Geosci. Model Dev. Discuss., https://doi.org/10.5194/gmd-2018-266-AC1, 2019 © Author(s) 2019. This work is distributed under the Creative Commons Attribution 4.0 License.



Interactive comment on "Global emissions pathways under different socioeconomic scenarios for use in CMIP6: a dataset of harmonized emissions trajectories through the end of the century" by Matthew J. Gidden et al.

Matthew J. Gidden et al.

matthew.gidden@gmail.com

Received and published: 5 March 2019

C1

1 Reviewer 1

1.1 General Comments

Comment

Additionally, the reference list includes mostly papers authored by the authors themselves. A broader literature review is recommended.

Response

We thank the reviewer for their comment regarding the cited literature. The authors of this manuscript do indeed span a number of research teams comprising much of the scenario literature, thus there are inevitably self-references. However, we do agree with the reviewer that our literature review could be more expansive. Therefore, we have increased, where applicable, references to the available literature not including authors on this paper. Notably, we expand our assessment in the user community of emissions scenarios and also with respect to uncertainties in species contribution to radiative forcing.

1.2 Specific Comments

Comment

2.1.3 Region-to-Country Downscaling

Some questions about region-to-country downscaling emerge, particularly about assumptions made. For instance, the use of a linear downscaling method "means that

the fraction of regional emissions in each country stays constant over time" (page 9 line 30). This seems an oversimplification for sectors that may represent a large share of emissions in many countries today (mostly developing countries). However, this may not be true in the future, especially as countries develop and energy use increases. Thus, holding the share of LUC emissions constant over time overestimates their contribution to total emissions, and downplays the potential contribution of energy use in these countries. More importantly, it downplays any mitigation efforts potentially implemented by such countries. Is it reasonable to assume that in SSP1 the shares of Agricultural Waste Burning emissions will remain constant? Why?

This assumption is made without any justification or analysis of its validity and potential impacts on results. The reader is referred to the "downscaling wiki", an online documentation site, but very little explanation is to be found there as well.

Response

We apologize for the confusion. Emissions amounts do change over time, it is just the sub-regional distribution amongst countries that is constant for open-burning emissions. We will update the text to clarify that this as:

means that the fraction of regional emissions in each country stays constant over time. Therefore, the total amount of open-burning emissions allocated to each country will vary over time as economies evolve into the future, following regional trends from the native IAM. However there is no sub-regional change in the within-model-region spatial distribution of landuse related emissions over time. This is in contrast to other anthropogenic emissions, where the IPAT method is used to dynamically downscale to the country level as discussed above.

C3

Comment

Additionally, this linear method may well have different impacts on different GHG species. Aerosols in particular, especially as these are the only species for which the downscaling results are presented in Section 3.4. Surely, agricultural waste burning has an impact on aerosol formation so that the choice of downscaling method in the first step (region to country) is fundamentally important.

Wouldn't it be more appropriate to also use some form of convergence as is done for other sectors? It seems to me this assumption of constant shares for LUC emissions should be fully justified or, preferably, be the subject of sensitivity analysis of some form. This should be further explored, explained and justified. In my view, it should have its own section in the supplementary information. Understandably, there is high uncertainty with these categories of emissions, but that is only more reason to explicitly address it. Also, it would be useful to have a comment by the authors about how this assumption might impact results of the CMIP6 activities using these scenarios as input.

Response

As noted above, open burning emissions are not constant over time. These evolve according to the parent IAM's modeling. The simple IPAT downscaling method was not used because this could lead to illogical results, such as ag waste burning emissions in one country becoming unrealistically large relative to agricultural land area. Certainty improved downscaling methods specific for each type of open-burning emissions are possible. Given that the code has been made available as open source, anyone can implement alternative downscaling methods and test their impact on, for example, aerosol loadings.

Comment

Appendix C: Emissions Gridding

Page 30, line 5: "For each aggregate sector the spatial pattern of emissions within a country, therefore, does not change over time in the future scenarios, although the spatial pattern of total emissions will change due to changes in the sectoral distribution of emissions."

This sentence is important and should be better formulated. I am having difficulty understanding what stays constant and what changes over time. Maybe the crux of the problem is what is meant by the word "total emissions". Is meant as "global emissions", as contrast with "emissions within a country"?

In my opinion, this section should be expanded to include the points I raised above in my previous comment on Section 2.3.1.

Response

This section will be expanded, as follows, to better explain this concept. Note, a companion paper is being submitted that will more fully explain the gridding methodology, results, and implications.

Emissions data were mapped to a spatial grid generally following the methodologies described in Hoesly et al. (2018). A brief description is given here, and a fuller discussion of the gridding process will be provided in Feng (2019). For most anthropogenic sectors, emissions at the level of country and aggregate sector are mapped to a 0.5° spatial grid by scaling the 2010 base-year country-level spatial pattern. Open-burning emissions from forest fires, grassland burning, and agricultural waste burning on fields

C5

are mapped to a spatial grid in the same manner, except that the spatial pattern is taken to be the average from the last 10-years of the historical dataset (e.g., 2005-2014). For each aggregate gridding sector the spatial pattern of emissions within a country does not change over time in the future scenarios. This means that, for example, the ratio of energy-sector NOx emissions from Shannxi and Beijing provinces in China is constant over time, even though total NOx emissions from China vary over time. Because sectors are mapped to the grid separately, however, total anthropogenic emissions (e.g. sum from all sectors) from any two regions within a country will, in general, not have the same time evolution.

Comment

Technical Corrections Page 2, line 32: "where" should be "were" Page 21 line 13: there is no Section 2.4

Response

We have corrected both typographical errors identified by the reviewer. To note, we have updated all section numbers in Section 2. This was due to an erroneous use of \subsubsection instead of \subsection.

2 Reviewer 2

2.1 Specific Comments

Comment

Page 7, lines 10-13. This makes sense only if there's no trend in the data. Did you check whether there are increases or decreases in land burning? Please clarify.

Response

The reviewer here requests clarity on the use of a decadal mean as a harmonization value for land-burning sectors rather than using the value in the most recent historical period as is done for anthropogenic emissions sectors. The variability of emissions species in these sectors is largely due to weather and other climatic variation. In the attached figure we show historical data for the past 25 years normalized to 2005 (the beginning of our data selection window). The harmonization window is shown in grey. While there are some inter-decadal trends, it is hard to observe any specific year-on-year trends. Furthermore, IAM scenarios do not model these year-to-year variations, but only provide a long-term trend. Thus, in order to capture these decadal trends without introducing noise from observations in any specific year, we harmonize to a decadal mean "climatology" as described in the text. We have added the below figure in the SI and referenced it for further clarity to the reader.

Comment

Eq. 3. If I_c is the growth rate, shouldn't the rhs be a sum rather than a multiplication? Please check whether this equation is correct.

C:7

Response

Equation 3 is correct, however the growth term was not clearly described in the text. It can more accurately be described as a growth factor. The text has been revised to clarify this, using a less ambiguous notation by introducing the growth factor as β . Equation 2, provides the annualized amount, as a multiplier, that the initial intensity should grow over time such that intensity will equal the convergence year intensity by the convergence year.

Comment

Eq. 5. Please define c'. I have problems understanding this equation.

Response

This equation simply normalizes results to the regional emissions as produced by a given model. c' here is an element of set R (all countries in a region), where the prime superscript is used to differentiate from the index c, which is used in the equation definition. This is common practice in describing elements of sets in order to differentiate the use of the index, thus any further comments here as to if this is still considered confusing would be appreciated (and we can incorporate accordingly). We have provided more clarifying text to describe this operation in the manuscript by explicitly stating it is a normalization.

Comment

Section 3.4. The manuscript does a good job describing the methodology for harmonizing the datasets, but it does not describe with the same level of detail the methodology

for obtaining the spatial distribution of emissions. I think it's important to add more detail on this methodology to better understand the results presented in section 3.4. In particular, the section describes different values of emissions for the different scenarios in different countries. Since population and GDP change drastically over time and across scenarios, it is very difficult to make sense of sentences such as ". . . SSP1-2.6, emissions across countries decline dramatically such that by the end of the century, total emissions in China, for example, are equal to that of the USA today". It'd be great if you help the reader by pointing out whether the results in emissions are driven mostly by changes in population, GDP or the other SSP drivers.

Response

We agree with the reviewer that more detail of the methodology is useful, and a companion paper describing this methodology in much more detail will be submitted to GMD (https://www.geosci-model-dev.net/special_issue590.html) within the week. It is important to note, however, that different models have different regional resolutions. For example, IMAGE is used for SSP1-2.6 and models the USA and China directly. Therefore, there are no downscaling effects associated with the results cited. We strove to discuss the country-specific results where relevant for models in which they are explicitly accounted for. For models which provide country-specific resolution, there are myriad drivers which result in the analyzed emissions trajectories, including not only population, GDP, etc., but also inertia in socioeconomic and energy systems which are endogenously included by the models. Thus, it is hard (if not impossible) to point to a specific subset of these drivers as the primary constituents inducing such changes. In any case, we agree that this can be further clarified, and as such have added the following text to Section 3.4:

Original submission -

C9

CO2 and CH4 are well-mixed climate forcers (Stocker et al., 2013) and thus their spatial variation have a higher impact from a political rather than physical perspective. Aerosols, however, have substantive spatial variability which directly impacts both regional climate forcing via scattering and absorption of solar radiation and cloud formation as well as local and regional air quality. Thus in order to provide climate models with more detailed and meaningful datasets, we downscale emissions trajectories from model regions to individual countries using the methodology described previously in Section 2.4, which are subsequently mapped to spatial grids (Feng, 2018). We here present global maps of two aerosol species with the strongest implications on future warming, i.e., BC in Figure 9 and sulfur in Figure 10. We highlight three cases which have relevant aerosol emissions profiles: SSP1-2.6 which has significantly decreasing emissions over the century, SSP3-7.0 which has the highest aerosol emissions, and SSP3-LowNTCF which has similar socioeconomic drivers as the SSP3 baseline but models the inclusion of policies which seek to limit emission of near-term climate forcing species.

Revision -

CO2 and CH4 are well-mixed climate forcers (Stocker et al., 2013) and thus their spatial variation have a higher impact from a political rather than physical perspective. Aerosols, however, have substantive spatial variability which directly impacts both regional climate forcing via scattering and absorption of solar radiation and cloud formation as well as local and regional air quality. Thus in order to provide climate models with more detailed and meaningful datasets, we downscale emissions trajectories from model regions to individual countries. In most cases, models explicitly represent countries with large shares of emissions (e.g., USA, China, India,

etc.). MESSAGE-GLOBIOM and REMIND-MAGPIE are notable exceptions; however, their regional aggregations are such that these important countries comprise the bulk of emissions in their aggregate regions (e.g., the MESSAGE-GLOBIOM North America region comprises the USA and Canada). For regions constituted by many countries, country-level emissions are driven largely by bulk region emissions and country GDP in each scenario (per Section 2.4). Afterwards, country-level emissions are subsequently mapped to spatial grids (Feng, 2018). We here present global maps of two aerosol species with the strongest implications on future warming, i.e., BC in Figure 9 and sulfur in Figure 10. We highlight three cases which have relevant aerosol emissions profiles: SSP1-2.6 which has significantly decreasing emissions over the century, SSP3-7.0 which has the highest aerosol emissions, and SSP3-LowNTCF which has similar socioeconomic drivers as the SSP3 baseline but models the inclusion of policies which seek to limit emission of near-term climate forcing species.

Comment

Conclusions. This section is relatively long and reads more like a summary. It can be improved by shortening it, focusing only on the main take home messages of the manuscript, and not summarizing it.

Response

We agree with the reviewer that the conclusions section could be more concise. Thus, we have revised the first few paragraphs in order to encapsulate the primary points of interest to the reader, while keeping the important points regarding data usage and availability.