of the 20th century in the historical run depicts a less vigorous circulation in the Atlantic, with a mean value for AMOC index of 10.5 ± 0.9 Sv. The convection sites have been slightly
modified between the two versions of the model. While the strongest convection was located in the North Atlantic subpolar gyre (SPG), south of Iceland, in the IPSL-CM5A version (Escudier et al., 2013), in IPSL-CM5A2, the Greenland sea shows the deepest mixed layer depth, which is larger than 1500 meters in winter (Fig. 12cd). Nevertheless, there is still a very active deep convection site south of Iceland, which is not realistic. The strengthening of the Nordic Seas convection can be related with the decrease of its sea-ice cover in winter (Fig. 12). As a consequence, denser water is now produced in this site, which improves the overflow through Greenland-Iceland-Scotland straits. Indeed, the water denser than 27.8 kg.m$^{-3}$ flowing southward reaches 1.5 Sv in IPSL-CM5A2, while it was only 0.2 Sv in IPSL-CM5A. It remains largely underestimated as the observations from (Olsen et al., 2008) indicates around 6 Sv of overflow in total from observation-based estimates over the last few decades. Besides, the poleward shift of the westerlies in the North Atlantic stimulate a northward shift of the boundary between the Atlantic subtropical and subpolar gyres, as shown by the barotropic streamfunction anomaly (Fig. 14b). This may contribute to bringing more salt into the subpolar gyre, which may also act to intensify the AMOC in IPSL-CM5A2.

**Figure 12.** a,b. March sea-ice cover (%) for IPSL-CM5A (left) and IPSL-CM5A2 (right) historical runs. Black contour indicates the 15% sea-ice cover interval in observations. c,d. January-February-March average of mixed-layer depth (meters) for IPSL-CM5A (left) and IPSL-CM5A2 (right).
4.2.2 Fluxes in the major ocean gateways

We computed oceanic mass transport through selected longitudinal and latitudinal cross-sections corresponding to the major gateways (Table 3), and compared them to IPSL-CM5A, estimates from observations, and values simulated by CNRM-CM5.1, that uses NEMO as well (Voldoire et al., 2013). These diagnostics highlight that the Antarctica Circumpolar Current (ACC) is underestimated by IPSL-CM5A2 by around 25%, as for the other configuration considered here, whereas despite the absence of high resolution and explicit resolution of Indonesian gateways, the overall Indonesian throughflow tends to be close to observation-based estimate (underestimation of 16%). Compared to observations and CNRM-CM5.1, the flow in the Florida strait is quite weak (weaker than observation by 37%), in line with the simulated “sluggish” AMOC. The increase of the AMOC
Figure 14. Atlantic meridional streamfunction (a) and global barotropic streamfunction (b) for the preindustrial experiments at steady-state (100-year averages). Color shading shows the anomaly IPSL-CM5A2 minus IPSL-CM5A. Contour lines shows the IPSL-CM5A2 values (in Sv).
from IPSL-CM5A to IPSL-CM5A2 has not been enough to strengthen this current, also largely resulting from wind forcing. Also, the absence of resolved eddies can also play a significant role in this underestimation.

4.2.3 Net Primary Production (NPP)

To illustrate some of the capabilities of IPSL-CM5A2, we also evaluate the simulated oceanic NPP (Fig. 15). The global integrated NPP over the historical period (here averaged over 1980-1999 from the historical simulation) is 47.5 PgC.y\(^{-1}\), which is a strong increase when compared to 30.9 PgC.y\(^{-1}\) simulated for IPSL-CM5A (Bopp et al., 2013), making IPSL-CM5A2 being closer to the 52.1 PgC.y\(^{-1}\) global estimate based on SeaWifs remote-sensing observations (Behrenfeld and Falkowski, 1997). The representation of NPP at the spatial scale compares well to observation-based estimates for the tropical and Southern Hemisphere, but depicts large biases in the Northern Hemisphere, especially in the North Atlantic where NPP is largely under-estimated (Fig. 15c).
4.2.4 Ocean variability

The main modes of variability of the AMOC are analyzed with an empirical orthogonal function (EOF) of the yearly Atlantic meridional overturning streamfunction (Fig. 16, top). The first EOF is a monopole in IPSL-CM5A and IPSL-CM5A2, but IPSL-CM5A2 shows less loading in the subpolar gyre between 40°N and 60°N. The associated principal component shows a significant 20-yr variability in the case of IPSL-CM5A. Such variability was previously linked to advection of surface salinity anomalies from the western boundary to the Eastern Atlantic (Escudier et al., 2013), and westward propagating planetary waves in subsurface (Ortega et al., 2015). North Atlantic observations (Vianna and Menezes, 2013; Swingedouw et al., 2013) and proxy records indicate a 20-year preferential variability in this region in the atmosphere (Chylek et al., 2011), sea ice (Dion and Dick, 2006) and the ocean (Sicre et al., 2008; Cronin et al., 2014). However, the peak of variability at 20-yr timescale is not significant any more in IPSL-CM5A2 (see spectra in Fig. 16, center). This difference can be explained by the different tuning of the two model versions.

Gastineau et al. (2018) used IPSL-CM5A-MR, a variant of IPSL-CM5A with an improved horizontal resolution in the atmosphere (2.5° × 1.25° vs 3.75° × 1.87°), and also found that a warmer mean state led to a reduction of the 20-yr variability in IPSL-CM5A-MR. Such a warmer mean state modifies the subpolar gyre currents and the subsurface stratification in the Eastern Atlantic Ocean where the mixed layer is deepest in IPSL-CM5A models. These changes were suggested to explain a smaller growth of baroclinic instabilities triggering the propagation of westward propagating anomalies (Gastineau et al., 2018). It is therefore likely that the different tuning of IPSL-CM5A2 led to similar changes. The SST associated with the first principal component of the AMOC also shows weaker SST anomalies in the subpolar gyre when compared to IPSL-CM5A (not shown). The AMV patterns are similar in IPSL-CM5A and IPSL-CM5A2, with a slightly more intense subpolar gyre SST (6.8°C vs 6.2°C for the maximum value in the subpolar gyre) in IPSL-CM5A2. Larger positive SST anomalies in IPSL-CM5A2 are also simulated in the Nordic Seas and Irminger sea associated with the AMV. The larger AMV in IPSL-CM5A2 is consistent with the smaller sea-ice cover in the North Atlantic, and a more intense signature of the atmospheric forcing. Yet, the observed AMV pattern computed from the historical period (1870-2008) using a similar methodology does not show such an intense subpolar pole: the SST anomalies in the subpolar region are only twice as intense as the tropical ones (Deser et al., 2009).

The Pacific variability in IPSL-CM5A2 is identical to that found in IPSL-CM5A. The El Niño Southern Oscillation (ENSO) is characterized by the Niño3.4 time series. IPSL-CM5A and IPSL-CM5A2 show a similar power spectrum (Fig. B3) which is comparable to observations (Bellenger et al., 2014). Both versions have the same bias in the seasonal phase locking (Fig. B3), as the monthly standard deviations are larger in May and June in CM5A and CM5A2, while the observed maximum is in January and December. The ENSO impacts are evaluated using spatial composites of the SST over ocean, 2-meter air temperature over land and sea level pressure. The composites are built using the Niño3.4 time series, and calculating the difference of all years where the Niño3.4 exceeds one standard deviation, minus years less than 1 standard deviation. Fig. B4 illustrates that the ENSO impacts are identical in both versions of the model, with a systematic bias in the position of the Pacific equatorial SLP anomalies which extend too much west toward the Pacific warm pool. However, the SLP anomalies display
Figure 16. a. and b. Spatial patterns of the leading EOF of the yearly AMOC computed from 2490 years of the IPSLCM5A2 PREIND (model years 4510 to 6999) and from 1900 model years of the IPSLCM5A PREIND (model years 1800 to 3699) respectively. The patterns are given in Sverdrup for one standard deviation of the corresponding principal component and the amount of variance explained by the pattern is given in the top right of each panel. c. and d. show the variance spectrum of the associated leading principal component respectively. The dashed lines show the 95% and 99% confidence level of the best-fit first-order Markov red noise spectrum respectively. e. and f. AMV pattern, in K, in IPSLCM5A2 PREIND and IPSLCM5A PREIND respectively, from the same model years as above. The Atlantic multidecadal variability (AMV), was calculated from the low-pass filtered (61-month running mean) North Atlantic SST in 0°N-60°N 80°W-0°E. The AMV pattern is given by the regression of the SST onto the AMV index.
a realistic pattern in the Northern and Southern Hemispheres, with the associated Pacific North American and Pacific South
American pattern, as in other CMIP5 models (Weare, 2013). The Pacific decadal oscillation (PDO) was also calculated as the
first EOF of the monthly Pacific SST anomalies north of 20°N (Fig. B5), which illustrate a pattern with positive SST anomalies
over the equatorial Pacific, negative anomalies in the northwest Pacific, and horseshoe pattern of positive SST anomalies in
the Northern Hemisphere. The IPSL-CM5A model generally simulates the characteristics of the observed PDO, as found by
(Fleming and Anchukaitis, 2016) or (Nidheesh et al., 2017), as well as the associated anomalies in the Atlantic and ocean
basins. The IPSL-CM5A2 version simulates a PDO nearly identical to that of IPSL-CM5A, so that the mean state changes
have no impact on the PDO state.

5 IPSL-CM5A2 for deep-time paleoclimate modelling, a case-study

This section depicts both the technical developments and the choices to be made regarding boundary conditions for a typ-
cical deep-time simulation with IPSL-CM5A2. We describe the generation of a new grid for the ocean model for deep time
applications and the generation of boundary conditions, component-wise. To illustrate our point, we run a 3000-year-long
numerical simulation of the Cretaceous as a case study. We present a very brief analysis of the model timeseries to illustrate
the long adjustment required to reach equilibrium in deep-time simulations. As the focus of the paper is the description of
IPSL-CM5A2, we only provide a very basic description of the simulated climate. We refer readers to Laugié et al. (2020) for
a detailed description of a Cretaceous simulation run with IPSL-CM5A2.

5.1 Generating a new grid for NEMO

In addition to the aforementioned computational performances, technical issues related to the oceanic component of the model
also have long prevented the use of any version of the IPSL ESM for deep-time climate modelling. The main lock was that
NEMO runs on a tripolar grid (ORCA2, with two poles in the northern hemisphere and one over Antarctica) that was designed,
amongst other objectives, to avoid any singularity points on the computational domain (i.e. the ocean domain) (Madec, 2015;
Madec and Imbard, 1996). Specifically, north of 20°N the mesh is made of a series of embedded ellipses, the foci of which
are the two north poles of the grid that are by design placed over continents (Fig. 17a). Southward the grid is a Mercator (i.e.
identical longitudinal ant latitudinal grid spacing) grid, defined down to 78°S, except from local refinements applied in the
tropics, so that the meridional resolution reaches 0.5° at the Equator and over the Red, Black, Caspian and Mediterranean Seas
where resolution is about 1°.

Because the grid poles are located over North America, Asia and Antarctica, changing the land-sea mask to account for
continental drift can shift the singularity points into the computation domain. In addition, a number of ocean routines include
hard-coded specifications that apply to some narrow modern straits, such as the Gibraltar strait. In these regions, the resolution
of the grid is much larger than the width of the straits, thereby requiring the outflow of water to be prescribed. Although needed
to better simulate modern climates, these features are irrelevant for deep-time simulations. To overcome these issues, we design
a new PALEORCA grid based on techniques and code designed by G. Madec and A. Sellar. The southern hemisphere (SH)
pole is conserved, which restricts the application of IPSL-CM5A2 to the last 100 million-years (Antarctica has been located over the South Pole since ca. mid-Cretaceous). In the northern hemisphere (NH), new ellipses are computed, changing the location of the two singularity points to eventually obtain a grid that can effectively be used for the last 100 million-years. To this end, and because the grid structure requires both NH poles to be diametrically opposed and limits the range of latitude in which the location of the poles can be chosen, the two poles are placed respectively at 106°E, 60°N and 74°W, 60°N (Fig. 17b). Above-mentioned refinements over the Mediterranean and Eurasian Seas are removed, but the tropical refinement is kept. We also take advantage of the well-defined coding guidelines of NEMO to create a “PALEORCA” configuration within the available oceanic configurations of the model. In this configuration, no hard-coded water exchanges across the globe are prescribed.

5.2 Application: Generating boundary conditions for deep-time climate modelling

The Cretaceous is historically known for its warmer than present climatic state without permanent ice sheets at the poles. Here, we present a simulation of the Turonian stage (circa 90 Ma), during which the apex of the Cretaceous warmth was reached. CO2 estimates reconstructed from proxy-based studies vary from 500 ppm to > 2000 ppm with large uncertainties (Wang et al., 2014). In our simulation, we impose an atmospheric CO2 concentration of 1120 ppm (4x preindustrial values) and we remove polar ice sheets. In addition, we keep the modern orbital configuration as no numerical solution for orbital parameters exists for the Cretaceous (Laskar et al., 2004). The basis of our Turonian configuration is the Cenomanian/Turonian paleogeography of (Sewall et al., 2007).
Figure 18. Topography (a) and bathymetry (b) (in meters) imposed for the Cretaceous simulation.

5.2.1 Oceanic boundary conditions

Bathymetry. The Turonian bathymetry (Fig. 18b) is obtained from the merging of two datasets. First, the regular longitude-latitude Cenomanian/Turonian paleogeography from (Sewall et al., 2007) is masked over continental areas and remapped onto the curvilinear PALEORCA grid. Given that the bathymetry of the Pacific and Indian oceans is mostly flat (without ridges) and
deep (~5000 m) in Sewall et al. (2007), we have decided to merge the more recent bathymetric dataset of Muller et al. (2008) with the land/sea mask and continental margins of Sewall et al. (2007) to create a more realistic deep bathymetry. Handmade corrections are then applied to the bathymetry to ensure that narrow straits (e.g. West Africa) are at least two grid points wide. This avoids the implementation of specific hard-coded schemes in the model to represent water advection across those straits.

**Geothermal heating.** The model modern boundary conditions include a realistic map of geothermal heating, whose impact on the ocean abyssal dynamics is non-negligible (Emile-Geay and Madec, 2009). As the geothermal heat flow $q(t)$ can be related to the ocean lithosphere age $t$, we construct a geothermal heating map for the Turonian based on the reconstruction of the Turonian ocean crustal age of Müller et al. (2008). We derive the geothermal heatflux using the age-heatflux relationship of the Stein and Stein (1992) model.

For $t \leq 55$ Ma:

$$ q(t) = 510t^{1/2} $$

For $t > 55$ Ma:

$$ q(t) = 48 + 96e^{-0.0278t} $$

In regions where crust age is missing in (Muller et al., 2008) dataset, we impose the minimal geothermal heat flux obtained with the above equations. The geothermal heating map created is then remapped on the PALEORCA grid and we limit the maximal heat flux to 400 mW.m$^{-2}$ because the Stein and Stein (1992) parameterization becomes singular for young ages (Emile-Geay and Madec, 2009).

**Tides.** Modern boundary conditions also include forcings of the dissipation associated with internal wave energy for the M2 and K1 tides component - see (de Lavergne et al., 2019). Here, in the absence of an appropriate model to estimate M2 and K1 dissipation values for the Cretaceous, we neglect this forcing and set these fields to zero.

**Initial conditions.** Salinity is initialized as a uniform value of 34.7 per mil equal to the modern mean ocean value (Lunt et al., 2017). The ocean initial temperature follows a slightly altered formulation of the zonally symmetric temperature distribution proposed by Lunt et al. (2017):

if $depth \leq 1000$ m :

$$ T = 10 + \frac{(1000 - depth)25\cos\varphi}{1000} $$

if $depth > 1000$ m :

$$ T = 10 $$

These initialization temperatures are quite high, as the ice-free Cretaceous periods have been shown to have far much warmer oceans than present, associated with high atmospheric pCO2. No sea ice is prescribed at the beginning of the simulation.
5.3 Continental boundary conditions

**Topography** We use the Cenomanian/Turonian topography from Sewall et al. (2007) but modify the Antarctic continent because the PALEORCA mesh is not defined beyond 78°S, which corresponds to the maximal extent of the modern Antarctic coastline (including ice shelves). This requires that the Antarctic coastline reaches at least the same latitude for paleoclimate simulations, which is not the case in the Cenomanian/Turonian continental configuration of Sewall et al. (2007). We therefore impose the maximal Antarctic topography proposed by Wilson et al. (2012) for the Eocene-Oligocene, which has the advantages of meeting the required Antarctic coastline criterion and is a relatively recent estimate of a deep-time Antarctic topography, which remains generally poorly constrained. The final topography/bathymetry is shown on Fig. (18b). LMDZ requires a 10-minute global elevation dataset to compute subgrid-scale parameters (elevation standard deviation, slope, peaks and valleys) that are needed to compute surface roughness length as well as to activate orographic drag parameterizations (Lott and Miller, 1997). As the paleotopography provided by Sewall et al. (2007) is 2.8°x1.4° in resolution, we simply artificially upscaled it to a 10-minute resolution through the use of cdo remapping tools (Schulzweida, 2019).

**River routing** The river routing system (i.e. length of rivers, direction of flows, locations of river outflows) is constructed by first upsampling the low-resolution topography of Sewall et al. (2007) to a 1°x1° resolution and second using the (Wang and Liu, 2006) algorithm within the QGIS software (QGIS Geographic Information System. Open Source Geospatial Foundation Project.). This algorithm ensures that inland depressions are filled and therefore that all continental freshwater is routed to the ocean in the coupled model.

**Vegetation** In the absence of globally consistent vegetation reconstructions for the Turonian, the imposed vegetation follows a very simple latitudinal distribution of plant functional types (4).

5.4 Results of the case-study experiment

5.4.1 Equilibrium

Time series of SST and t2m shows the strong model departure from the initial conditions in the first centuries of the simulation (This departure is also observed in QTOA timeseries, in Appendix C). After 3,000 years, SST and t2m are adjusted at 28.3°C
Figure 19. Adjustment of the Cretaceous simulation over 3,000 years. Latitudinal cross-section of initial (a,c) and final (b,d) states of salinity (top, PSU) and temperature (bottom, °C) for the Cretaceous simulation. e. Timeseries of SST (° C), t2m (° C) and total OHC (J).

and 26°C, respectively (Fig. 19e). OHC of the first 300 meters is stabilized as well. The long-term adjustment of the ocean is, as for our initial pre-industrial experiment, driven by the initial radiative unbalance resulting from the initial ocean temperatures and the greenhouse forcing. The vertical cross sections at initial and final states of the experiment shows the spatial patterns of adjustment of ocean temperature and salinity (Fig. 19abcd). The constant salinity together with the latitudinal profile of temperature imposed as initial conditions also induce a latitudinal gradient in seawater density, and thereby an adjustment of ocean dynamics. Surface salinity and temperature are also adjusting driven by the precipitation minus evaporation balance (PE) and freshwater inputs from river routing (see next section) + radiative balance. Contrary to surface signals, the total OHC shows that the deep ocean is still adjusting after 3,000 years (Fig. 19e), showing the need for very long integrations when one model deeptime paleoclimates.
5.4.2 Mean surface climate

The elevated values of SST and t2m at equilibrium are primarily explained by the high atmospheric CO2 concentration prescribed. In the ocean, warmest temperatures (> 36°C) are found in the equatorial band in the Atlantic and Indian oceans as well as in the western Pacific (Fig. 20ab), whereas the eastern equatorial Pacific is cooler by a few degrees. The Southern Ocean is relatively warm (> 10°C) as is the North Pacific while the Arctic Ocean exhibits the coldest surface temperatures (reaching less than 8°C at the highest latitudes). There is no sea ice formed in the polar oceans. In continental areas, the radiative impacts of the high atmospheric CO2 concentration and the absence of ice sheets also substantially affect the mean annual t2m. Values > 40 °C are found in equatorial Africa, South America and southeastern Asia, whereas below 0 °C mean annual continental t2m are only located in Antarctica (Fig. 20b). The winter season yields well below freezing t2m over large parts of the NH high latitudes, with lowest values in northern Siberia. Similarly, the whole Antarctic continent suffers below freezing t2m, with values below -16°C over large parts of the continental interior. In the summer however, Antarctic t2m are mostly above 20 °C and reach up to more than 28 °C close to the modern Weddell Sea (not shown). Tropical continents are affected by extremely high summer t2m reaching more then 44°C whereas NH high latitudes exhibit the lowest summer values with < 16°C in northern Siberia.

Mean annual PE shows expected patterns across the globe (Fig. 20d), including the double ITCZ as in the preindustrial and historical simulations. Net evaporation areas located at subtropical latitudes under the descending branch of the Hadley circulation and net precipitation areas located in the equatorial band (except in the equatorial Pacific and Atlantic) and in the mid- to high-latitudes. Note the particularly dry tropical continental regions of Africa and South America, which are the results of their more southern position during the Cretaceous (the same is true for South Asia in the Northern Hemisphere).

PE generally explains well the patterns of open ocean sea surface salinities (SSS, Fig. 20c). Highest SSS are thus found in the subtropical areas whereas lower values are found in the mid- to high-latitudes and in the equatorial band. This spatial SSS distribution is altered by the supply of water from continental rivers. The very high amount of precipitation occurring in the equatorial band over India and Northern Africa indeed increase the freshening of nearby coastal waters. Low salinity waters are also found in the Australo-Antarctic sector of the Indian Ocean and in the Arctic Ocean, where large paleorivers can supply significant amounts of freshwater to the ocean. The relative isolation of the Arctic from the global ocean limits possible exchanges of water and therefore may amplify the freshening of this basin. The Pacific sector of the Southern Ocean is more saline than other corresponding high latitudes oceanic regions (North Pacific, Atlantic and Indian parts of the Southern Ocean) with SST of similar or lower values. The winter cooling of the dense surface waters in this region therefore leads to deep water formation as shown by the winter deepening of the mixed-layer depth (Fig. C3). Contrary to the modern, there is no deep water formation in the Northern Hemisphere, in relation with the peculiar Turonien paleogeography, in which the northern Atlantic is only partially open and NH landmasses reach low latitudes. This simulation, and by extant the overall response of IPSL-CM5A2 to a Cretaceous paleogeography without ice sheets and elevated CO2 levels, would need to be evaluated in more details against previous work and proxy data. Still, the basic analysis presented here demonstrates that the mean climate simulated by our model shares many similarities with that simulated by...
other investigations of Cretaceous warmth with coupled models (Niezgodzi et al., 2017; Tabor et al., 2016; Upchurch et al., 2015). It therefore suggests that our goal of adapting the IPSL ESM to deep-time boundary conditions has been reached with success.

Figure 20. Climate mean state of the Cretaceous experiment after 3000 years. a. SST (°C), b. t2m (°C), c. SSS (PSU), d. PE (mm day⁻¹).
This article aimed at providing a detailed description of the building of the IPSL-CM5A2 ESM, developed with the aim of performing multi-millennial earth system simulations. We expect IPSL-CM5A2 to be used mostly for deep-time paleoclimate modelling and, given its reasonable computing performances, attempts of transient simulations for the Quaternary and the future. Starting from IPSL-CM5A, our goal to obtain a faster model with reduced biases has been reached successfully. IPSL-CM5A2 pre-industrial integration shows a reduced cold bias, no drift in the OHC after 1500 years, and reasonable energy leaks and salinity drift. Despite persistent biases in the tropics, the historical runs also show a more vigorous AMOC, together with better-located westerlies. The simulation of global NPP, although still underestimated, has been largely improved. Our case-study for the Cretaceous shows the numerous hurdles, often undocumented, that need to be crossed for setting-up a climate simulation for deep time.

Documenting the strength and weaknesses of IPSL-CM5A2 stimulates questions for the future of paleoclimate modelling. Although numerous interdisciplinary studies can be triggered by efficient climate modelling at the frontier of geology, climatology and biology (see e.g. Haywood et al. (2019) for a synthesis), the question of the stability of biases with radically different boundary conditions remains open. Recently, Krinner and Flanner (2018) opened a path by exploring the stationarity of ESMs large scale biases when forced by four-time atmospheric CO2 concentration. Answering the same question for paleoclimate contexts in which the boundary conditions, such as land-sea mask, topography and bathymetry, insolation and greenhouse gas forcing, are so different than present that they impose very strong forcings on the climate system, remains to be done. Until this challenge is tackled, the use of a computing-efficient ESM to run sensitivity experiments to individual forcing factors, for which the geologic record provides good constraints, appears to be the key.

On the technical point of view, the ongoing race to higher resolution in ESMs asks for consistency between code architecture and supercomputer design. For example, a simple comparison of computing performances of IPSL-CM5A2 and IPSL-CM6-LR, the higher resolution IPSL model designed for CMIP6, shows that the latter requires ~8 times more computing resource than IPSL-CM5A2 to run a simulation of a given length (Fig. C1). Specifically, for a 1000-year experiment, IPSL-CM6 requires more than 1.6 millions of computing hours, simulating 16 SYPD on more than 1000 cores. This model is being used for paleoclimate snapshots experiments, like in the PMIP project, but its use for long, multi-millennial simulations that are typical of deep-time paleoclimate studies is compromised by these performances and requirements. Similar constraints have led colleagues from other climate modelling group to keep updating and using climate models at the resolution they were originally designed more than 15 years ago (Valdes et al., 2017). Given the apparent contradiction between (i) the need for long integrations in deep-time climate modelling, (ii) the enormous resource requirements of modern ESMs, and (iii) the limited number of computing time available on supercomputers, this strategy may well be the most appropriate, in the absence of a technological leap. At IPSL, the next generation of ESMs will likely include an atmospheric model with a new dynamical core (LMD-DYNAMICO) for which scaling will be far better than LMDZ. The very recent version of NEMO (version 4) might also help breaking this computing lock, as the code now benefits from high level parallelization. Still, there is a supplemental challenge for paleoclimate modelling: numerous developments have been made in the recent years in the different components.
of IPSL ESMs, with the inclusion of passive tracers in the ocean (Arsouze et al., 2007), the water isotopes in the atmosphere (Risi et al., 2010), and the explicit resolution of chemical photoreactions in the stratosphere (Jourdain et al., 2008). All of them appear to be particularly relevant for paleoclimate studies (Botsyun et al., 2019; Sepulchre et al., 2014; Szopa et al., 2019), but most drastically reduce the computing efficiency of the model. A trade-off between the number of tracers and/or processes to be included in the model, its spatial resolution, and the length of simulations required for paleoclimate simulations will thus have to be found in the next years to setup relevant experiments designs in the deep-time climate modelling field.

Code availability. LMDZ, XIOS, NEMO and ORCHIDEE are released under the terms of the CeCILL license. OASIS-MCT is released under the terms of the Lesser GNU General Public License (LGPL). IPSL-CM5A2 code is publicly available through svn, with the following command lines: svn co http://forge.ipsl.jussieu.fr/igcmg/svn/modipsl/branches/publications/IPSLCM5A2.1_11192019 modipsl
cd modipsl/util;./model IPSLCM5A2.1

The mod.def file provides information regarding the different revisions used, namely:

- NEMOGCM branch nemo_v3.6_STABLE revision 6665
- XIOS2 branches/xios-2.5 revision 1763
- IOIPS/L src svn tags/v2.2.2
- LMDZ5 branches/IPSLCM5A2.1 rev 3591
- branches/publications/ORCHIDEE_IPSLCM5A2.1.rev5307 rev 6336
- OASIS3-MCT 2.0_branch (rev 4775 IPSL server)

The login/password combination requested at first use to download the ORCHIDEE component is anonymous/anonymous. We recommend to refer to the project website: http://forge.ipsl.jussieu.fr/igcmg_doc/wiki/Doc/Config/IPSLCM5A2 for a proper installation and compilation of the environment.

Data availability. All model outputs analysed in this paper are available as netcdf files on the zenodo repository with the DOI 10.5281/zenodo.3549769. We also published a jupyter Notebook gathering most of the scripts used to create the figures. This notebook is embedded in a binder container that allows anyone to run it and analyse the code. It is published on the zenodo repository with the DOI: 10.5281/zenodo.3549652. Most of the figures were done using NOAA pyferret within jupyter notebooks, thanks to the ferretmagic add-on developed at LSCE by Patrick Brockmann. Ferret is a product of NOAA’s Pacific Marine Environmental Laboratory. Information is available at http://ferret.pmel.noaa.gov/Ferret, distributed under the Open Source Definition. The Jupyter Notebook is an open-source web application. Most colored figures in this paper were made with perceptually uniform, colour-vision-deficiency friendly scientific colormaps, developed and distributed by Fabio Crameri (Crameri, 2018a, b), to prevent visual distortion of the data. We also used the C-ESM-EP (CliMAF Earth System Model Evaluation Platform) developed at LSCE by Jérôme Servonnat https://github.com/jservonnat/C-ESM-EP/ to evaluate the model: https://vesg.ipsl.upmc.fr/thredds/fileServer/IPSLFS/pselpuch/C-ESM-EP/CM5A2vsDATA_pselpuch/C-ESM-EP_CM5A2vsDATA.html. C-ESM-EP is based on CliMAF (Climate Model Assessment Framework) an open-source library distributed with a GPL-compatible license. We acknowledge the NCAR Climate Analysis Section’s Climate Variability Diagnostics
Appendix A: Configuration and performances

A1 Setting up IPSL-CM5A2 configuration

Before making our choice regarding the new model configuration, we explored the computing performances of two configurations of the IPSL coupled model, differing by the spatial resolutions of the atmospheric and oceanic components. The first configuration keeps the spatial resolution of IPSL-CM5A, i.e. 3.75°x1.875° and 39 vertical levels in the atmosphere (hereafter 9696L39), and around 2° in the ocean (with a refinement to 0.5° in the tropics, hereafter ORCA2). The second configuration has higher spatial resolution: the atmosphere is at 1.25°x2.5° and 79 vertical levels (hereafter LMDZ-144143L79) and has a 1° resolution in the ocean (NEMO-ORCA1). This last version is the one used for the CMIP6 project. Here we analyze the computing performances of the separate components to optimally allocate computing resources and determine the best trade-off between resolution and computation time. Scalability information for each component is obtained from LUCIA (Maisonnave and Caubel, 2014) which is directly incorporated in OASIS3-MCT. LUCIA measures at each time step the time spent for (i) component calculations, (ii) variables received/sent from/to other components, and (iii) interpolations. This method has the advantage of measuring the exact scalability of the components during the coupled system production phase. The maximum speed of the coupled system is approximately the speed of the slowest component because the model is run on concurrent mode (i.e. both components run at the same time because they require fields from the previous coupling time step). Fig. A3 shows that despite tests with more than 500 cores, LMDZ and NEMO scalability with configuration 2 remains low, with maximal performances reaching 8 simulated years per day (SYPD) for LMDZ144143L79 and 12 SYPD for NEMO-ORCA1. Scalability is much higher in configuration 1, with NEMO saturating at more than 100 SYPD with 200 cores and LMDZ reaching 56 SYPD with 256 cores (32 MPI processes x 8 OMP threads). These results lead us to use the first configuration as a basis for designing a fast version of the IPSL coupled model.

A2 Performances analysis

A2.1 IPSL-CM5A2 vs IPSL-CM5A: computing performances on supercomputer CURIE

This section assesses how the technical developments depicted in section 2.2 improve the computing performances of the model. From (Balaji et al., 2017) we use a set of metrics relevant for studying computational performances of IPSL-CM5A2. Besides SYPD, four metrics are used here:

- CHSY (core-hour per simulated year): This is the product of the model elapsed time for 1 simulated year times the number of computing cores. It indicates the computing cost of a simulation.
- Parallelization: number of computing cores allocated for the simulation.
Table A1. Computing performances metrics collected on CURIE and JOLIOT-CURIE platforms. Parallelization is defined as the total number of computing cores (Number of MPI (x Number of OMP) per component). One core is always used for XIOS process.

<table>
<thead>
<tr>
<th>Platform</th>
<th>Configuration</th>
<th>SYPD</th>
<th>CHSY</th>
<th>Parallelization (TOTAL/ATM-OCE-CPL)</th>
<th>Coupling</th>
<th>I/O</th>
</tr>
</thead>
<tbody>
<tr>
<td>CURIE</td>
<td>IPSL-CM5A</td>
<td>10</td>
<td>91</td>
<td>38 (32+5+1)</td>
<td>12%</td>
<td>10%</td>
</tr>
<tr>
<td></td>
<td>IPSL-CM5A2 (T)</td>
<td>38</td>
<td>101</td>
<td>160 (32+4 + 31 + 1)</td>
<td>&lt;1%</td>
<td>10%</td>
</tr>
<tr>
<td></td>
<td>IPSL-CM5A2 (S)</td>
<td>56</td>
<td>129</td>
<td>302 (32x8+45+1)</td>
<td>&lt;1%</td>
<td>10%</td>
</tr>
<tr>
<td>JOLIOT-CURIE</td>
<td>IPSL-CM5A2</td>
<td>80</td>
<td>91</td>
<td>302 (32x8+45+1)</td>
<td>&lt;1%</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
<td>81</td>
<td>94</td>
<td>317 (32x8+60+1)</td>
<td>&lt;1%</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
<td>98</td>
<td>107</td>
<td>437 (47x8+60+1)</td>
<td>&lt;1%</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
<td>98</td>
<td>110</td>
<td>452 (47x8+75+1)</td>
<td>&lt;1%</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
<td>103</td>
<td>145</td>
<td>625 (47x12+60+1)</td>
<td>&lt;1%</td>
<td>-</td>
</tr>
<tr>
<td></td>
<td></td>
<td>104</td>
<td>148</td>
<td>640 (47x12+75+1)</td>
<td>&lt;1%</td>
<td>-</td>
</tr>
</tbody>
</table>

– Coupling cost: overhead due to the coupling, computed as the cost of the coupling algorithm itself, i.e. the ratio of CHSY with and without coupling (CHSY – CHSY no coupling) / CHSY. The “no coupling” means atmospheric model only.

– Data output cost: cost of performing I/O, computed as the ratio of CHSY with and without I/O : (CHSY – CHSY no I/O) / CHSY.

Compared to IPSL-CM5A, which could be run on 38 cores only, hybrid parallelization for IPSL-CM5A2 permitted to design two configurations using more than 100 cores: a high throughput version (IPSL-CM5A2-T) on 160 cores, and a “fast” version (IPSL-CM5A2-S) running on 302 cores. IPSL-CM5A2-T configuration allows to reach 38 SYPD (3.8 times faster than IPSL-CM5A) with a 10.8 % increase in computing cost (CHSY, Table A1 and Fig. A1). IPSL-CM5A2-S reaches 56 SYPD, for a CHSY 42 % larger than for IPSL-CM5A (129,000 hours for 1,000 simulated years). Besides, CHSY is greater for IPSL-CM5A2-S, which means there is a better use of a resource allocation than with IPSL-CM5A. The replacement of the sequential OASIS3.3 coupler by the parallel OASIS3-MCT library in IPSL-CM5A2 also allows reducing the cost of the coupling from 12 to 1 % of the total CPU time of a simulation. Regarding inputs/outputs aspects, IPSL-CM5A used the IOIPSL library, which performed sequential writing and imposed a rebuild post-processing step to merge multiple files into one single file. Instead, IPSL-CM5A2 uses the XIOS library that writes out data asynchronously using dedicated I/O servers. Parallel writing allows users to obtain directly one single file without any time-consuming rebuild step. Still, although the writing is performed asynchronously, the I/O cost remains important in IPSL-CM5A2 (10 %) due to operations (temporal operations, variables combinations) that are now performed by the library (on the client side, i.e. model process) rather than previously into the model.
A2.2 Performances of IPSL-CM5A2 on the supercomputer JOLIOT-CURIE

Switching from CURIE to JOLIOT-CURIE supercomputer opened opportunities to increase performances of IPSL-CM5A2 through several aspects: first, replacing Intel Sandy Bridge by more recent Intel Skylake processors was expected to reduce computation time. Second, we decided to test higher parallelization levels, by combining (i) two-latitude band parallelization (instead of three), (ii) an increased number of OMP processes for LMDz, and (iii) more MPI processes for NEMO. Figure A1 depicts the scaling of IPSL-CM5A2 on JOLIOT-CURIE and CURIE. It shows that for the same parallelization set at 302 cores, IPSL-CM5A2 computes about 43% faster on JOLIOT-CURIE than on CURIE (80 SYPD compared to 56 SYPD), with a cost reduced by $\sim30\%$ (91,000 hours for 1000 year). Finding the most valuable configuration requires both to find a tradeoff between maximizing SYPD and minimizing CHSY for each component and to minimize each component waiting time in coupled configuration. To achieve the latter goal, we used LUCIA (Maisonnave and Caubel, 2014) to determine LMDZ and NEMO computing and waiting time. From the 302-cores configuration, we made a step-by-step increase in LMDZ and NEMO resources that shows that both components waiting times are minimized when LMDZ runs on 47 MPI processes and 8 OMP threads together with NEMO using 60 MPI processes (Fig. A3). This configuration uses 437 cores and allows simulating 98 years per day, leading to a cost of 107,000 hours for a thousand-year simulation. Scaling up does not provide significant improvements in SYPD numbers but the cost of the simulation is dramatically enhanced (39% increase when we dedicate more than 600 cores to the model (Fig. A1 and Table A1). Consequently, the 437-core configuration has been chosen as the distributed version of reference.
Appendix B: Supplemental mean-state diagnostics and modes of variability

This section includes supplemental diagnoses for the mean-state of the historical simulation, namely a comparison of 50°S-50°N SST annual mean value with CMIP5 climate models (fig. B1) and winter and summer 850 hPa wind divergence (fig. B2), that helps explain rainfall patterns in the tropics. Detailed diagnoses of modes of variability comparing IPSL-CM5A, IPSL-CM5A2 and data are also presented.
Figure A3. Waiting time of the model components and computation time of the coupler and interpolations, as a function of incremental resources provided to each component. The XX.X.X code on the x-axis indicates successively the number of MPI processes, OMP threads for LMDZ and MPI processes for NEMO. This shows that most efficient configuration is the one with 47 MPI and 8 OMP dedicated to LMDZ and 60 MPI dedicated to NEMO.
Figure A4. IPSL-CM5A2 computing performances compared to IPSL-CM6-LR. Scalability, i.e. SYPD as a function of parallelization, of IPSL-CM5A2 on CURIE and JOLIOT-CURIE supercomputers compared to IPSL-CM6-LR (blue disks) on JOLIOT-CURIE. The surface area of the disks is proportional to the computing cost (CHSY).

Figure B1. Simulated SST of the historical experiment compared to results from CMIP5 models
Figure B2. 850 hPa wind (vectors) and wind divergence ($10^{-6}$ s$^{-1}$) for DJF (a) and JJA (b).
Figure B3. (Left) Power spectra of the Niño3.4 index (SST in 5°N-5°S, 170°W-120°W) in variance-conserving form, in observations (top, HadISST, 1920-2016), and in the PREIND simulations of (center) IPSL-CM5A and (bottom) IPSL-CM5A2. The best-fit first-order Markov red noise spectrum (red curve) and its 95% (blue curve) and 99% (green curve) confidence bounds are shown on each panel. (Right) Monthly standard deviation of the Niño3.4 index for the same observations and simulations.
Figure B4. Composite of all years greater than 1 standard deviation (El Niño) and all years less that -1 standard deviation (La Niña), based on the December Niño3.4 index (SST in 5°N-5°S, 170°W-120°W) smoothed with a 3-point binomial filter; for observations (1979-2016 period), and for the PREIND simulations of (left) IPSL-CM5A and (center) IPSL-CM5A2. The number of El Niño/La Niña events composited is shown in the right subtitle. Temperatures are color shaded and in units of Celsius and are show the SST in the ocean and the 2m temperature over the continent. Sea level pressure is contoured from -16 to 16hPa by 2hPa; negative contours are dashed. Top row is for DJF, almost simultaneous with the December Niño3.4, while bottom row is for JJA preceding December Niño3.4 by about 5 months.
Figure B5. SST patterns for the Pacific Decadal Oscillation (PDO). The PDO is defined using monthly leading principal component of North Pacific (20°N, 110°E:100°W) area-weighted SST* anomalies, where SST* denotes that the global mean SST anomaly has been removed at each timestep. Pattern created by regressing SST anomalies (in °C) onto the normalized PC timeseries.
Appendix C: Cretaceous supplemental analysis: Equilibrium, seasonal temperatures and mixed-layer depth

This section presents simulated surface conditions for the Cretaceous after 3000 years of simulation (last 100 years average).
Figure C1. (a) Time-series of TOA imbalance for the 3000 years of the Cretaceous simulation. (b) TOA imbalance versus near-surface temperature.
Figure C2. Simulated JAS and JFM near-surface temperature after 3000 years of run for the Cretaceous experiment.

Figure C3. Simulated JAS and JFM mixed-layer depth after 3000 years of run for the Cretaceous experiment.
Appendix D: Glossary

ACC: Antarctica circumpolar current
AMOC: Atlantic meridional overturning circulation
AMV: Atlantic multidecadal variability
CERES: Earth’s Radiant Energy System
CHSY: core-hour per simulated year
CMIP: Coupled Model Intercomparison Project
CRF: cloud radiative forcing
DGVM: dynamic global vegetation model
DJF average: December-January-February average
EBAF: Energy Balanced and Filled Fluxes
EMIC: Earth System Models of Intermediate Complexity
EOF: empirical orthogonal function
ESM: Earth System Model
GCM: general circulation model
GPCP: global precipitation climatology project
HPC: high-performance computing
I/O: input/output
IPSL: Institut Pierre Simon Laplace
ITCZ: Intertropical Convergence Zone
JJA average: June-July-August average
LMD: Laboratoire de Météorologie Dynamique
LW: Longwave
Ma: million-years
MPI: Message Passing Interface
NH: Northern hemisphere
NPP: net primary production
OHC: ocean heat content
OMP: OpenMP=Open Multi-Processing
PDO: Pacific decadal oscillation
PE: precipitation minus evaporation balance
PFT: plant functional type
Q_TOA: Top of Atmosphere net radiation
SH: Southern hemisphere
slp : sea-level pressure
SSH : sea-surface height
SST: sea-surface temperature
SW: Shortwave
SYPD : simulated years per day
t2m : surface air (2-meter) temperature
TSI: total solar irradiance

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Disclaimer. The simulations analyzed in this paper required approximately 500,000 CPU hours, emitting approximately 1.7 teqCO2.

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