Interactive comment on “Synthesizing long-term sea level rise projections – the MAGICC sea level model” by Alexander Nauels et al.

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Received and published: 24 March 2017

Referee 2 General Comment (RC2.00): The ms tackles an important and nontrivial point and makes several new contributions. The authors are to be applauded for the efforts and providing the new insights. Alas, the current ms suffers from several (in my view severe, but pretty easy to fix) shortcomings that need to be mitigated to enable a careful review and to hopefully (and eventually) rise to the level of quality one expects from papers in GMD. These specific points are discussed below. The ms requires in my view a careful and responsive revision and re-review. I would be happy to look at this ms again.

Author Response (AR2.00): We thank the referee for this thorough review leading to numerous revisions and additions, which, in our eyes, have substantially improved the manuscript. We regret not having spent more time on the topic of code availability before the initial submission. We hope that our response and the provision of the MAGICC source code address the issues raised by the referee. Please note, that all revised and additional Figures as well as one additional Appendix Table are provided at the end of this document for your convenience.

Referee 2 Comment 1(RC2.01): The ms does, in my view, not follow the data and code policy of GMD. As a reminder, here is the link describing the policy: GMD code and data policy. These rules state: “Preferably, this section should contain the instructions for obtaining the model code and/or data, either from the supplement or from an archive with a digital object identifier (DOI). Suitable repositories can be found at the Registry of Research Data Repositories, e.g. ZENODO for model code. After the paper is accepted, a link to the GMD paper should be added to the metadata of the archive.” What is available, as far as I can tell, is code for the sea-level rise model / module. However, it will be extremely tough to reproduce the results in the paper, due to several issues. For example, the interface to call this module is not well documented, nor are the data used to drive and calibrate the model provided. The web-page the ms links to states: “The Fortran code of the model as well as required additional MAGICC input files can be downloaded as a zip-file at the top right of this page. Please note that the MAGICC sea level model is not a stand-alone tool. It is fully integrated into the simple carbon cycle climate model MAGICC and cannot be run separately. The MAGICC model is described in Meinshausen et al. (2011). For a compiled MAGICC version including the sea level model, please contact alexander.nauelsclimate-energy-college.org. MAGICC is licensed under a Creative Commons Attribution-NonCommercial-ShareAlike3.0 Unported License”. Does this mean that the source code for the model needed to run the sea-level model is not accessible? Why is only a compiled version distributed? Given this situation, I would assign a very low score for reproducibility (see the discussion by Haas in the recent Risk Analysis volume (Risk Analysis, Vol. 36, No. 10, 2016, DOI: 10.1111/risa.12730). Many other studies have abided by the nice standards outlined...
in GMD and a publication in GMD should clearly abide by these standards. This is a
decision for the editor. I cannot carefully review the ms without the ability to see and
run all the code and cannot support publication as a peer reviewed study in GMD with
the current state of data availability.

AR2.01: We acknowledge the concerns expressed by the referee and regret that we
have not clarified the availability of the MAGICC source code earlier. We hope that
the following update resolves this issue. In line with the GMD code and data policies,
the sea level model presented here is available as open-source code on a transparent
and version-controlled repository. Once the manuscript is accepted in its final version,
we will also deposit the sea level code on ZENODO with a DOI. We agree that it is
challenging to reproduce the sea level model output without the parent model code.
The MAGICC model itself is not open-source, which is mainly due to the long model
development history. The model was first introduced in the 1980s, a time when the
open-source concept wasn’t well established in the science community. Subsequent
MAGICC developments have all been carried out without the scope and resources to
fully migrate the model to an open-source platform. Unfortunately, the authors of this
manuscript are not in a position to go open source with the code in its current version.
The model source-code is available for interested researchers under a non-commercial
license agreement on request, which is now communicated correctly in the repository
documentation. A MAGICC model package with the source-code including the sea
level model has been provided to the editor for distribution to the referees to ensure
transparency and reproducibility of the presented work. The most recent MAGICC
model version 6 has been extensively documented in Meinshausen et al. (2011a) and
Meinshausen et al. (2011b). Data for the atmospheric and ocean variables used for
the model calibration can be downloaded from the CMIP5 archive. The interpolated
calibration data for the MAGICC ocean model are available on request. This is now
clarified in the code and data section (p.17 l.20). The other sea level components are
calibrated with additional published datasets that were received by courtesy of the
respective author teams. With the sea level model code openly provided and the full

MAGICC model available to the referees as part of this review and for interested peers
upon request, we hope that we have addressed the referee’s concerns regarding the
data and code policy of GMD.

RC2.02: The ms emulates model runs that are mostly samples of best guesses on
choices such as model parameters and structures. This sample is likely missing im-
portant uncertainties. Communicating the range of the projections and the discussing
“probabilistic uncertainty analyses” (abstract) can lead to unfortunate misunderstand-
ings by the users. This needs to be discussed and the communication made more clear
to help the users of this tool and the results to better understand the caveats. There is
a nice start towards this goal on the bottom of page 9, but this needs to be made much
more clear to the reader (e.g., in the abstract). This potential misunderstanding is the
even strengthened in the design of the projections, that uses a “probabilistic MAGICC
design” (p. 14). For example, Figure A1 discusses a 66% range. One question the ms
should address is how the ranges they present compare to expert assessments (that
are cited in the ms).

AR2.02: The referee is correct in noting that the presented sea level model does
not account for deep uncertainties, i.e., uncertainties due to processes that are
not yet known or not captured by process-based models. It is not probabilistic
in that sense. Rather, we present an approach to reflect key uncertainties for all
major SLR drivers based on the process-based projections presented in the Fifth
Assessment Report of the IPCC. Our model is probabilistic in the sense that we
combine parameter uncertainties in a way that results in each calibrated parameter set
reasonably reproducing the specific reference data. By applying the well-established
probabilistic Metropolis-Hastings Markov chain Monte Carlo projection method from
Meinshausen et al. (2009), which also underlies, e.g., Meinshausen et al. (2011c),
Rogelj, Meinshausen and Knutti (2012), Perrette et al. (2013), Levermann et al.
(2014), we can additionally map key climate and carbon-cycle related uncertainties;
above all, the range of equilibrium climate sensitivities presented in IPCC’s AR5. The MAGICC sea level emulator therefore provides projections that, besides scenario related uncertainties, reflect model uncertainties included in the reference data (i.e. from the CMIP5 ensemble), as well as uncertainties from carbon-cycle responses (Meinshausen et al. 2011a, Friedlingstein et al. 2014) and latest climate sensitivity estimates (Flato et al. 2013). In order to avoid confusion regarding the scope of uncertainties that can be assessed with the presented approach, we have modified the abstract, with the relevant sentence on p.1 l.18 now reading: “The MAGICC sea level model provides a flexible and efficient platform for the analysis of major scenario, model, and climate uncertainties underlying long-term SLR projections”. In addition, we have elaborated on the probabilistic method and the comparison of our Antarctic solid ice discharge projections (90% model ranges now included in the Appendix figures) to the ones presented in DeConto and Pollard (2016). We have revised p.14 l.17 to read: “In order to also provide a sufficiently large number of model runs for 2300, we use 600 historically-constrained parameter sets that have been derived using a probabilistic Metropolis-Hastings Markov chain Monte Carlo method (Meinshausen et al. 2009). This approach has been extended to also reflect carbon-cycle uncertainties (Friedlingstein et al. 2014) and the climate sensitivity range of the latest IPCC assessment (Flato et al. 2013, Rogelj et al. 2012, Rogelj et al. 2014)”; the section on p.17 l.1 has been updated: “The Antarctic ice sheet response, for example, could be subject to more rapid, non-linear dynamics that are not captured by the emulated process-based projections. Only recently, DeConto and Pollard (2016) have revised potential future Antarctic contributions to global sea level based on indicators from paleoclimatic archives. For RCP8.5, they project 2100 contributions of around 1m from Antarctica alone, with 2300 contributions reaching up to around 10 m. The MAGICC sea level model projections for the Antarctic SID contribution are based on Levermann et al. (2014) and only yield up to around 0.35 m in 2100 and 2.68 m in 2300 for the upper bound of the 90% range. As the more recent research suggests, these estimates may be too low, indicating that the Antarctic contribution to future SLR is subject to additional uncertainties. This illustrates the need to handle long-term SLR projections with care and to note the corresponding methodological caveats; in particular, those surrounding the representation of Antarctic ice-sheet changes.”

RC2.03: The ms makes several broad and somewhat ambiguous claims that need to be updated. For example, the ms states: “This MAGICC sea level model has been designed to emulate the behavior of the latest available process-based sea level projections” (p2). This does seem to be in contrast with the findings of DeConto and Pollard (2016), which the ms cites but in my reading not considers. This also needs to flow into the discussion of ice sheet dynamics on page 13 as well as the summary in the discussion (p. 15).

AR2.03: We agree with the referee that our statement regarding the selected reference datasets was deceptive. We have revised the corresponding statement on p.2 l.13 to read: “This MAGICC sea level model has been designed to emulate the behavior of process-based sea level projections presented in the IPCC’s Fifth Assessment Report (Church et al. 2013a), with thorough calibrations for each major sea level component.” The discussion now includes a direct comparison of the MAGICC sea level model results with the findings from DeConto and Pollard (2016). After searching for other instances of using the potentially ambiguous adjective “latest”, we have also decided to change the land water related statement in the abstract on p.1 l.15 to read: “The land water storage component replicates recent hydrological modeling results.”

RC2.04: How are “semi-empirical models” (p2) specifically defined? Has there been no use before 2007 (as the ms seems to imply)?

AR2.04: We now provide more specific details on the design of semi-empirical models and clarify that the semi-empirical concept has been applied well before 2007. The
section on p.2 l.30 now reads: “In the 1980s, first Semi-Empirical Models (SEMs), which estimate global sea level changes based on the evolution of global mean temperature, were introduced together with early attempts to model thermal expansion based on simplified ocean processes (Gornitz et al. 1982). Generally, SEMs establish statistical relationships between observed/reconstructed global mean temperature or radiative forcing changes and observed/reconstructed global mean sea level changes. Assuming that such relationships stay the same for the future, they are used to estimate future SLR from projected global temperature/forcing changes (Jevrejeva et al. 2010, Kopp et al. 2016, Rahmstorf 2007, Vermeer and Rahmstorf 2009). As such, these SEMs do not calculate sea level by resolving the underlying physical processes.”

RC2.05: Emulators and semi-empirical models are not necessarily alternatives, as the ms seems to imply (p.2). Overall, the ms does not do justice to the very large body of literature on the design, use, and potential usefulness of emulators in the field of climate change research. This comes back to haunt the ms in the discussion (e.g., page 16). Please add to the introduction a more careful review on this issue and revise the statements on this issue in the discussion. How is “robust” (p.2) defined?

AR2.05: In order to define where the presented MAGICC sea level sits methodologically, we have referenced two different methods to efficiently estimate future sea level rise. This distinction is not meant to suggest that semi-empirical models and sea level emulators are alternatives. Both methods have their individual strengths and weaknesses and are often closely related with regards to underlying parameterizations. We have revised the manuscript to include more information on expert elicitation as well as sea level modeling methods that use paleoclimatic information to provide future projections. However, we think that providing a comprehensive discussion on “emulators in the field of climate change research” would unnecessarily lengthen the paper and distract readers interested in the sea level model description. We hope that the referee understands our reluctance to extend the introduction beyond the revised content. On p.3 l.13, we used “robust” to indicate that the calibrations for all parameterizations converge to reach their global optima (see AR2.10 and AR2.11). We acknowledge that this term is misleading and refined the corresponding statement. The following changes have been made to the text: p.3 l.5: “Sea level rise projections have also been provided based on expert elicitation (Horton et al. 2014). Furthermore, sea level expert judgments have been combined with statistical models synthesizing sea level projections for individual components (Kopp et al. 2014). Other studies have used an extended suite of methods, analysing paleoclimatic archives, modeling parts of the SLR response with a reduced complexity model, and deriving future projections for land ice contributions based semi-empirical considerations (Clark et al. 2016). The growing efforts in the sea level modeling community to provide fully transparent and freely available model code are reflected by the recent introduction of a transparent, simple model framework to estimate regional sea levels (Wong et al. 2017). Previous MAGICC versions provided sea level rise estimates based on parameterizations for selected components (Wigley 1995, Wigley and Raper 2005, Wigley and Raper 1987, Wigley and Raper 1992). Here, we adopt the approach of deriving a total sea level response by emulating existing process-based projections for individual sea level components (Perrette et al. 2013, Schleussner et al. 2016). Future sea level dynamics are synthesized by calibrating parameterizations to the selected complex model projections for all major sea level contributions. Progress in the understanding of individual sea level processes and the availability of revised future sea level contributions require sea level emulators to be updated regularly. With this study, we are able to complement the existing sea level projection emulators with a platform based on a comprehensive set of individual sea level components that allows for projections consistent with IPCC AR5 estimates.”, p.3 l.13: “It mimics process-based sea level responses for the seven main sea level components with thoroughly calibrated parameterizations that extend global sea level projections to 2300.”, p.16 l.25: “Sea level emulators complement the comprehensive but computationally expensive, process-based sea level models due to their flexible and
efficient design. They can be quickly adapted to, e.g., incorporate previously unknown uncertainties from newly quantified ice sheet processes (Clark et al. 2016, DeConto and Pollard 2016). Being directly coupled to MAGICC, our sea level model can also account for additional climate system response uncertainties and provide consistent projections for a wide range of climate change scenarios beyond the standard IPCC pathways. The latter aspects describe key strengths of the MAGICC sea level model and make it a useful tool to assess SLR for scenarios that are not covered by larger, more comprehensive models. The emulated MAGICC sea level projections reflect, independently, the reference responses for each individual sea level component, assuming that the implemented parameterizations fully capture the process-based simulations. Underlying model uncertainties differ substantially for the individual sea level components (Church et al. 2013b).

RC2.06: “In this study, we provide a first series of updates for MAGICC version 7 which will be consistent with the ensemble output of CMIP5” (p. 3). Is this a statement of fact or about a possible future?

AR2.06: This is a statement of fact. Currently, the MAGICC carbon cycle is being updated. We expect the fully CMIP5-consistent MAGICC version 7 to be published by the end of this year.

RC2.07: “The tuned model captures key features of the individual ocean heat uptake and vertical redistribution behavior of every CMIP5 model” (p. 4). Which ones?

AR2.07: We use ocean-layer specific potential ocean temperature change as proxy for ocean heat uptake behaviour. As can be seen from Figures 2 and 3, the calibrated MAGICC ocean provides potential ocean temperature warming profiles and thermal expansion projections that are in line with the CMIP5 reference data. We have revised the statement on p.4 l.18: “The tuned model captures ocean-layer specific

RC2.08: Please add a table of all model parameters, definitions, units, and values. Please review the formatting of all equations (e.g., equation 3) and symbol uses in the text (e.g., p6 l13).

AR2.08: We have now included an overview Table A.1 in the Appendix, listing key calibration variables and the calibration parameters for every sea level component. The calibration values have all been provided before in Tables 1 to 5. Thanks for notifying us about the equation formatting issues. All equations have been checked, equation 2, 3, and 6 have been updated (see also AR2.09). We have included an additional sentence on p.11 l.32 to refer to the new Table: “For an overview of all relevant variables and calibration parameters please see Table A.1.”

RC2.09: Why select only 9 parameters for the calibration? How are the chosen? What would happen if you use all parameters? (p. 11)

AR2.09: For the MAGICC ocean model calibration, we have selected all MAGICC parameters that directly determine the ocean-layer specific potential ocean temperature and corresponding thermal expansion responses. These 9 parameters drive the band-routine of the hemispheric upwelling-diffusion ocean model and are now also listed in the additional Appendix Table A.1 for clarity (see AR2.08). The MAGICC sea level model calibration has been carried out per sea level contribution, only optimising the free parameters of the respective component, while prescribing the annual mean surface temperatures. As such, it was easier for us to ensure consistency with the reference data, separate the effects of individual parameter variations, and analyse the calibration quality. While carrying out additional checks of the calibration routines after the submission of the manuscript, we unfortunately discovered a bug in the...
MAGICC routine to prescribe annual mean surface air temperature as well as a problem with the Greenland SID functional form. Also, one specific CMIP5 model input had to be removed due to quality issues with the reference data (BCC-CSM1.1M). All issues have been resolved. The revised prescribed temperature routine required the re-calibration of all sea level model components, with all figures now showing the corrected sea level model results. The updated optimal parameter sets are provided in the corresponding Tables. For the Greenland SID parameterisation, the calibrated constant was removed, because it prohibited reproducing lower bound projections (see revised Equation 6). In addition, the routine was adapted to allow for hindcasts without depleting the maximum outlet glacier volume, which was determined based on the year 2000. The code repository has been updated to account for these changes. We have revised the following sections to clarify the ocean model parameter selection and to account for the necessary code revisions: p.11 l.11 “Here, we select all MAGICC parameters that directly influence the ocean-layer specific potential ocean temperature and corresponding thermal expansion responses. These 9 parameters drive the band-routine of the hemispheric upwelling-diffusion ocean model”, p.7 l.25 “Applying the scaling suggested by Church et al. (2013a), our total Greenland SID maximum ice discharge volumes amount to around 180 mm and 268 mm SLE for the minimum and maximum cases presented in Nick et al. (2013)”, p.15 l.11 “The overall historically-constrained, probabilistic MAGICC global mean tas response for the 21st century is stronger than in the CMIP5 reference data for RCP4.5, RCP6.0 and RCP8.5 scenarios. This slightly steeper 21st century global mean tas slope is also reflected in the corresponding probabilistic MAGICC 2100 SLR estimates, given the strong air temperature dependence of the sea level model (see panels (a) to (d) of Figure 4).”

RC2.10: Why do you choose the 5000 random parameters (I assume as a starting point)? (p. 11). Is this because there are concerns about / evidence for multiple maxima?

AR2.10: Yes, there were concerns about local optima or too flat gradients for the optimization routine to converge properly. Initial calibration testing with 100 initial random runs pointed to local convergence of the MAGICC ocean parameter optimization routine. Extensive testing with 500, 1000, 5000, 7500, and 10000 initial random runs was conducted to establish the necessary number of pre-optimization random parameter sets to appropriately map the parameter space. While all tests with 1000 parameters showed global convergence, we opted for 5000 random runs to ensure a sufficiently large set of samples for deriving the initial parameter set of the optimization routine. The text on p.11 l.22 has been revised: “5000 random parameter sets are drawn prior to each model optimization procedure. The number of initial random runs has been determined through iterative testing to ensure convergence to a global optimum. The resulting best fit is subsequently used for the initialization of the automated Nelder-Mead simplex optimization routine (Lagarias et al. 1998, Nelder and Mead 1965) with a termination tolerance of $10^{-8}$ and a maximum iteration number of 10,000.”

RC2.11: You should (i) cite and discuss the ground-breaking paper: Hargreaves, J. C., and J. D. Annan (2002), Assimilation of paleo-data in a simple Earth system model, Climate Dynamics, 19(5-6), 371-381 and (ii) provide evidence for the assessment that the chosen method has a decent chance to be close to the global maximum. How are the weights for the RSS chosen (p.11, I also assume Table 1)? Is this choice consistent with the properties of the residuals?

AR2.11: Thank you for pointing us to this relevant paper which is now cited in the revised draft. The Nelder-Mead Simplex optimisation routine as described in Lagarias et al. (1998) has been successfully applied for previous MAGICC calibration procedures (Meinshausen et al. 2011a). We are using 5000 iterations of randomly drawn calibration parameter sets to ensure that the subsequent optimisation routine shows global convergence in providing the optimal parameter fits. It is only for the
ocean calibration that weights different from 1 have been applied. The different units of the ocean target variables (K for potential ocean temperatures per layer, mm for thermal expansion) call for a sensible choice of variable specific weights to not introduce an optimisation error bias for the RSS. To ensure that the calibration routine prioritises the ocean calibration while still globally converging for the thermal expansion scaling coefficient, we have iteratively derived the thermal expansion weight (0.001) and the ocean layer temperature anomaly weight (10). The following sections of the calibration section have been revised: p.11 l.11: “Previous studies have shown that highly parameterised simple models do successfully show global convergence when calibrating a large number of free parameters (Hargreaves and Annan 2002, Meinshausen et al. 2011a),” p.11 l.26: “The zostoga optimization component is given four orders of magnitude less relative weight than the thetao component in order to prioritize the accurate layer-by-layer emulation of the respective CMIP5 model thetao time series.”

RC2.12: The claim that the presented results are “superior” (p. 16) is unclear and not backed up by evidence. Please define specific metrics (maybe hindcast cross validation error, etc.) and provide the evidence for this.

AR2.12: We admit that “superior” should not be used in this context. However, we still think that the MAGICC sea level model has advantages over other simple sea level models that either provide a global sea level response only, e.g. semi-empirical approaches like Rahmstorf (2007), use less up-to-date reference data (Perrette et al. 2013, Schleussner et al. 2016), or do not reproduce historical data as well (Mengel et al. 2016). We have revised the corresponding statement on p.16 l.3 to now read: “The close reproduction of selected reference data (Figure 3, Figure A.3), the consistent translation of climate forcing into a SLR response within the MAGICC model, and the comprehensive representation of relevant processes (e.g. the thermal expansion contribution produced by the CMIP5-consistent MAGICC ocean model and

RC2.13: Please review the format and missing information in the citations.

AR2.13: Done. We have updated references which were either incomplete or not in the right format.

RC2.14: Please show the residuals for the calibration (Figure 2). Is there a discrepancy in the lower layers? If so, how is this handled in the calibration?

AR2.14: The calibration residuals for Figure 2 are now provided as potential ocean temperature anomalies for every MAGICC ocean layer in the Appendix, Figure A.1. In addition we have decided to highlight specific model outliers in Figure 2 if they are not covered by the 90% CMIP5 reference and resulting 90% MAGICC model ranges. These figures now more clearly show what has already been stated on p.12 l.20, i.e. the tendency of the MAGICC bottom layers to warm more than the CMIP5 reference data. Calibration results for 2 of the 36 CMIP5 models show a major bottom layer warming bias. The GISS-E2-R reference data shows strong mid-layer warming combined with actual bottom layer cooling, the HadGEM2-CC data shows cooling in the upper 500 m over the historical period. In both cases, the MAGICC hemispheric upwelling-diffusion ocean model cannot fully capture these characteristics and, for the HadGEM2-CC model, overcompensates the surface cooling with strong bottom layer warming. The calibration routine itself gives all ocean layers the same weight and terminates the optimisation when the reduction in the RSS stays with the 10-8 termination tolerance. Both model calibration results represent the optimal parameter fits for the given reference data. We have revised the text to more clearly state these aspects. p.12 l.17 now reads: “The Figure also provides information on individual model outliers for reference data and calibration results. Corresponding potential ocean temperature

the inclusion of the land water storage sea level component) make the MAGICC sea level model a powerful addition to the existing sea level emulators.”

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residuals are shown in Figure A.1”, p.12 l.18: “The updated MAGICC ocean deviates from the CMIP5 data in a few cases. Generally, there appears to be less warming in the mid-ocean between around 1500 m and 2500 m than in the CMIP5 reference data. Also, there is a tendency for the MAGICC bottom layers to warm more than the CMIP5 reference data. However, it is only for 2 of the 36 CMIP5 models used that calibration results show a major bottom layer warming bias. The GISS-E2-R reference data show strong mid-layer warming combined with actual bottom layer cooling, while the HadGEM2-CC data show cooling in the upper 500 m over the historical period (Figure A.2). In both cases, the MAGICC hemispheric upwelling-diffusion ocean model cannot fully capture these characteristics. For the HadGEM2-CC emulation, MAGICC overcompensates the surface cooling with strong bottom layer warming.” Please see also AR2.09, AR2.10, and AR2.11 for the calibration related part of the referee comment.

RC2.15: The fonts and line sizes in Figure 3 are too small to read. Please provide readable font sizes and line / symbol separations (as SOM if need be). Why do you choose only the 90% range for Antarctic solid ice discharge? What explains the change in range in several of the panels (e.g., a and c)? Please show the hindcasts and the residuals for the results shown in Figure 4.

AR2.15: Thanks for pointing this out. We have increased font size, line size, and contrast where possible to improve the readability of the Figure. Hopefully, the editor will agree to provide it as a full page graphic. We have purposefully designed the Figure in the current format to provide all calibration results in one place and to allow for immediate visual checks of the calibration quality. We would like to keep this format. The ranges in each panel have been adapted to fit the projection magnitude and reference data time frame for each sea level component. This has now been stated in the caption. We only show the 90% range of the Antarctic SID component because we were not able to receive more detailed comparison data from Levermann et al. (2014). The hindcast for total SLR is now provided in Figure A.3 with three selected comparison data sets from Mengel et al. (2016), Hay et al. (2015), and Church and White (2011). The author teams of the observational datasets have been included in the acknowledgements. Hindcast for the individual SLR components have been included in Figures A.4 to A.7. For the Greenland SMB, SID and the Antarctic SMB components, the parameterisations are run freely for the historical period as the process-based model projections used as reference data do not cover this period. The historical sea level response, in particular for the GIS SID component, could be further improved in the future by also including reanalyses or observational calibration data as part of a sea level model update. This, however, is beyond the scope of the present work. We updated the text and now refer to the newly provided hindcast on p.15 l.5: “In Figure A.3, we provide MAGICC SLR hindcast results and three comparison datasets for the period 1900 to 2000. The MAGICC sea level model shows good agreement with the observational datasets based on Church and White (2011) and Hay et al. (2015). The global 1900-2300 SLR responses are provided for all RCPs and each sea level component in the Appendix Figures A.4 to A.7.” Additional text has been added to the discussion section on p.16 l.3: Our sea level model transparently emulates and combines long-term sea level projections from process-based models. It also reproduces well observed past total sea level change (see Figure A.3). Therefore, we are confident that this model contributes to advancing efficient long-term sea level projections.”

References:


Please also note the supplement to this comment:
Fig. 1. Figure 2. Potential ocean temperature depth profiles for MAGICC and reference CMIP5 warming under RCP2.6, RCP4.5, RCP6.0 and RCP8.5 scenarios, 2081-2100 anomalies with respect to 1986-2005. [...]

RCP26

RCP45

RCP60

RCP85

(a) (b) (c) (d)
Fig. 3. Figure A.1. Potential ocean temperature residuals for calibrated MAGICC and reference CMIP5 ocean warming under RCP2.6, RCP4.5, RCP6.0 and RCP8.5 scenarios. [...]
Fig. 4. Figure A.2. Annual RCP4.5 ocean warming anomalies for the CMIP5 models GISS-E2-R (1850-2300), HadGEM2-CC (1850-2100), and CCSM (1850-2300), relative to 1850 and globally averaged. [...]

Fig. 5. Figure A.3. Historical modelled and observed SLR from 1900 to 2000, relative to the 1986-2005 mean. Median and 90% uncertainty ranges are shown for the MAGICC hindcast (orange) [...]