We thank the two referees' useful and constructive comments and detail our responses as below. We have revised the manuscript substantially, taking into account all comments, and tried to address all issues raised; we hope that the manuscript has been greatly improved in both its scientific value and the presentation. Significant changes include the inclusion of detailed CO and HCHO budgets, allowing the model differences and the differences in the model response to changes in biogenic emissions to now be more quantitatively addressed. While we aim to address every comment, there are some limitations that we cannot fully resolve at this stage, but we will try to follow up in a future study. In our revised manuscript, the modified text is shown in bold, and the new/modified tables and figures are highlighted. The referees' comments are copied below, with our responses in bold.

Anonymous Referee 1

The purpose of this manuscript is to investigate the causes of model biases for CO and formaldehyde in the southern hemisphere. Overall, this is a well written and very thorough manuscript, but I do have one major concern. I would like to see more discussion on the chemistry that, I believe, will significantly improve the presentation and overall impact of this manuscript.

We greatly appreciate the referee's very thoughtful comments and have revised the manuscript substantially to include chemical production and loss terms for CO and HCHO and to illustrate the differences of these terms among models and between the two simulations carried out in this study. (Section 5)

1) It would be helpful to have tables clearly showing the southern hemispheric budgets for CO and formaldehyde for each model. A graphic showing how the budgets change monthly may be instructive.

We have included the detailed CO and HCHO budget terms globally and for the southern hemisphere (divided into three sub-regions), where the model data are available, and these are shown in three new tables (Tables 4-6) in the revised version. We also include a new plot showing the monthly budgets from all available model output (Fig. 20).

2) In the same vein, it would be helpful to show the major contributors to the chemical production and loss terms for CO and formaldehyde for each model and how they vary in time. How do the four models' Henry's Law constants for formaldehyde compare? My recollection is that formaldehyde is not highly soluble, so there could be some sensitivity to the choice of Henry's Law constant. Could one model do two simulations, one each using the high and low of the uncertainty range?

As stated above, we have included new tables showing detailed CO and HCHO budgets. HCHO is moderately soluble therefore wet deposition of HCHO is an important loss process. Very similar effective Henry's Law Coefficients are applied in each model although TM5 takes into account hydration once HCHO enters the solution, resulting in an increased effective solubility. From the available HCHO budget terms in the new Table 5, wet deposition accounts for about 10% of total loss in TM5 and about 3% in NIWA-UKCA, which we now state in the text. The absolute differences in wet deposition of HCHO are small compared to other loss terms, i.e. oxidation by OH and HCHO photolysis. Adopting the TM5 approach in other models, or even one, is not a trivial task at this stage so we leave this to a possible future investigation. We agree with the referee that there is room for differences in this respect between the models. However, such differences are not big enough to explain the significantly low modelled HCHO in the SH compared to the column measurements. (Section 5, page 29)

3) There is considerable measurement uncertainty associated with the isoprene oxidation scheme. You say that looking into this issue is "beyond the scope in this study" and cite Archibald et al. (2010). I recommend that you at least provide a paragraph or two discussing the current uncertainties in individual reactions reported in the literature and hypothesize how they may or may not be relevant to your model exercise.

We have now considerably expanded the discussion of the impact of differences in the isoprene oxidation schemes. Considering the complexity of these schemes, we are unable to fully answer the question of how individual differences in these schemes affect HCHO and CO production rates. (Section 5; Page 35 Section 7)

Minor Comments

There are too many typos and grammar issues.

We have gone through extensive proofreading of the revised version.

Why don't you use MOPITT CO data to constrain the models?

We have now shown comparisons between MOPITT CO and modelled CO for January and September 2005 in Fig. 8 and their differences in Fig. 9, and added some discussion in Section 3.2.

Anonymous Referee 2

General comments

This paper presents the results of a modelling study simulating carbon monoxide and formaldehyde concentrations in the Southern Hemisphere, using two different biogenic

emissions estimates and four atmospheric chemistry and transport models, over a period of 5 years (2004-8). Model performance was evaluated through inter-model comparisons as well as observational data from surface stations. This is a timely and well-structured study. The use of atmospheric columns of a variety of marker compounds to constrain biogenic emissions estimates and elucidate the skill of atmospheric chemistry models in simulating the oxidation and transport of these species and their products is an important method in regions of sparse observations (ironically those with the highest emissions). The inclusion of simulations with artificial tracers to disentangle the contributions of transport, primary emissions and secondary production of CO is particularly nice. Given recent work indicating the influence of biogenic precursors and their climatically and air quality active products, such an evaluation is urgently needed. This study marks and important first step in this process. Overall, the study is well presented, described and discussed. I do have concerns over the level of detail provided regarding certain aspects of the work, and these are outlined below. In particular, I feel that the analysis is not of sufficient depth, especially with regard to an attribution of the observed differences to the chemistry mechanisms included in the four atmospheric chemistry models used. There are also sections where the English is poor, particularly earlier in the paper, and the manuscript is in need of a thorough editing.

We thank the reviewer for the very detailed review and many useful comments. We are encouraged by the positive view of the usefulness this study to the community, especially the use of column measurements in this vast and data sparse region. We have carried out a substantial revision, also in accordance with Reviewer 1's comments, by including detailed CO and HCHO budgets and analyses of differences in chemical production and loss terms in each model where the data are available. We have gone through thorough proof-reading and revision to improve the presentation of the paper.

Specific comments

2. Model simulations and observations One of my chief concerns with the paper is the level of detail provided regarding the model set-up and simulations. The information provided here for the processes, boundary conditions and driving data for each model is not consistent, making comparison difficult. The authors refer to previous model descriptions too often with regard to key processes. While it is extremely difficult to ensure full consistency between models, the paper would benefit from greater elucidation of some of the inconsistencies that might affect the conclusions drawn from their results.

We will address these comments below.

2.1 Simulations

Biogenic emissions estimates are strongly dependent on land cover (i.e. vegetation distribution), and driving meteorology. Presumably, although the two inventories are generated off-line, it is done with the "default" input fields for MEGAN-CLM and LPJGUESS. The authors should state clearly here the input data used, as well as the spatial and temporal resolution of the emissions inventories generated, as this is a clear source of inconsistency with the chemistry models that are then driven with these emissions. For some of the model set-ups, the emissions will be more compatible with the land surface and meteorology driving data for the chemistry and transport than others. Differences in land cover, for example, will also affect deposition of the primary emissions and their oxidation products; differences in meteorology will affect chemical production and loss rates. It is possible that some of the observed differences between simulated concentrations of oxidation products are a result of these inconsistencies.

We agree that inconsistencies exist in meteorology between the fields used to generate the biogenic emissions and those used to drive the chemical transport models (CTMs) and chemistry-climate models (CCMs). CLM-MEGANv2.1 emissions were generated offline using the Community Land Model (reference given

in the text), forced by meteorological reanalyses for each year (CRUNCEP; http://dods.extra.cea.fr/data/p529viov/cruncep/readme.htm). . LPJ-GUESS was generated using the CRU TS 3.1 climate data (eccad.sedoo.fr). It would be impractical, within the constraints of this study, to harmonize the driving meteorology in global models, especially in multimodel intercomparion studies given the complex nature of such models. It is precisely the purpose of this study to minimize the differences from the dependence of biogenic emissions on the driving meteorology that we chose to use prescribed monthly mean biogenic emissions in the model rather than calculate the isoprene emissions online (e.g. GEOS-Chem and CAM-chem both have the option to calculate isoprene emissions interactively). Also we speculate that differences in the driving meteorology might be a smaller factor in defining the biogenic emissions than differences in assumptions about biological factors. (Page 7 Section 2.1)

The use of model specific natural emissions of ocean CO is inconsistent with the use of specified biogenic emissions. Given that one aim of the study is to assess the impact of different biogenic emissions on chemistry transport model output and that anthropogenic emissions are the same for each model, surely this would have been a simple difference to eliminate.

We use the same emission dataset for ocean CO and we have clarified this in the revised version. (Page 7 Section 2.1)

Furthermore, in the descriptions of the GEOS-Chem and CAM-chem model simulations, the authors state that diurnal variability was imposed on the biogenic emissions inventories. Was this also done for NIWA-UKCA and TM5? And why? What was the temporal resolution of the biogenic emissions calculated by MEGAN and LPJ?

Yes, the diurnal variability of isoprene emissions are also imposed in NIWA-UKCA and TM5. We use monthly mean CLM-MEGANv2.1 and LPJ-GUESS emissions so we impose the diurnal variation on this emission dataset. This is done routinely in global models when using prescribed monthly mean isoprene emissions. Relevant text has been added. (Section 2.2)

Likewise the use of model specific lightning NOx emissions is inconsistent with the approach of harmonizing emissions used to drive the four chemistry models. NOx levels are critically important to the atmospheric oxidation of volatile organic compounds. Again this would have been a simple source of discrepancies between models to eliminate from the study. For both lightning NOx and ocean CO emissions, I would like to see further justification of the use of model specific inventories. How substantial are the differences between the models, and how sensitive are the models to these inputs?

We use the same ocean CO emissions (now clarified). Yearly lightning NOx emissions are now shown in Table 1, organised from the lowest to the highest used in the models; we can say that there are no substantial inconsistencies among the models with values in CAM-chem (3.3-3.6 TgN yr-1) being moderately lower than from other models (5.8-6.5 TgN yr-1). Most tropospheric chemistry models now use an interactive lightning-NOx scheme so making these emissions non-interactive would have required significant additional programming in some of the models, with limited impact on our work which focuses on HCHO and CO.

2.2 Models

The details provided about the four atmospheric chemistry and transport models used in this study are not consistent. In addition to the temporal and spatial resolution of the model and the driving meteorology, other aspects off particular importance to VOC oxidation and atmospheric concentrations are imposed methane concentrations, the dry deposition

scheme and the treatment of partitioning of gas-phase condensable species to the aerosol phase. The level of detail of the chemistry mechanism is also critical. At the very least, the authors should specifically (i.e. not simply by referring the reader to previous model descriptions) provide information of all of these in either the text or Table 2 (or both). Details of the chemistry mechanisms should include the number of chemical species (transported, reactive, fixed) and reactions (photolysis and thermal) in the full mechanism, and in the isoprene and monoterpene oxidation schemes. Furthermore, details of relevant compound lumping (e.g. are methacrolein and methyl vinyl ketone treated separately or lumped as a single species) should be provided, particularly with regard to species within the biogenic chemical degradation paths, and related species such as methanol and acetone. I would also like to see the authors elucidate further on the treatment of organic nitrates, PAN, peroxy radicals and (in the case of isoprene) epoxides. Monoterpene chemistry, in particular, varies considerably between atmospheric chemistry models and a full description of the schemes used within each model would be useful in the form of an Appendix or as Supplementary Information.

We appreciate the reviewer's suggestions in giving full model details, especially regarding chemical mechanisms used in the models. We now include a table and added text showing many of the relevant details of the four models, including reference to the isoprene oxidation schemes used. Further detail of chemical mechanisms is difficult to include given the complexities of the model physical and dynamical processes and the large number of chemical species and reactions (e.g. hundreds of chemical reactions). Such a comprehensive description of the models is beyond the scope of this paper and would also not necessarily help in understanding the causes of differences in model behaviour. All four models have been clearly and comprehensively described in various publications which we cite in this paper.

Our focus here is not mainly on comparing the differences in mechanisms but the effect of applying different mechanisms; such effects are reflected in the budgets

of the various production and loss terms which we have added in the revised version (Tables 4 and 5, and the new Section 5 of the text). Uncertainties arise from the different chemical mechanisms implemented in the model as well as the physical processes, and all processes interact to produce the differences in modelled chemical species. Short of any targeted investigation, it is not straightforward to identify in detail which process/processes might be responsible for the differences in the modelled species. The detailed differences in chemical mechanisms are best investigated in a box model where one can control the variables, as we now state in the text. Such an approach has been exemplified in the work of Archibald et al. (2010), cited in our paper, which includes all four models used in this study. We do not think that elucidating the different treatments of organic nitrates, PAN, peroxy radicals and epoxides here will be able to explain the differences in modelled CO and HCHO alone. As stated above, we have revised the paper and added detailed budgeting terms for CO and HCHO and this allows a more complete description of the model differences in chemistry. (Section 5)

2.3 Observations

Please give details of the temporal resolution of the raw measurement data so that this can also be clearly compared to that of the models.

Column data are in daily mean format where there were measurements on that day. Indeed, we compare the modelled daily mean output to that measured on the same day when the measurements were carried out. We take monthly mean observed surface CO data and compare with the monthly mean model data. This has been more clearly stated in the revised version.

Sections 3-6 My other main concern is the lack of detailed consideration and discussion of the differences in chemistry between the models and the impact this has on the simu-

lated HCHO and CO concentrations. It seems that it should be straight-forward enough to provide details of the main chemical sources and sinks of HCHO and CO in the various chemistry schemes, and the relative production and loss rates to each. How do these differ between low and high NOx environments in the models? This lack of detail makes these sections extremely unsatisfying. The authors speculate on numerous possible causes for the differences between modelled and measured HCHO, but do not explore any of these further. I recommend that they select the most likely causes and perform some sensitivity tests to establish whether it is indeed a contributor to the persistent under-estimation of HCHO in these models.

We have now included the detailed budget terms for CO and HCHO in global, SH and 3 sub-regions of the SH, as well as a new section (Section 5) detailing these terms. We have also deleted some of the speculations as to what may cause the low bias of modelled HCHO other than citing the previous studies, and we rather leave this (finding the cause) to a future study. Unfortunately we do not have sufficient resources at the moment to further explore the various causes and to carry out further sensitivity studies. We also suggest that more targeted observations might lead to the cause of low modelled HCHO in the remote SH. (Section 3.4)

3. Comparison between models and observations Again, there is a lack of consistency, this time between sites. While it is unavoidable that Darwin has only total column data, while partial columns are preferable for this application, the authors need to consider the likely effect that this has on the comparisons they are making and the conclusions that can be drawn. Ideally, I would like to see the authors show comparisons of total columns for another site to assess how model performance differs between total and partial column values. It could be expected that the use of total columns might tend to smooth differences between models as they have greater spatial coverage.

We can see the reviewer's point of view. But two of the four models do not have detailed stratospheric chemistry and one model has no saved output from the stratosphere so it is impossible to assess how realistic modelled total columns would be compared to the observed FTIR total columns. The observed total columns of CO include the CO production from the mesosphere which accounts for up to 10% in the high SH latitudes in the spring when the polar vortex, which is enriched in CO of upper-atmosphere origin, makes up a substantial part of the total air column. That is why we rather use observed partial columns (0-12km) for comparison. In the case of Darwin, the contribution of CO from the upper atmosphere is minimal given its latitudinal location. We believe that this is the best possible approach we can apply in comparing modelled and FTIR columns of CO. We have added some statements in the revised version.

4. Model differences in chemistry and transport Again there is an unnecessary inconsistency in approach. Previous evaluations of model performance had used partial (or total in the case of Darwin) columns. Here the authors choose to use tropospheric columns, with total heights set by the model specific tropopause.

For observed FTIR columns, we are restricted to the columns below 12km (for sites other than Darwin) due to the nature of the retrieved CO data, so we have to compare model results and observations in a like-for-like manner. However, when we compare the model differences we are not restricted to the region below 12km. For global (and SH) comparisons, a 12 km cut-off would not be a good choice as some model levels are the stratosphere (polar regions) and some tropospheric levels will be missing in the tropics which make significant contributions to the tropospheric columns. Likewise, for tropospheric budget calculations we define the chemical tropopause in the model using their modelled ozone as a marker as commonly done in multi-model comparison studies (e.g. Stevenson et al. 2006). Again we believe that this approach

would not introduce substantial inconsistencies among the models, given the sharp gradients of the modelled ozone in the tropopause region.

Technical corrections

Title - The authors do not, to my mind, currently address model uncertainties in a quantitate enough way to justify including this in the title.

We have now changed the title to "Multi-model simulation of CO and HCHO in the Southern Hemisphere: Comparisons with observations and the impact of biogenic emissions".

Abstract – the Abstract contains far too many undefined acronyms (e.g. NOAA GMD, NDACC, etc.). Please either spell these out in full or define them here.

Done.

1. Introduction – the English is particularly difficult to understand in this section of the paper. I suggest the authors read it through and edit it appropriately. Here, and throughout the paper, abbreviations and acronyms (particularly for the models) are used without having been previously defined (e.g. MEGAN).

Text has been modified accordingly.

The section on p2618, L4-10, seems rather contradictory. The authors state that global models are suitable tools for investigations such as these, and follow this up with the assertion that "For instance, global atmospheric models systematically overestimate observed: ::", which seems to indicate their lack of suitability. Perhaps some rearrangement would

help – the authors are right to highlight the shortcomings of the tools they are using, and to identify them as suitable.

Revised

P2618, L17 – are the authors aware of any atmospheric chemistry models that are still using a value of 200 Tg y-1 for isoprene emissions? I was under the impression that as model chemistry had improved, this value had not been used since TAR.

We stated that this low limit was used in Stevenson et al. (2006) so it is not very recent.

P2618, L24-29 – I'm not sure that I understand the point the authors are making here, regarding the Marais et al. study and the sparsity of direct observations. Again, rephrasing would be helpful.

Revised.

P2619, L1-4 – Again, this section appears to contradict the previous section. Further details of the model set-up used would perhaps help. Were biogenic emissions included in this study? If so, were they also generated by MEGAN?

The IMAGESvv used in Vigoroux adopts isoprene emissions from Muller et al. 2008 and was generated based on MEGAN using ECMWF meteorology. The amount of yearly isoprene emissions are comparable to that of LPJ-GUESS used here. In Jones et al (2009) that we cited here the box modelling was carried out assuming initial isoprene concentrations and therefore no isoprene emission applied. The purpose of citing these two studies are that modelled HCHO in the remote SH has been un-

derestimated by previous studies. We have moved the citations to Section 3.3, and modified the text.

P2619, L14 – does the "They" used here refer again to the Morgenstern et al., study

Yes. Clarified.

2. Model simulations and observations

P2621, L2 – Please state the % difference in monoterpene emissions as has been done for isoprene.

Added in page 7.

P2622, L10 – Figure 2 does not show differences; it shows the absolute values from two models side-by-side.

Text corrected.

P2623, L13 – Please provide further information here (or elsewhere in this paper) of the simple monoterpene oxidation scheme used.

Added in page 9.

P2623, L13-4 – Describe further how this gradient is implemented.

Two different constants are applied for the NH and SH respectively. Clarified. Page 9.

P2624, L19-21 – How does this assumed yield of CO from monoterpenes compare with yields in the other chemistry schemes? What is the assumed timescale of this conversion, and again how does this compare?

This is a long-standing assumption in GEOS-Chem, adopted due to the absence of explicit monoterpene oxidation and justified in the references given in the text. We do not have sufficient diagnostics to calculate the specific yields of CO from monoterpenes from other models for this study, therefore cannot compare them in this paper.

P2628, L20 – The sites are not shown anywhere in the paper. However, I would very much like to see them shown on a map to better understand their locations and the differences between them.

A new figure (Fig. 3) has been added to show all locations of ground-based measurement sites used in this study.

3. Comparison between models and observations

P2628, L17-21 – Could the differences in direct CO emissions from the ocean (different in the various models) also be a factor?

No. They are the same among models, and this has now been clarified in the text.

P2628 – Please include a direct comparison between modelled and FTIR CO columns for LPJ simulations, either as an addition to Figure 3 or as a new figure.

Now new Fig. 5.

P2629, L2-3 –Were the emissions inventories generated from the same meteorological driving data (in which case, the model columns might be expected to show similar seasonal and inter-annual variations)? Or, if not, please could the authors indicate how seasonal and inter-annual differences in relevant meteorology compare between the different models used here.

The emissions inventories were generated using similar meteorology (see the previous response). The two emission datasets show similar seasonal variation but do not show close correlation between their inter-annual variations in some source regions (see Figure 1). We do not know if such differences are caused by differences in meteorology and/or other forcing that constrains the models. We are not able to compare the different meteorology used in the models and those for generating the emissions which have been provided using entirely different models. However such differences in the various meteorological forcing adopted in each model should have similar features in terms of seasonal and inter-annual variations in general, because most of the climate forcing used in the models are derived from the observed data. The exception is that NIWA-UKCA is a free-running general circulation model. But this model is constrained by monthly observed sea surface temperatures, and calculated chemical species concentrations do not feed into the dynamics. This is to minimise inconsistencies with other models that are either entirely driven by prescribed meteorology or nudged towards the observed meteorology.

P2629, L5 Was the multi-annual ensemble mean data derived as daily or monthly averages? If monthly, why (given that the earlier comparisons were made with daily data)? If daily, why is daily data not presented in Figure 5?

The multi-annual ensemble mean data were shown as monthly averages in the new Figure 7 so the seasonality is more clearly shown. The monthly data are the average of daily mean data of course. The purpose of showing monthly mean (averaged from

daily data) is to show the seasonality of the deviation of modelled data from the observed data. Observed monthly data are also generated from the daily data.

P2629, L17 – Please make clear in the text that these are annual averages (presumably).

Done.

P2629, L18-19 – Why should the magnitude of the model biases have this order? Could the authors speculate here, or refer the reader to the section in which this is explored.

See revised text on page 16.

P2629, L20-22 – Again please speculate on the possible reasons for this seasonal difference between model variability.

See revised text on p16.

P2629, L26 – Is the greater inter-model variability due to the relative scale of the biogenic emissions from each of MEGAN and LPJ? Have the authors tried normalizing the biases by the magnitudes of the emissions inventories?

We have revised the text to indicate that this is "primarily due to the larger response of modelled CO to its higher precursor emissions". We have not normalized the biases.

P2630, L8 and elsewhere – Please provide % differences as well as absolute.

We have modified the text accordingly. (Page 18)

P2630, L19-25 – It seems that this could be easily checked by selecting only the model data from dates that coincide with the sampling dates at Cape Grim. This would provide a much more rigorous evaluation in any case.

We have now replaced Cape Grim data with the Pacific Ocean data which is more representative of background atmosphere of the SH, and is consistent with comparisons from other sites.

P2630, L26-7 – Please speculate as to why NIWA-UKCA and GEOS-Chem should perform so well when compared against surface observations in spite of the fact their deviations were not always the lowest when assessed against CO columns.

We should clarify that models have systematic low biases compared to the FTIR column CO. Therefore the good agreement between modelled and observed surface CO are not entirely reflected in the column comparison. The modelled column CO are attributed to not only surface distribution but also the vertical distribution which is controlled by the vertical transport of both CO and its precursors. We have noted this point in the revised text. (Page 18 in Section 3.3)

P2631, L5-7 – What are the characteristics of the sites for which LPJ performs better than MEGAN? Why might this be the case?

It is clearly shown that modelled surface CO shows a lower bias from the simulation with LPJ-GUESS than that with CLM-MEGANv2.1 emissions in all cases but this does not mean that, in absolute terms, the MEGAN simulation is better than GUESS in all locations and seasons. If in the case MEGAN has a high bias (e.g. at Ascension Island in JJA), the GUESS simulation is closer to the observed surface CO. We cannot characterise the sites where GUESS performs better as there is no statistical significance due to the limited locations selected.

P2631, L9-13 – See above regarding the anomalies at Cape Grim. The strong positive bias here provides further motivation for removing polluted dates from the analysis and selecting only the sampling dates from the model output data. For the purposes of transparency, the authors could show the current time series (i.e. with all dates included) on the same plot.

We have now replaced the comparison for Cape Grim surface CO with surface CO from the "Pacific Ocean" site, which is more representative of the remote SH. This is because unlike at Cape Grim, in the models we do not only sample air during clean-air days.

P2631, L22-28 – Does Figure 8 show model simulations driven with emissions data from MEGAN or LPJ? Please make it clear here and in the caption of Figure 8.

Thanks, done. They are MEGAN simulations.

P2632, L1 – Again what are the key differences between these sites that might drive that difference in variability.

Lauder is further from the source regions than Wollongong; this indicates that the difference in chemistry among the models dominates the HCHO variation in the more remote location. Relevant statements have been added in the text.

P2632, L6-7 – 10-15% is not a small difference. What is the % underestimation?

Texts have been modified to read as: "This resulted in smalloverall changes, with 10-15% increases that occur in some summer months, andthe increases were not sufficient to close the gap between the models and the observations." Note that we assume the maximum daily HCHO and 10-15% increase is the upper limit. We do not

have data with higher temporal resolution than daily mean from three models. (Page 19 Section3.4)

P2632 – Again please show the comparisons between the modelled and observed columns for the chemistry models driven by the second emissions inventory (presumably LPJ again) either as an addition to Figure 8 or as an additional figure.

It does not add much new information to show the comparison between observed and modelled HCHO with LPJ-GUESS emissions because the two simulated HCHO fields are almost indistinguishable, as also shown in Figure 13. We have too many plots already together with some new plots so prefer to leave as is.

P2632, L10 – the multi-annual monthly mean (?)

Yes. Clarified.

P2632, L23-5 – Have the authors tried to scale methanol emissions input into the chemistry models to determine whether this makes a substantial difference to HCHO concentrations?

We have now removed this discussion. We do not yet have observational data of methanol in this region so do not want to speculate at this stage. This could be the subject of future investigations.

P2632, L25 – P2633, L1 – This section is highly confusing and poorly explained. I would strongly urge the authors to include the full set of reactions to which they are referring. The authors jump from a discussion of methanol emissions to a reaction yield of HCHO from CH3OOH (acetic acid, rather than methanol). Please explain the connection between a reaction commonly included in methane oxidation in chemistry models and the methanol

budget. Are the specific reactions referred to here included explicitly in all of the models used in this study?

We have revised the text on this discussion and made some clarifications. The purpose was to cite the previous work on modelling HCHO albeit using a box model. We drew a common conclusion which is that models underestimate HCHO in the remote SH. (Page 20 Section 3.4)

P2633, L11 – What % of global terrestrial isoprene emissions?

Related text deleted.

P2633, L21-29 – Please give details of the chemistry included in the IMAGES model. How does it compare to that included in the models used in this study, and how do the conclusions drawn by Vigouroux et al. apply to the results here?

These points have been added/discussed in the revised text. Page 21.

P2634, L1-2 – Why should a reduced oxidative capacity result in an increase in HCHO?

Corrected - should be "lower OH in the model".

P2634, L9 - "differnces" should read "differences"

Corrected.

P2634, L21 – Please explain here rather than later in the text why January and September have been chosen for this analysis.

Revised. Page 22.

P2635 – Please provide the same figure for CO_{OH} as for CO₂₅ (Fig. 11)

Added. Fig. 16.

P2635, L25 – "towars" should read "towards"

Corrected.

P2636, L5 - "reflects" should read "reflects"

Corrected.

P2636, L15-6 – Please explain/speculate here why NIWA-UKCA should show such a small contribution from primary emissions, and relate this to the results/discussion already presented from comparisons of the partial CO and surface CO data.

The relatively small contribution from primary emissions to CO column in NIWA-UKCA is related to the relatively large chemical production compared to other models; this now has been shown in Table 4.

P2636, L21 – "agreement" would read better than "correspondence"

Changed.

P2637, L4 – "deduce" would read better than "derive"

Changed.

P2637, L25-9 – As noted previously, the authors should include full details of the isoprene and monoterpene oxidation schemes of the four chemistry models used in these simulations. Presumably, the authors do not mean that isoprene itself is lumped, but it would be instructive to see which of its products are lumped and how this differs between models.

The paragraph has been removed and is replaced with a section on the CO and HCHO budget analysis. We show that the underlying differences in chemistry are reflected in the rates of chemical production of CO and HCHO presented here, although we cannot identify in more detail what processes in the chemical mechanisms are responsible. Such investigation would be better studied to a box model in which parameters can be controlled. Page 26-29.

P2637, L25-29 – It would be helpful if the authors were to refer the reader back to the bottom row of panels in Figure 13.

This has now been replaced with new text.

P2638, L1 – Please state in the text that the profiles shown in Figure 14 are again in latitudinal zones.

Added in now Fig.19.

P2638, L1-3 – Why are there no observations included in Figure 14?

This figure only shows model-model comparisons; the model data are averaged over three sub-regions in the SH and are compared. The observational data we present in this study are only for specific locations, therefore are not appropriate for comparison in this figure.

P2638 – Although I appreciate that the authors have chosen to focus on output from the MEGAN simulations, I would suggest that it would be useful to include at least a discussion of the LPJ simulations and how they compare. Are the results in keeping with those from MEGAN (even if of lower magnitude)? Are there any notable differences that might highlight the behaviour of the chemistry mechanisms under different NOx:VOC ratios?

We have now added detailed CO and HCHO budgets for both simulations (new Section 5). Hence there is now more discussion on LPJ-GUESS results. Substantial revisions have been carried out as a result.

P2638, L8 – Again, inclusion of the detailed isoprene and monoterpene oxidation schemes employed by each of the models would greatly facilitate the discussion here. What are the relative rate constants for the photolysis of HCHO in each of the models? What are the production rates of CO and HCHO from the oxidation of the biogenic VOCs in each?

We now show reaction fluxes of HCHO photolysis from TM5 and NIWA-UKCA (Table 5), and find that the photolysis rates of HCHO are very similar in these two models. However we note that wet deposition of HCHO in TM5 is much higher than in NIWA-UKCA (10% and 3% relative to total chemical loss of HCHO, respectively). This is now discussed in Section 5.

P2638 – See above comment. It would be useful to see the production rates of CO and HCHO from OH oxidation of biogenic VOCs for each model. Ditto loss rates of HCHO to OH-initiated reactions.

We now include CO budget from all models and HCHO budget from two models as stated above. The chemical production terms include oxidation of methane and NMVOCs. We are not able to provide oxidation rates from biogenic VOCs here as not all relevant reaction rates have been saved.

P2638, L17-8 – What is the correlation coefficient for each model? Or the ratio of HCHO to OH?

We have not calculated such a correlation coefficient. In the remote SH, HCHO formation is mainly through methane oxidation; therefore, OH and HCHO are correlated.

P2638, L25 – I suggest that the authors could easily produce vertical profiles of H2O vapour for each model, or H2O/OH in line with the HO2/OH profiles in Figure 14.

Water vapour fields are very similar among the models as prescribed meteorological fields are used to drive the models. The differences in OH between the models are likely due to the differences in their photolysis schemes. This has been noted in the revised text. (Page 26.)

We have now removed HO2/OH ratio panel from Fig.19 as the conversion rates of NMVOC to CO is now better diagnosed with CO budget terms.

P2637, L28-9 – Again, the authors should calculate the production/loss rates for these reactions to determine which is the cause, or if it is a combination of the two.

Substantial revisions have been carried out based on calculated production /loss terms. Section 5.

P2639, L1-3 – Given the complexities of the chemistry involved, it is hard to see how the authors can draw this conclusion without having conducted a more detailed analysis than is presented here.

We have more quantitative discussion on this in Section 5.

5. Sensitivity of modelled SH CO and HCHO to uncertainties in biogenic emissions

Please could the authors explain why they have (again) introduced an unnecessary inconsistency into their analysis. In the previous section they chose to focus on January and September; here January and July. Please justify fully this decision, or ideally, change one of he sections so that the same months are discussed in both.

This has been explained in the revised text. Page 30.

P2639, L17-19 – The right-hand panels in Figure 15 appear to show increases in the CO columns in LPJ relative to MEGAN in some parts of northern and central Africa. However, the choice of colour scale is poor. Furthermore, as this is a discussion of the columns over the Southern Hemisphere, there is no reason for the plots to extend north of the equator.

Increases of CO in this region reflect the higher isoprene emissions in this region in the LPJ-GUESS inventory. The "meteorological equator" (i.e. the ITCZ) divides the atmosphere into northern and southern-hemisphere air masses but the ITCZ is often north of the geographical equator. We therefore prefer the plot to also cover some areas north of the Equator.

P2640, L2 – "reginos" should read "regions"

Thanks - corrected.

P2640, L4-6 – What is the relative change in biogenic emissions between the models for each season? Does this correlate with the changes in columns? What are the seasonal differences of NOx emissions in the region?

There are no changes in biogenic emissions between the models as the same emissions are prescribed in each model. The differences in columns here are due to differences in chemistry. Given the same changes in emissions, each models reacts in different degrees to such changes due to the underlining differences in chemistry in each model. We have removed this note to avoid confusion.

P2640, L15 – Please could the authors explain what they mean by spatial variations of the biogenic emissions. Is that variations between the two inventories? Are these the only two regions with marked spatial variations? At what resolution are these variations apparent?

We mean the increase of LPJ-GUESS isoprene emissions in these regions relative to MEGAN inventory. We have modified the text to make it clear. Page 30.

P2640, L23-29 – As these are both oxidation products of biogenic VOCs perhaps the authors could explain more clearly why one should be correlated and the other anticorrelated with OH concentrations. Again, this would be best done via an analysis of production and loss rates. What are the correlation coefficients?

We have replaced "anti-correlated" with "in opposite sign". We have also modified the text to clearly explain why in remote region models have lower CO with LPJ-GUESS simulation but higher OH and HCHO. As suggested by the reviewer, this has been backed up by the calculated production and loss rates.

P2641, L1-4 – Again, please quantify the relative speeds of the isoprene oxidation processes in each model, and present the mechanisms used in each.

Unfortunately we cannot provide directly the HCHO production from isoprene oxidation as we do not have sufficient diagnostics to quantify it. We assume that isoprene oxidation is a dominate component. Table 5 shows that oxidation rates of

NMVOCs in the SH are larger in TM5 than in NIWA-UKCA in both the absolute values and relative to the total chemical production rates.

P2641, L5-14 – Given the limited nature of the discussion of the zonal vertical profiles presented in Figure 18, I would suggest that showing the panels for January and July only would suffice. There is little difference between those for October and January in particular.

We understand the reviewer's concern but would still like to leave the plot as is showing all four seasons. There are some seasonal differences that are reflected both in differences in two emission inventories and the difference in transport pattern. We have added a sentence regarding seasonality in the revised text. Page 32.

6. Summary and conclusions

P2641, L23 – I would not describe -19.2% as comparing well, certainly not relative to the other locations.

Changed to "reasonable agreement" Page 32.

P2642 – The word "significant" has a specific statistical meaning. As the authors have not performed any tests of significance they should not use it in this context.

We have dropped the word "significant".

P2642, L9-12 – While the complexities of the system preclude a simple solution to the discrepancies between modeled and observed HCHO, a more detailed analysis of the differences between the chemical mechanisms and the HCHO production and loss rates to the various reactions included in each would at least indicate where future research should be directed.

Agreed. However the inter-model differences are much smaller than the differences between the models and the observations, indicating that the cause of such a large discrepancy goes beyond what differences in chemical mechanisms can explain, at least based on our study. We have modified the text accordingly. Page 35-36.

P2642, L17-20 - Please state here in the text the

Added. Page 33.

P2642, L21-2 – "neither" : : : "nor" should read "either" : : : "or" as the sentence is already negative.

Thanks - corrected.

P2642, L25-26 - "reduced" would read better as "reductions in"

Replaced.

L2642, L25-28 – What is the % difference between biogenic emissions in MEGAN and LPJ for these source regions?

Added in the text. Page 33.

P2642, L1-2 – As before, the inclusion of details of the oxidation processes and a more thorough analysis and discussion of how the differences between the models affect the production/loss of HCHO and CO would strengthen this assertion.

We now show production/loss terms and include a discussion of differences in production rates of CO and HCHO from the oxidation of NMVOCs between simulations using these two emission inventories.

P2642, L6 – Without the in-depth analysis of the chemical production and loss rates of HCHO in the various models, the authors are over-stating the case, by saying that they "show". Their results may suggest this, but they have not conclusively demonstrated causality.

This statement is now based on the detailed budget shown in Table 5.

P2643, L8-12 – To what extent do the models include the latest findings on isoprene/ OH in low NOx environments?

To various degree. For example, GEOS-Chem includes many recent updates on isoprene oxidation mechanism and does include formation of epoxide species in the model which regenerate OH under low NOx conditions. Epoxides are not included in the other models. This point has been noted in the revised text. Page 35.

P2644, L1-2 – It would be instructive for the authors to perform sensitivity tests with the models using the same methane fields for all. This would allow them to assess the effect of NMVOC chemistry differences alone.

Unfortunately it is not realistic for all models to perform such sensitivity tests. We have now included mean surface methane concentrations used in each model (see Table 2) and hope it will satisfy. The differences in methane among the models are not sufficient to explain the differences in NMVOC chemistry.

P2644, L3 – "distributions" should read "distributions"

Corrected.

P2644, L3-18 – This entire section would be strengthened if the authors included the analyses of the biogenic VOC oxidation mechanisms, together with production and loss rates, as suggested previously.

Agreed. We have now quantified the production and loss rates through detailed budget calculations. Text modified accordingly.

P2644, L19-20 – a more detailed analysis of the sources of the differences in modelled CO and HCHO concentrations between the mechanisms, might result in an improvement in these mechanisms, allowing a more robust use of HCHO and CO columns to constrain biogenic emissions and reduce this uncertainty.

Text modified to reflect this comments Page 36.

Table 1 – The text in section 2 states that the models use internally generated emissions of lightning NOx and ocean CO emissions. These emissions should be clearly stated somewhere either in the text or in this table, which currently implies identical CO emissions across all models.

Lightning NOx emissions have been added in Table 1. Ocean CO emissions are the same for all models, as is now stated in the revised text.

Table 1 – While C5H8 and C10H16 have been identified within the Introduction as isoprene and monoterpenes respectively, these compounds are referred to almost exclusively by name rather than formula in the text. Please use the names here.

Modified.

Figure 3 – The title of 3rd panel should read "Wollongong"

Corrected.

Figure 3 – The fourth panel would benefit from being (vertically) larger so that differences between the models can be more clearly seen.

We can see the reviewer's point of view. However we would like to keep the plotting aspects the same for all panels. The inter-model differences are relatively smaller at this site (Darwin) compared to at other sites.

Figure 5 – Why are the observations plotted as symbols and the model output by lines? If the observations and model output is equivalent then they should all be depicted in a consistent way (i.e. all as symbols or lines).

The symbols here are not observations but the deviations of model ensemble mean from the observations. We have now changed all plots to lines.

We ahve also modified now new Fig. 11 to use lines.

Figure 15 – The choice of colour scale is poor. It is particularly hard to distinguish between the different reds used for positive changes, and for the greens used for changes of -5-10% and -10-15%. Why do the panels extend north of the equator when this paper is focused entirely on the Southern Hemisphere? Ending the plots at 0S would help with the issue of the scale, as well as removing the area in northern Africa where the changes are strongly positive.

We have now changed the colour scale, addressing the reviewer's concern. We don't entirely agree with the reviewer's suggestion regarding the domain of the plots

for the reasons we stated before. Therefore we would like to leave the plots as they are to include the domain 0-30N.

Figure 15, Caption – What do the authors mean by "multi-annual and ensemble mean"? Surely the plot only shows the multi-annual mean for each of the four models?

We have corrected this - thanks for pointing out.

Figure 16 – See the comments regarding the colour scale and geographical extent of Figure 15.

Colour scale has been adjusted in now Fig. 22.

Figure 17 – As above.

Colour scale changed in now Fig. 23.

Figure 18 – While I realise the contours are clearly labelled this figure would again benefit from a better choice of colour scale.

Colour scales have been adjusted in now Fig.24.

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Multi-model simulation of CO and HCHO in the Southern Hemisphere: Comparison with observations and impact of biogenic emissions

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Abstract

We investigate the impact of biogenic emissions on carbon monoxide (CO) and formaldehyde (HCHO) in the Southern Hemisphere (SH), with simulations using two different biogenic emission inventories for isoprene and monoterpenes. Results from four atmospheric chemistry models are compared to continous long-term ground-based CO and HCHO column measurements at SH Network for the Detection of Atmospheric Composition Change (NDACC) sites, the satellite measurement of tropospheric CO columns from the Measurement of Pollution in the Troposphere (MOPITT), and in situ surface CO measurements from across the SH, representing a subset of the National Oceanic and Atmospheric Administration's Global Monitoring Division (NOAA GMD) network. Simulated mean model CO using the CLM-MEGANv2.1 inventory is in better agreement with both column and surface observations than simulations adopting LPJ-GUESS emissions, which markedly underestimate measured column and surface CO at most sites. Differences in biogenic emissions cause large differences in CO in the source regions which propagate to the remote SH. Significant inter-model differences exist in modelled column and surface CO, and secondary production of CO dominates these inter-model differences, due mainly to differences in the models' oxidation schemes for volatile organic compounds, predominantly isoprene oxidation. While biogenic emissions are a significant factor in modelling SH CO, inter-model differences pose an additional challenge to constrain these emissions. Corresponding comparisons of HCHO columns at two SH mid-latitude sites reveal that all models significantly underestimate the observed values by approximately a factor of 2. There is a much smaller impact on HCHO of the significantly different biogenic emissions in remote regions, compared to the source regions. Decreased biogenic emissions cause decreased CO export to remote regions, which leads to increased OH; this in turn results in increased HCHO production through methane oxidation. In agreement with earlier studies, we corroborate that significant HCHO sources are likely missing in the models in the remote SH.

1 Introduction

Carbon monoxide (CO) is ubiquitous throughout the troposphere and is an important ozone (O_3) precursor; it originates from both primary emission sources (fossil fuel and biomass combustion, biogenic and oceanic processes) and in-situ chemical production. The dominant chemical source term in the troposphere is the photo-oxidation of methane (CH_4) and non-methane volatile organic compounds (NMVOCs) (e.g., Duncan et al., 2007). Its principal sink is the reaction with the hydroxyl radical (OH), hence CO plays a key role in controlling the oxidizing capacity in the atmosphere (e.g., Levy, 1971). The oxidation of methane and NMVOCs, such as isoprene (C_5H_8) , monoterpenes $(C_{10}H_{16})$, acetone (CH_3COCH_3) and higher aldehydes, leads to the formation of formaldehyde (HCHO), which, through photolysis and reaction with OH, is the major chemical source of CO (Atkinson, 2000). Once formed, CO has a relatively long lifetime of around 1–2 months, and therefore it is often used as a chemical marker for characterizing the long-range transport of air pollutants away from important source regions (e.g., Staudt et al., 2001; Heald et al., 2003; Liang et al., 2004; Fisher et al., 2010).

Due to a lack of strong regional emission sources, the Southern Hemisphere (SH) acts as a global sink for many of the polluting trace species emitted in the tropics, where polluted plumes are transported away out over the relatively clean ocean becoming subject to chemical processing. The relatively low population density, and thus low anthropogenic activity, in the SH means that direct emission sources of CO are principally limited to biomass burning (BB) and direct biogenic processes (e.g., Swinnerton et al., 1970; Watson et al., 1990; Fishman et al., 1991). Satellite and ground-based observations of CO in the SH have been used to identify the effect of BB and its footprint through long-range transport in the SH, which dominates the CO seasonal cycle there (e.g., Rinsland et al., 2005; Edwards et al., 2006; Gloudemans et al., 2006; Morgenstern et al., 2012). Global distributions of HCHO are much more inhomogeneous than CO, due to the much shorter lifetime of HCHO (on the order of a few hours), and the concentration of HCHO drops off sharply away from the source regions. Observations of HCHO are commonly used to constrain isoprene emis-

sions in high-emission regions, because it is a high-yield product of isoprene oxidation (e.g., Palmer et al., 2003; Shim et al., 2005; Barkley et al., 2008).

Global chemical models have been extensively used to estimate the sources and sinks of CO (e.g., Holloway et al., 2000; Duncan et al., 2007). However, systematic discrepancies between modelled and observed CO still exist, with models generally underestimating CO in the more polluted NH and overestimating CO in the SH (e.g., Shindell et al., 2006; Naik et al., 2013; Stein et al., 2014). In the remote SH, however, the extremely low HCHO concentrations are expected to further complicate the comparisons of model results with observations.

In contrast to the anthropogenic emissions dominating CO sources in the more polluted NH, biogenic VOCs are important sources of CO and HCHO in the SH, and isoprene oxidation contributes significantly to the regional CO and HCHO abundances in this region (Pfister et al., 2008). However, large uncertainties exist in biogenic emissions inventories, in particular for surface fluxes of isoprene and monoterpenes (Arneth et al., 2008). Bottom-up estimates of annual isoprene emissions vary between 400 and 600 Tg C yr $^{-1}$ (Arneth et al., 2008), and the typical range of annual total isoprene emissions implemented in global atmospheric chemistry models is ~ 200 –600 Tg C yr $^{-1}$ (Stevenson et al., 2006). The effect of such uncertainties in biogenic emissions on SH composition, such as CO and HCHO, has not been adequately assessed. Moreover, the sparsity of the ground-based CO and HCHO measurements in the SH also limits our ability to constrain these biogenic emissions.

In this study, we perform a number of simulations using an ensemble of chemical transport models (CTMs) and chemistry-climate models (CCMs) as part of the Southern Hemisphere Model Intercomparison Project (SHMIP), to compare modelled CO and HCHO to observations and to investigate the factors that influence the distributions of CO and HCHO in the SH. Given the relatively low anthropogenic emissions in the SH and the dominance of biogenic emissions of VOCs (mainly isoprene), we determine the influence that different emission inventories of isoprene and monoterpenes have regarding their effects on modelled CO and HCHO columns in the SH. Satellite observations

of SH CO usually are in good agreement with ground-based observations, however the data quality of the satellite data deteriorates towards the poles (Morgenstern et al., 2012). Morgenstern et al. (2012) find that CO columns exhibit a large-scale mode of variability in the remote SH that does not exist in the NH. For our purposes, we make use of highprecision ground-based Fourier Transform Infrared Spectroscopy (FTIR) measurements of CO columns from four SH sites that have not previously been fully exploited for model evaluations, namely Darwin (12.43° S, 130.89° E) and Wollongong (34.41° S, 150.88° E) in Australia, Lauder (45.04° S, 169.69° E) in New Zealand, and Arrival Heights (77.82° S, 166.65° E) in Antarctica. We also compare the modelled HCHO columns to those observed by the FTIR instruments at Wollongong and Lauder. In a companion paper, Fisher et al. (2014) have evaluated the vertical gradients of CO from the SHMIP models in the vicinity of Cape Grim, Australia, which is representative of SH mid-latitude background air, using multi-year aircraft measurements available from the Cape Grim Overflight Programme (Langenfelds et al., 1996). The influence of both chemistry and transport on the modelled vertical gradients of CO are addressed. Although there are biases of various magnitudes across the different models, the seasonal variability and extent of the gradients in tropospheric CO are shown to be captured reasonably well, especially during the tropical BB season.

In this paper we address the sensitivity of CO and HCHO distributions in the SH to biogenic emissions of isoprene and monoterpenes as provided by the LPJ-GUESS inventory (Arneth et al., 2007a, b; Schurgers et al., 2009) and the MEGANv2.1 model (Guenther et al., 2012) across the models included in SHMIP. In Sect. 2 we provide model descriptions, the common emission inventories used to drive the models, and the observations used in the study. In Sect. 3 we compare results for the period 2004–2008 and show comparisons between modelled CO and HCHO columns and the FTIR measurements at the four SH sites mentioned above, **modelled and MOPITT CO columns**, as well as comparisons between modelled and observed surface CO. In Sect. 4 we analyse differences in the models' abilities to reproduce SH CO and HCHO columns, and the underlying differences in the models' chemistry and transport. **In Sect. 5 we further analyse the chem-**

ical production and loss terms to address differences in models' NMVOC oxidation mechanisms. In Sect. 6 we assess the sensitivity of modelled CO and HCHO to changes in biogenic emissions and the effect of such changes on the oxidizing capacity in the clean SH. Finally, in Sect. 7 we present our conclusions.

2 Model simulations and observations

The SHMIP intercomparison uses four global models, including two CTMs (TM5, GEOS-Chem) and two CCMs (CAM-chem, NIWA-UKCA). In this section we provide the description of the simulations performed, the common emission inventories employed, a brief description of each model, the meteorological drivers, and the observations used for evaluating the performance of the models.

2.1 Simulations

We perform simulations covering the period of 2004 to 2008 using a one-year spin-up for 2003. The two CTMs are driven by the meteorological analysis for the same period from their respective sources, whereas NIWA-UKCA uses observed sea surface temperature and sea ice data sets. CAM-chem runs in the specified-dynamics mode, using meteorological fields from the reanalysis data. Two simulations are performed in all models with identical emission inventories for the anthropogenic and BB components, but different inventories are adopted for biogenic isoprene and monoterpene emissions. We also include passive CO tracers in the simulation defined as having the same global primary, surface emission sources as CO, but with one having a fixed lifetime of 25 days and a second having the lifetime determined by OH distribution in each respective model. These tracers allow the differentiation of the inter-model variability with respect to transport of CO to the SH from the main source regions.

Although we have been careful to harmonize the emissions used across models, differences in the chemical mechanisms which are employed result in the aggregated emissions of the NMVOCs being somewhat different across the models. For anthropogenic emis-

sions, we adopt the yearly specific MACCity global emission estimates (Lamarque et al., 2010; Granier et al., 2011) nested with the Regional Emission Inventory in Asia (REASv2.1) for the East Asia region (Kurokawa et al., 2013). Interannually varying monthly mean BB emissions are taken from the Global Fire Emissions Database version 3 (GFEDv3) (van der Werf et al., 2010). For lightning-NO $_{\rm x}$ emissions, each model adopts individual parameterizations, which interact with the models' convection schemes. Natural emissions of soil-NO $_{\rm x}$ and CO from the ocean are taken from the POET database (http://eccad.sedoo.fr). The annual total emission fluxes for key species are listed in Table 1 for the simulation period of 2004–2008.

Biogenic emissions for isoprene, monoterpenes, CO, methanol, and acetone are based on MEGANv2.1 (Guenther et al., 2012) and are calculated offline using the Community Land Model (CLM4.0; Lawrence et al., 2011), driven by CRUNCEP reanalyses (http://dods.extra.cea.fr/data/p529viov/cruncep/readme.html) for each year. We refer to this dataset as CLM-MEGANv2.1 hereafter. We then replace the MEGANv2.1 emissions for isoprene and monoterpenes with the GUESS-ES emissions (http://eccad.sedoo.fr) in our second set of simulations for comparison (hereafter referred to as LPJ-GUESS simulations); this emission dataset is generated using a dynamical vegetation model LPJ-GUESS driven with CRU TS 3.1 climate data (Arneth et al., 2007a; Schurgers et al., 2009). None of the models currently include any higher terpenes. The yearly-varying annual global total emissions of isoprene from CLM-MEGANv2.1 (462-508 Tg vr⁻¹) are markedly larger than the LPJ-GUESS emissions (431–450 Tg yr $^{-1}$), i.e. \sim 5–10 % relative to CLM-MEGANv2.1 (Table 1). The corresponding differences are much larger for the SH between these two inventories, i.e. $\sim 10-20$ % relative to CLM-MEGANv2.1, over the same period. For monoterpenes, the annual total CLM-MEGANv2.1 emissions are substantially larger than the LPJ-GUESS emissions, i.e. a factor of 4 and 6 larger globally and in the SH, respectively. The two datasets have been generated from the respective land surface models driven by similar meteorological fields, as specified in the above references. Here, we do not harmonize the model meteorologies to those used in generating the biogenic emissions. Instead, we prescribe the

monthly mean biogenic emissions in the models to ensure the consistency. Fig. 1 shows the SH and regional monthly total isoprene emission fluxes from LPJ-GUESS and CLM-MEGANv2.1 for Australia and part of Indonesia (0–44° S, 94–156° E), Southern Africa (0–37° S, 9–44° E) and South America (0–57° S, 34–84° W), the regions with high isoprene emissions. The largest monthly emissions occur in austral summer in both datasets when the differences between these two datasets are also largest. Overall, the integrated CLM-MEGANv2.1 isoprene emissions (especially the summer maxima) are substantially higher than LPJ-GUESS emissions with the exception of Southern Africa, where LPJ-GUESS emissions are larger. **Fig. 2 shows the spatial distribution of** the fluxes for both isoprene and monoterpenes from the two emission inventories for January 2005. In Amazonia, tropical Africa, and Australia emissions are visibly larger in CLM-MEGANv2.1 than in LPJ-GUESS.

2.2 Models

2.2.1 NIWA-UKCA

NIWA-UKCA stands for the version of the UK Chemistry and Aerosols Model (UKCA) that is used and further developed at the National Institute of Water and Atmospheric Research (NIWA). The background climate model is similar to HADGEM3-A (Hewitt et al., 2011) with a horizontal resolution of $3.75^{\circ} \times 2.5^{\circ}$ and 60 vertical levels extending from the surface to 84 km. The physical processes in NIWA-UKCA, including interactive dry and wet deposition of the trace gases and the FAST-JX photolysis scheme, have been described in detail by Morgenstern et al. (2013) and O'Connor et al. (2014). Unlike the version described by O'Connor et al. (2014), here we now apply dry deposition following Wesely (1989) only to the bottom model layer rather than throughout the boundary layer. The model setup used here comprises a coupled stratosphere—troposphere chemistry scheme. The stratospheric reactions are the same as in Morgenstern et al. (2009) and include explicit chlorine and bromine chemistry. We have updated the NIWA-UKCA chemical mechanism from that described in Morgenstern et al. (2013) to account for emissions and degradations of

ethene (C_2H_4) , propene (C_3H_6) , methanol (CH_3OH) , isoprene, a representative monoterpene, and a lumped species that accounts for missing NMVOCs in the model similar to the approach taken in the IMAGES model (Müller and Brasseur, 1995), with primary emission sources. In addition to the above, the NIWA-UKCA chemical scheme includes nitrogen oxides (NO_x), CO, ethane (C₂H₆), propane (C₃H₈), HCHO, acetone (CH₃COCH₃), and acetaldehyde (CH₃CHO) as primarily emitted species (O'Connor et al., 2014). The isoprene oxidation scheme is the mechanism described by Pöschl et al. (2000), as previously used by Zeng et al. (2008), but with rate coefficients of reactions between OH and isoprene nitrates and between NO and isoprene peroxy radicals updated following Paulot et al. (2009a, b). A diurnal cycle is imposed on isoprene emissions as a function of the solar zenith angle. We adopt a set of monoterpene oxidation reactions initiated by OH, O₃, and NO₃, described by Brasseur et al. (1998). Methane mixing ratios are prescribed at the surface with a constant value of 1813 ppbv north of the Equator and a constant value of 1707 ppbv south of the Equator, and are the same for each year. Surface emissions are as specified in Sect. 2.1. Lightning NO emissions are based on the parameterization of Price and Rind (1992, 1994), as a function of convection. The model uses prescribed sea surface conditions following the Atmospheric Model Intercomparison Project (AMIP) II (http://www-pcmdi.llnl.gov). The chemistry is run in a semi-offline mode, i.e. chemistry feedbacks to meteorology and hydrology are deactivated. Table 2 summarizes key model properties for all models.

2.2.2 TM5

TM5 is a global 3-D CTM driven by meteorological fields from the ECMWF ERA-interim re-analysis (Dee et al., 2011) using an update frequency of three hours. Interpolated fields are used for the interstitial time periods. The version used here is identical to that described by Williams et al. (2014) and uses the modified CB05 chemical mechanism (Williams et al., 2013) for describing the chemistry which occurs in the troposphere, along with online photolysis rates. Details relating to the convection, advection and deposition processes employed are given by Huijnen et al. (2010). TM5 includes a full description of HO_x and NO_x

reactions, as well as explicitly treating all C1 to C3 organic species in addition to ammonia (NH₃), sulphur dioxide (SO₂) and dimethyl sulphide (DMS). For this study a horizontal resolution of $3^{\circ} \times 2^{\circ}$ is chosen adopting 34 vertical layers from the surface up to $0.5\,h\text{Pa}$. The isoprene and monoterpene oxidation schemes are based on the mechanisms developed by Yarwood et al. (2005), with modifications to both the oxidation rate of organic hydroperoxide (ROOH) and the production efficiency of HO₂ from the OH initiated oxidation of isoprene following recommendations by Archibald et al. (2010). Also in TM5, a diurnal cycle is applied to the monthly mean isoprene emissions. Methane emissions are included and the simulated surface concentrations are then nudged towards a latitudinally and monthly varying climatology based on surface observations; a detailed description of the approach is given by Williams et al. (2013). Mean surface methane mixing ratios for the year 2004 are listed in Table 2, and their interannual variation is small.

2.2.3 GEOS-Chem

The GEOS-Chem global 3-D CTM (www.geos-chem.org) is driven by meteorology from the NASA Global Monitoring and Assimilation Office (GMAO) Goddard Earth Observing System (GEOS-5) assimilated product (Bey et al., 2001). The native GEOS-5 product with $0.667^{\circ} \times 0.5^{\circ}$ horizontal resolution and 72 vertical levels (surface to $0.01 \, hPa$) is regridded for computational efficiency to $2.5^{\circ} \times 2^{\circ} \times 47$ levels (with all vertical lumping in the stratosphere). Here we use the v9-01-03 coupled $O_3-NO_x-HO_x-VOC$ —aerosol simulation with the Caltech isoprene mechanism (Paulot et al., 2009a, b), which includes 57 transported species (both gas-phase and aerosol) and an additional 49 species that interact chemically but are not transported (The detailed chemical mechanism used in this study can be found at: http://wiki.seas.harvard.edu/geoschem/index.php/New_isoprene_scheme_prelim). As monoterpenes are not included as an explicit chemical species, their emissions are used to produce CO (assumed 20 % molar yield; Hatakeyama et al., 1991) and acetone (assume a 12 % molar yield; Jacob et al., 2002). CH₄ mixing ratios are prescribed annually and latitudinally based on NOAA GMD

surface observations and are listed in Table 2. Interactive chemistry is computed in the troposphere only, with stratosphere production and loss rates for most species taken from the NASA Global Modeling Initiative (GMI) Combo CTM Aura4 model (Murray et al., 2013). Stratospheric ozone is simulated using the Linoz linearized ozone scheme (McLinden et al., 2000). Lightning NO emissions are based on the cloud top height parameterization of Price and Rind (1992) with regional correction to match lightning distributions from satellite, as described by Murray et al. (2012). Other processes in GEOS-Chem v9-01-03 including mixing and deposition are described in detail by Mao et al. (2010, 2013a). The version used here was modified from the standard v9-01-03 to include irreversible HO₂ uptake by aerosols with a gamma value of 0.2 (Mao et al., 2013b), and to include methanol as an interactive tracer based on the offline simulation of Millet et al. (2008). In the standard GEOS-Chem simulation, biogenic emissions are computed online using a coupled version of the MEGAN model. Here, to ensure consistency with the other SHMIP models, we used the pre-computed biogenic emissions described above (Sect. 2.1) and applied an imposed diurnal variability tied to solar zenith angle.

2.2.4 CAM-chem

The Community Atmosphere Model with Chemistry (CAM-chem) is a component of the NCAR Community Earth System Model (CESM). The version used for this study is the same as that used for the Chemistry–Climate Model Initiative (CCMI) (Eyring et al., 2013), and very similar to the CAM4 version described in Tilmes et al. (2015). For SHMIP, CAMchem was run in the specified dynamics mode, using meteorological fields from the Modern-Era Retrospective Analysis For Research And Applications (MERRA) reanalysis product (http://gmao.gsfc.nasa.gov/merra/), regridded to the model horizontal resolution of $1.9^{\circ} \times 2.5^{\circ}$, using the lowest 56 levels. In this study, the internally derived meteorological fields (temperature, winds, surface heat and water fluxes) are nudged at every time step (30 min) by 1 % towards the reanalysis fields (i.e., a 50 h Newtonian relaxation time). The chemical mechanism, based on MOZART-4 (Emmons et al., 2010), includes both tropospheric and stratospheric chemistry, with 150 compounds and 400 photolysis and ki-

netic reactions, and a tropospheric bulk aerosol model (Lamarque et al., 2012). Heterogeneous reactions on aerosols are included as described in Emmons et al. (2010), including the uptake of HO₂ with a reaction probability of 0.2 producing H₂O₂. While the option of running with online MEGANv2.1 biogenic emissions is available in CAM-chem, this was not used and all surface emissions were based on those specified for this intercomparison, with diurnal variation imposed for isoprene and monoterpenes. Methane surface mixing ratios are specified for monthly zonal averages, as used for CCMI, based on RCP6.0 (Meinshausen et al., 2011). Lightning NO emissions are determined according to the cloud height parameterization of Price and Rind (1992) and Price et al. (1997). The vertical distribution follows DeCaria et al. (2006) and the strengths of intra-cloud and cloud-ground strikes are assumed equal, as recommended by Ridley et al. (2005).

2.3 Observations of CO and HCHO in the SH

Long-term measurements of trace gases up to the upper troposphere in the remote SH are sparse. Continuous multi-year tropospheric columns of CO are observed at four SH sites: Darwin, Wollongong, Lauder, and Arrival Heights, with high spectral resolution FTIR spectrometers. In addition, HCHO columns have been retrieved at Wollongong and Lauder. The data records and retrieval methods have largely been presented before (Rinsland et al., 1998, 2002; Jones et al., 2001, 2009; Paton-Walsh et al., 2010; de Laat et al., 2006; Zeng et al., 2012; Morgenstern et al., 2012) and therefore we only give a brief description here.

At Wollongong, Lauder, and Arrival Heights, mid-infrared (MIR) spectra from the FTIR measurements are used to retrieve CO columns, and these stations are part of the Network for the Detection of Atmospheric Composition Change (NDACC; http://www.ndacc.org). The retrieval of trace gas information from these recorded spectra was performed based on the SFIT2 profile retrieval algorithm using the 4.7 µm band, and is similar to that described by Rinsland et al. (1998, 2002). At Arrival Heights, there are no measurements during the polar nights which last 4–5 months per year. CO total columns have been observed at Darwin since 2005 with solar remote sensing using FTIR measurements in the near infrared (NIR),

as part of the Total Column Carbon Observing Network (TCCON) (Deutscher et al., 2010; Wunch et al., 2011). The spectra used for CO retrieval are analysed with the GFIT spectral fitting algorithm (Washenfelder et al., 2006) for total column CO. Details of the retrieval method and data are described in Paton-Walsh et al. (2010). Daily averaged time series of CO columns from 2004 to 2008 are used for comparison with the models. Due to the very small "smoothing" error for CO retrievals, which indicates the difference between realistic and retrieved CO columns, averaging kernels are not applied when comparing with the modelled CO data (de Laat et al., 2006; Zeng et al., 2012). Comparisons are made against daily mean output from each model.

Total columns of HCHO were retrieved at Wollongong and Lauder from the mid-infrared spectra using the SFIT2 inversion algorithm (Jones et al., 2009). HCHO is a very weak absorber in the mid-infrared spectral region. Due to its large "smoothing" errors, the averaging kernels and a priori applied in the retrieval were also applied to the modelled data for a like-with-like comparison between the modelled and retrieved HCHO columns following the method described by Zeng et al. (2012) and references therein.

In order to provide comparisons on a larger spatial scale, we also perform multi-year comparisons for surface CO against flask measurements available from the NOAA Global Monitoring Division network (Novelli et al., 1998). The selected sites are all situated in the SH and cover an extensive latitudinal range. They are typically located away from regions which exhibit strong local emissions of CO. The sites shown are Mahe Island (4.7° S, 55.5° E), Ascension Island (8.0° S, 14.4° W), Pacific Ocean (30.0° S, 176.0° W), Baring Head (41.4° S, 174.9° E), Crozet Island (46.4° S, 51.9° E), Tierra del Fuego (54.9° S, 68.3° W), Syowa Station (69.0° S, 39.6° E) and South Pole (90° S, 24.8° W). **The locations of all sites used in this study are displayed in Fig. 3.**

To illustrate how the models perform on the global scale in general, we also show comparisons between modelled CO and observed CO columns made by the Measurements Of Pollution in the Troposphere (MOPITT) satellite instrument (https://www2.acd.ucar.edu/mopitt). We use the MOPITT version 6 level 3 Thermal-Infrared product, and the data are monthly averages. A description of the data and

the retrieval method is given by Deeter et al. (2003, 2014). Here the daytime monthly mean MOPITT CO columns for January and September 2005 are used for comparison. Model outputs are monthly averaged and have been interpolated to the MOPITT horizontal grid of 1°x 1° and 10 vertical levels with a 100 hPa spacing. The MOPITT CO averaging kernel and a priori data are applied to the calculation of the modelled CO columns, as described by Morgenstern et al. (2012) and the references therein; such an approach is generally recommended when comparing modelled data to data from satellite remote sensing instruments (Rodgers et al., 2003).

3 Comparison between models and observations

3.1 FTIR CO columns

Fig. 4 shows the direct comparison between modelled and FTIR daily mean CO columns for the CLM-MEGANv2.1 simulation. Here, we use the tropospheric FTIR partial columns, for the reasons that not all models have well-resolved stratospheric chemistry in the model, there is a significant contribution of CO from the mesosphere during polar spring (Velazco et al., 2007), and that all models lack or have deficiencies in handling the mesospheric chemistry. FTIR CO partial columns (0-12 km) are used for comparison at Arrival Heights, Lauder, and Wollongong instead of total columns. However, the partial columns of CO at Darwin are not available so we use total columns for comparison. **Note** that the contribution of mesospheric CO to the total column is expected to be minimal at Darwin given its tropical location (Velazco et al., 2007), therefore the differences between the partial and the total columns are expected to be small. The model data at all locations have been interpolated to the dates when the measurements were carried out. Fig. 4 shows that CO seasonal cycles are well reproduced by all four models at all locations. Models accurately reproduce the total columns of CO at Darwin with very small inter-model differences. The Darwin measurement site is the closest to the tropical source regions; this indicates that the emissions in this area are well represented in the models.

Inter-model differences are notably larger at other sites which are located further from the source regions, with consistent overestimation by TM5 and underestimation by CAM-chem at both Arrival Heights and Lauder. Such differences are possibly associated with both differences in the oxidative capacities in these two models and differences in transport (discussed in Sect. 4). All models underestimate CO columns at Wollongong, especially during the peak BB season; this may be due to its proximity to large forested areas and/or the cities of Sydney and Wollongong whose direct emissions may be underestimated in the MACCity inventory. Note that due to the coastal location of Wollongong, model grid boxes may not be representative of the measurement site.

We performed a second set of simulations using LPJ-GUESS isoprene and monoterpene emissions (see Fig. 5); the models visibly underestimate the observed FTIR tropospheric CO columns at all sites. The deviation of model ensemble-mean CO columns from the observed FTIR columns are shown at the four measurement sites (Fig. 6) in comparison with the simulation using CLM-MEGANv2.1 biogenic emissions. The differences between these two simulations are also shown (i.e. CLM-MEGANv2.1 minus LPJ-GUESS). It appears that a larger negative bias exists when adopting the LPJ-GUESS emissions for all of the column measurement stations (i.e. CLM-MEGANv2.1 results in better agreement with the FTIR observations). The deviations of both simulations from the observed CO columns exhibit large seasonal variations but seasonal and inter-annual variations are consistent between these two simulations.

Fig. 7 shows differences between the modelled and observed FTIR CO columns at the four measurement sites from the multi-annual ensemble mean data for both CLM-MEGANv2.1 and LPJ-GUESS simulations. As in Fig. 6, the seasonal variations of the biases from these two sets of simulations follow a very similar pattern, implying that the effect of different biogenic emissions is reflected in the differences in the background CO columns in the SH. The biases shown in the ensemble model means from both simulations are largest during the SH tropical BB season of September, October and November (SON), although at Darwin the negative biases are also high in July and increase from October to December. For Wollongong, Lauder, and Arrival Heights the largest negative biases are in

October, November, and December, respectively; this suggests an underestimation of SH BB sources in GFEDv3 and the subsequent effect on CO columns at SH remote locations through long-range transport. At Darwin, CO columns are more likely influenced by local or nearby BB sources which may have a different seasonality. The **annually** averaged biases of the model ensemble means for each site are shown in Table 3; the lowest biases are at Arrival Heights for both simulations, followed by those for Darwin, Lauder, and Wollongong. Note that the low ensemble bias at Arrival Heights is largely the result of cancellation of a positive bias in TM5 with a negative bias from CAM-chem with a similar magnitude. The large spread between the models indicates that substantial differences exist in other physical and/or chemical processes which are unrelated to emissions.

The individual model biases are also shown in Fig. 7. For both simulations, inter-model variability is notably larger during months that lie outside the seasons when most intensive BB occurs, i.e. typically in austral summer and autumn (covering December and January to May). Such a seasonal dependence of inter-model variability is consistent with that described by Fisher et al. (2014) who compare modelled vertical CO gradients in the Cape Grim region using the same simulations, and is due to the difference in chemistry that controls CO chemical production and loss processes in the seasons other than the peak biomass burning season. Inter-model variability is generally larger in CLM-MEGANv2.1 than in LPJ-GUESS for all seasons and locations, primarily due to the larger response of modelled CO to its higher precursor emissions.

3.2 MOPITT CO columns

Fig. 8 shows MOPITT CO columns and the model biases for January and September 2005. The model data are monthly means convolved with the MOPITT averaging kernels and a-priori data. MOPITT data exhibit a lot of gaps over the Amazon region in January and over Africa in September, due to persistent cloud cover. All models underestimate MOPITT CO columns in the NH (with the exception of East Asia), and overestimate CO in the plumes, particularly over the tropical Atlantic in January and the Pacific in September. These plumes originate in tropical Africa and South Amer-

ica, respectively, indicating an overestimation of biomass burning emissions in these regions. Comparing the models, TM5 shows higher CO columns throughout the SH and CAM-chem the lowest, in agreement with the comparison of FTIR CO columns. Fig. 9 shows the percentage differences between the ensemble-mean modelled and measured columns for both the CLM-MEGANv2.1 and the LPJ-GUESS simulations. There is a general underestimation of CO columns by both ensembles in the NH by up to $\sim -25\%$. Both ensembles overestimate CO in the source regions, with up to +30\% over tropical Africa in January and over Amazonia in September, i.e., during the months of peak biomass burning. In the SH, away from the CO plumes, the CLM-MEGANv2.1 ensemble clearly compares better with MOPITT CO than the LPJ-GUESS ensemble, with biases typically between -10% and +10% in January and September, whereas errors typically are in the range -20% to -5% from the LPJ-GUESS ensemble. Both ensembles also underestimate CO columns over Australia in September, which suggests an underestimation of biomass burning in GFEDv3. This is also reflected in the comparison between modelled and FTIR CO columns in the four SH locations shown above, which generally show negative biases in modelled CO. The two ensembles are fairly similar in the NH with regard to their CO columns but exhibit significant differences in the extratropical SH. This is consistent with a larger relative role of biogenic emissions in the SH versus the NH.

3.3 Surface CO

To assess the models' ability to capture both the seasonality and inter-annual variability of CO at the surface over the simulation period, we show in Fig. 10 comparisons between the CLM-MEGANv2.1 simulations and monthly mean CO values observed at the eight surface sites listed in Sect. 2.3. Consistent with the FTIR column comparisons, all models capture the seasonal cycles of observed surface CO at each location. In line with Fig. 3 in Fisher et al. (2014), TM5 typically exhibits a high bias and CAM-chem exhibits a low bias of the order of 5–10 ppbv. Large variations exist in seasonal cycles at both Mahe Island and Ascension Island, but the timing of the peaks are different. At Mahe Island, sur-

face CO peaks in January and February due to the influence of anthropogenic emissions from India (Wai et al., 2014), whereas at Ascension Island, the seasonal cycle is principally driven by CO which originates from BB in Southern Africa during June-July-August (e.g., Williams et al., 2012; Wai et al., 2014). The inter-annual variability and timing in peak mixing ratios is not captured well at Ascension Island, especially for GEOS-Chem and TM5; this is possibly related to too strong westerly transport out of southern Africa and too weak an oxidative capacity, especially in TM5. For the more southerly, oceanic sites, the seasonal cycles and amplitudes are remarkably similar, indicating that the variability in background CO is rather low at the surface in the SH remote locations. In general, NIWA-UKCA and GEOS-Chem display a better agreement with the observations in the remote SH, indicating that their oxidative capacities are more realistic. The consistent high and low biases in TM5 and CAM-chem, respectively, are related to the oxidizing capacity in these models; this is discussed in Sect. 4.

We quantify the differences between the multi-annual ensemble means for surface CO and the corresponding values derived from the observations for both the CLM-MEGANv2.1 and the LPJ-GUESS simulations (Fig. 11). As was seen in the model comparisons to the FTIR and MOPITT CO data, the observed distributions of surface CO in the SH are better reproduced by CLM-MEGANv2.1 for most of the chosen sites. A comparison of sites shows that the seasonal biases are more variable for the tropical sites which are affected by the inter-annual variability in tropical BB. For the mid- to high-latitude sites, the CLM-MEGANv2.1 ensemble mean accurately reproduces the observations in most cases, whereas the LPJ-GUESS ensemble is consistently biased low. The individual model biases (shown only for the CLM-MEGANv2.1 simulation) are up to $\pm 20\%$, and are much larger than the differences between the two ensemble means ($\sim 10\%$). The generally better agreement between modelled and observed surface CO, relative to the agreement between modelled and FTIR CO columns in the remote SH, reflects that there may be some deficiencies in the models' vertical transport of either CO and/or its precursors. This generally underestimation of observed vertical gradients of CO

by the models in the remote SH was shown by Fisher et al. (2014) for the Cape Grim region.

3.4 HCHO columns

Here we examine the models' ability to reproduce observed HCHO columns at the SH mid-latitude sites Lauder and Wollongong. Fig. 12 shows comparisons between modelled daily mean HCHO columns (from the CLM-MEGANv2.1 simulation) convolved with FTIR a priori data and averaging kernels, and observed daily mean HCHO columns from the FTIR measurements. The seasonal cycles are generally well reproduced across the entire model ensemble, with the seasonal maxima in austral summer and the minima in winter, but all models significantly underestimate observed columns in all seasons. Inter-model differences in modelled HCHO columns are larger at Lauder than at Wollongong; the highest HCHO columns are produced in GEOS-Chem, whereas the lowest are from TM5. Such variations between the models indicate that the differences in the models' chemistry are the driving factor, in particular at the sites that are further away from the emission sources. Significant and persistent low bias across all models cannot be reconciled by considering the diurnal cycle in HCHO; for testing purposes, we also calculated HCHO columns by replacing daily mean HCHO data shown in Fig. 12 with the daily maximum of the 3 hourly data from one of the ensemble members (CAM-chem). This resulted in small overall changes, with ~ 10 –15 % increases that occur in some summer months, and the increases were not sufficient to close the gap between the models and the observations. Therefore we are confident that using daily mean modelled HCHO columns for comparing to columns from FTIR observations that occur during the daylight is satisfactory. Fig. 13 shows the multi-annual monthly mean FTIR HCHO columns and model ensemble means averaged for the same years with both CLM-MEGANv2.1 and LPJ-GUESS emissions for isoprene and monoterpenes. Overall, the models underestimate the observed HCHO columns by approximately 50 %. Differences in biogenic emissions do not appreciably affect this.

In the case of Wollongong, proximity to Sydney and the influence of episodic BB events in the vicinity (Williamson et al., 2013) could introduce local direct and indirect sources of HCHO and chemical precursors which are unaccounted for and might have contributed to the low bias simulated in the models, particularly for the seasonal peaks. However, at Lauder there are no known significant local sources of HCHO. We therefore assume that the underestimation of observed FTIR HCHO columns by the models is very likely related to missing emissions of precursors.

The underestimation of measured HCHO by the models at the remote SH locations had been shown in some previous studies, and in those studies various assumptions about missing processes have been explored (e.g., Ayers et al., 1997; Jones et al., 2009; Vigouroux et al., 2009). Avers et al. (1997) used a box model to simulate the measured surface HCHO at Cape Grim and were unable to capture the magnitude of the observed mixing ratios of HCHO by including a set of standard methane oxidation reactions in the model. Among the major HCHO production channels, Avers et al. (1997) assumed a 100 % yield of CH₃OOH from CH₃O₂ + HO₂. Ayers et al. (1997) then experimented with an alternative oxidation pathway that involved the direct production of HCHO (40%) from CH₃O₂ + HO₂, which resulted in a much improved comparison. We have not applied such high direct yield of HCHO in our models. However, following the recent recommendation of the International Union of Pure and Applied Chemistry (IUPAC) (Atkinson et al., 2006), a 10% direct yield of HCHO has been adopted by NIWA-UKCA but no direct yield has been applied in the other 3 models. The recent IUPAC recommendations (Atkinson et al., 2006) assume a temperature-dependent branching ratio for the direct HCHO production channel (i.e. 0.09 to 0.29 for temperatures ranging from 298 to 218 K). Adopting this recommendation, an additional test was performed in TM5, showing some modest increases in HCHO in the extratropics of up to ~ 10 %. However this is not sufficient to explain the large bias shown here. Another hypothesis suggested by Ayers et al. (1997) is the possibility of a small marine biological source of isoprene (e.g., Bonsang et al., 1992). Recently, Lawson et al. (2015) found relatively abundant HCHO precursors (dicarbonyls) in two regions of the southwest Pacific, corroborating the hypothesis that marine biological activity might be responsible for the measured HCHO abundance. However, spatial sampling and understanding of the underlying biological processes remain poor.

The HCHO column dataset we use here is an extension of the 1992-2005 dataset described by Jones et al. (2009), retrieved using the same algorithm. They also derived HCHO mixing ratios at a coarse vertical resolution. Jones et al. (2009) performed a box model simulation based on subsets of the Master Chemical Mechanism (MCM) (Saunders et al., 2003) including the isoprene oxidation scheme of the MCM. They found that high HCHO mixing ratios retrieved at Lauder cannot be explained by methane oxidation alone and that additional local sources, possibly isoprene, are needed to explain the observed near-surface HCHO mixing ratios at Lauder. A recent study by Vigouroux et al. (2009) compared modelled and observed FTIR HCHO columns at Reunion Island, using the global chemical transport model IMAGESv2; they also underestimate the observed FTIR HCHO columns albeit with a smaller magnitude than that shown in this study. The time series shown by Vigouroux et al. (2009) are for August to November 2004 and for May to November 2007, respectively, and the differences between modelled and observed HCHO columns are around 30 and 25 %, respectively. The isoprene mechanism used by Vigouroux et al. (2009) is based on the MCM and is described by Müller and Stavrakou (2005). The isoprene emissions used by (Vigouroux et al., 2009) are from the MEGAN-ECMWF inventory (Müller et al., 2008), and the yearly totals averaged over 2004 to 2006 are around 10% lower than the CLM-MEGANv2.1 inventory used here. To investigate the possible causes for the low bias in modelled HCHO, Vigouroux et al. (2009) include methane oxidation by tropospheric chlorine, but the impact of this process on HCHO columns is only about 1-2% and therefore cannot explain the underestimation. They also experimented using a different OH climatology; this increase of OH abundance results in better agreement between observed and modelled HCHO columns but cannot fully reconcile the substantial differences, and a more probable explanation is an underestimation of HCHO precursors transported from Madagascar to Réunion Island. This finding, together with our finding here, suggests that the underestimation of HCHO columns is persistent throughout the SH. Observations of HCHO in the remote SH regions are extremely sparse, and it is impossible to fully constrain modelled HCHO. Note that in both studies (Jones et al. (2009), Vigouroux et al. (2009), FTIR HCHO columns compare well with satellite measurements, and with both satellite and MAX-DOAS measurements, respectively. This again suggests that the FTIR HCHO retrieval is robust at all sites, and that the likely cause for model-observation differences is missing sources of HCHO and/or its precursors in the models.

4 Model differences in chemistry and transport

Although the four models are constrained by the same emissions, there are significant differences in the models' abilities to reproduce observed CO columns and surface CO in the remote SH, as shown above. Here we explore the underlying factors driving these differences. To diagnose the extent of differences in transport between the models, we examine the two passive CO tracers defined in Sect. 2: one with a fixed lifetime of 25 days (referred to as $\rm CO_{25}$) and the second with first-order loss via model calculated OH (referred to as $\rm CO_{OH}$). Both tracers are subjected to the same surface emissions as the full simulations, but not subjected to any secondary production of CO from methane and NMVOC oxidation. Dry deposition of CO is not included for either of the additional CO tracers as it is considered a minor loss channel for the SH.

The global tropospheric CO columns from all models for January and September are shown in Fig. 14. January and September represent the seasonal maxima of biogenic and biomass burning emissions in the SH, respectively. Here, we define the tropospheric columns as the columns below the chemical tropopause marked by the 150 ppbv O₃ isopleth in each model (monthly mean O₃ used here is averaged over 2004–2008). Although here we focus on the SH, we note that the inter-model differences apparent in the SH are consistent with those occurring in the NH, namely, the lowest CO columns occur in CAM-chem, followed by NIWA-UKCA, with higher CO columns from GEOS-Chem and TM5

for both hemispheres, indicating systematic differences between the models. Comparing the seasonal variations, CO columns are generally higher in September than in January in the SH, primarily due to the timing of the most intensive tropical BB events in austral spring. Of the four models, CAM-chem simulates the lowest CO columns in both the source regions and in the remote mid- to high latitudes. Examining the distributions of the tropospheric columns of CO₂₅ shown in Fig. 15, CO₂₅ exhibits similar distributions among the models for both seasons in source regions as those shown in Fig. 14. The differences become more obvious in the extratropics, with NIWA-UKCA showing slightly weaker transport towards the poles, whereas GEOS-Chem shows somewhat stronger export of CO₂₅ out of the source regions and towards the poles. Overall, despite some differences, the magnitude and distribution of CO₂₅ are very similar among the models. However, such differences and similarities in transport among the models are not reflected in the differences in CO columns shown in Fig. 14 in which TM5 simulates the highest CO columns and CAM-chem the lowest in both the source and remote regions. CO_{OH} shown in Fig. 16 is a more realistic proxy of CO, reflecting to the influences of the models' variable OH concentrations. Like CO₂₅, the magnitudes and distributions of CO_{OH} are similar across the models, hence the main driver of the model differences in total CO cannot be attributed to primary emissions.

Next, we quantify the roles of transport and chemistry in determining the inter-model variability in CO columns in the SH. We examine three zonal bands defined as 0–30° S, 30–60° S, and 60–90° S. These latitude bands capture the main tropical source regions and mid- and high latitudes, respectively. Fig. 17 shows the monthly mean tropospheric columns of CO as well as ratios of $\rm CO_{25}/\rm CO$ and $\rm CO_{OH}/\rm CO$ columns, averaged across each of these zones for each model. $\rm CO_{sec} = \rm CO - \rm CO_{OH}$ is an estimate of the fraction of CO that is produced by oxidation of $\rm CH_4$ and NMVOCs; the ratio of $\rm CO_{sec}$ to CO is also shown in Fig. 17. These ratios define the contributions of $\rm CO_{25}$, $\rm CO_{OH}$, and $\rm CO_{sec}$ to the total CO columns in each model. Fig. 17 shows that CO columns decrease towards the high latitudes and the seasonal maxima are during the September/October BB season in all zones. Although $\rm CO_{25}$ is an idealized tracer designed to diagnose differences in the long-range transport

simulated in each model, CO_{OH} should be a more realistic measure of how much primary emissions of CO contribute to the CO columns because CO_{OH} reflects the locally varying lifetime of CO due to the spatial variability of OH. The ratio of CO₂₅ to CO drops sharply from the tropics to the pole for all models (from \sim 20 to \sim 5% in the annual average), as would be expected from the hemispheric distribution of emissions and the timescales for meridional transport. By contrast CO_{OH}/CO reduces only from \sim 30 to \sim 25 % in the yearly average. This reflects that the lifetime of CO is considerably longer outside of the source region due to lower background O₃ levels (and therefore lower OH levels) in the more pristine environment away from strong NO_x sources. CO_{sec}/CO shows a moderate increase from 70 to 75% from the tropical zone to the high latitudes. Overall, primarily-emitted CO makes up ~ 25 –45% of total tropospheric CO in the source region and ~ 20 –40% in the polar region, depending on season, while the secondary CO makes up the remainder of the tropospheric columns, i.e. $\sim 55-75$ % in the source region and $\sim 60-80$ % in the polar region. Regarding seasonal variability, CO₂₅ and CO_{OH} have proportionally larger contributions in austral spring when BB dominates the CO emissions, whereas CO_{sec} shows larger contributions in austral summer/autumn. Of all the models, NIWA-UKCA shows the smallest contribution from primary CO to the columns and the largest contribution from the secondary CO, relative to the other three models.

Inter-model differences in CO columns and the additional CO tracers are expressed as the ratio of individual model columns vs. the multi-model mean columns for each zone, shown in Fig. 18. For CO columns, the inter-model differences are smallest in the tropical zone and gradually increase towards the pole, with the highest CO columns from TM5 and the lowest from CAM-chem, in **agreement** with the FTIR comparisons and the surface comparisons shown earlier. Examining the inter-model differences in CO₂₅, the model spread increases substantially towards the polar zone, and is characterized by the strongest transport out of the source region from GEOS-Chem and the weakest from NIWA-UKCA (also shown in Fig. 15). Note that this behaviour is not reflected in the model spread of CO columns (i.e. the highest CO occurs in TM5 and the lowest in CAM-chem). By contrast, the patterns of model spread in CO_{sec} and to a lesser degree in CO_{OH} are consistent

with that seen in the CO columns, indicating that the inter-model differences in modelled CO columns are strongly influenced by the differences in CO_{sec.}, which is dependent on the oxidizing capacity in the model that also drives the loss of primary-emitted CO by OH. Considering also the absolute contributions of both primary CO sources and secondary CO production to the SH CO columns (these being ~ 35 and ~ 65 %, respectively), we can **deduce** that inter-model differences in CO columns are attributed about 1/3 to primary and 2/3 to the secondary CO production in the SH. Note that here we only take into account the accumulated effects of primary and secondary contributions to CO; we do not differentiate or individually identify the separate influences e.g. of transport and chemistry. For example, the large CO columns in TM5 can be the result of combined effects of slower chemical loss of CO due to lower OH levels in the model and a faster secondary CO production in the source region, as reflected in higher ratios of CO_{sec} to CO shown in Fig. 18. In contrast, GEOS-Chem CO has faster loss by OH than TM5 (but slower than the others), but this is outweighed by a stronger transport resulting in higher CO compared to that in NIWA-UKCA and in CAM-chem. For CAM-chem, moderately slow transport of CO out of the source region combined with slower secondary CO production result in the lowest CO columns. More quantitative analyses of differences in chemistry are carried out in Section 5.

To further probe the differences between the models, we show vertical profiles of modelled key species, namely CO, HCHO, O_3 , and OH mixing ratios We display data for January 2005 from the CLM-MEGANv2.1 simulation, because in austral summer the chemical production maximizes due to stronger photochemistry and higher biogenic emissions, and absolute inter-model differences in CO columns are also larger than in other seasons. TM5 is characterized by consistently high CO throughout the SH. The CO values in NIWA-UKCA and in GEOS-Chem are very close in all three zones, exhibiting differences of $\sim 5\text{--}10\,\%$, although CO in NIWA-UKCA is slightly higher than that in GEOS-Chem in the tropics but becomes lower towards remote regions. This may reflect slower meridional transport in NIWA-UKCA (shown in CO $_{25}$) combined with larger chemical production in the source region. The HCHO mixing ratios decrease sharply with altitude due to the dominant chemical precursors residing in the boundary layer and the efficient

photo-dissociation, but the vertical gradient becomes smaller away from the source region, particularly in TM5, due to depletion of the biogenic precursor emissions in the remote SH. HCHO abundances in the four models correlate with OH to some extent, i.e. both OH and HCHO are relatively large in GEOS-Chem, whereas both are relatively small in TM5; this reflects the approximate linearity between the modelled HCHO relationship between HCHO and OH; OH is involved in both the loss and the production of HCHO, and abundance and methane oxidation via OH in the remote SH. However, there is no simple linear HCHO is one of the OH sources. The modelled OH profiles do not seem to be closely related to O_3 (the primary source of OH) in that TM5 has the lowest OH but its O_3 values lie in the middle of the model range; this is likely due to differences in photolysis schemes. Water vapour fields are very similar among the models.

5 Analysis of chemical production and loss rates

To quantify the effects of differences in model chemistry, we analyse chemical production (CP) and loss (CL) rates of CO and of HCHO, as listed in Tables 4 and 5 for both simulations (i.e. with CLM-MEGANv2.1 and LPJ-GUESS emissions, respectively). The budget terms displayed are for year 2004 and for the whole globe, the SH, and the three SH latitudal bands defined above. The corresponding burdens of CO, HCHO, and OH are shown in Table 6. We define the tropopause of each model as the 150 ppbv O_3 isopleth in each model, as in Sect. 4.

Examining the CO budget terms shown in Table 4, the SH CPs and CLs of CO are under half of the global values. The main contribution to the SH CPs come from the 0-30S latitude band; production decreases sharply towards the southern polar region. In general, chemical production and loss rates of CO are larger in the CLM-MEGANv2.1 simulations, indicating larger biogenic emissions leading to larger CO production. However, the CO production from methane oxidation is generally larger in LPJ-GUESS as a result of increased OH (shown in Table 6) due to the reduction of biogenic emissions. In all models, the ratio of CP of CO to surface emissions is

markedly larger in the SH than the global values; this results in larger intermodel differences in the SH due to the differences in the models' underlying chemistry. NIWA-UKCA shows the largest total CPs for all domains, followed by GEOS-Chem, TM5, and CAM-chem, and the differences in total CPs are dominated by the differences in the oxidation of NMVOCs. Methane oxidation is more constrained among the models, and the differences in methane oxidation are mainly driven by differences in OH (shown in Table 6). Note that we do not calculate CO production rates from NMVOC oxidation explicitly in the models; instead they are deduced from the total CP and methane oxidation terms, assuming a 100% yield of CO from methane oxidation; this is only for diagnostic purposes and we do not make such assumptions in the actual mechanisms. Dry deposition of CO is a small loss term, particularly for the SH. TM5 and NIWA-UKCA have comparable dry deposition loss rates. Note that CO loss through dry deposition is not included in GEOS-Chem and is not provided for CAM-chem.

The seasonal variation of CP and CL of CO in the SH are shown in Fig. 20. The surface emission of CO peaks in September in the SH which is dominated by biomass burning emissions. CPs and CLs maximise in austral summer and minimise in austral winter. CAM-chem shows much lower total CP and NMVOC oxidation compared to the other three models, in particular during the summer months, indicating below-average oxidizing capacity in that model. Examining the contribution to the total CPs, methane oxidation and oxidation from NMVOCs are nearly equal in all models except CAM-chem where NMVOC oxidation is significantly lower than methane oxidation. Methane oxidation is largest in GEOS-chem, followed by NIWA-UKCA reflecting the higher OH in these two models. The peak chemical loss shown in all four models in October is in response to the peak of surface emissions of CO. We also display the ratio of CO production from NMVOCs to the total CP, showing that TM5 has the highest ratio, indicating a fast conversion of NMVOCs to CO in TM5. In comparison, CAM-chem has a substantially lower NMVOC oxidation to total CP ratio, indicating a slower NMVOC to CO conversion; this is the primary cause for low CO in

this model. We have not explicitly quantified the CO production from isoprene oxidation but assume that isoprene oxidation is the dominant contributor to NMVOC oxidation in the SH (Pfister et al., 2008). We therefore suggest that the different isoprene oxidation schemes used in the models are responsible for the differences in the chemical production rates of CO. Without a detailed comparison of the chemical mechanisms used in the models, we cannot identify which processes and/or parameters that make up the mechanisms are responsible for the differences in the models employed here, and such tests would be more suitably done in a box model in which parameters can be more straightforwardly controlled (e.g., Archibald et al., 2010). Four different isoprene oxidation mechanisms are included in the models presented here. They vary in complexity and also in the approaches to treat degradation products. The isoprene oxidation mechanism in NIWA-UKCA is based on a smaller mechanism (MIM; Pöschl et al. (2000)) than those used in GEOS-Chem (Paulot et al., 2009a, b) and in CAM-chem (Emmons et al., 2010). NIWA-UKCA contains some recently updated rate coefficients of reactions between NO and peroxy radicals from the OH-initiated isoprene oxidation reactions, and reactions between OH and isoprene nitrate (Paulot et al., 2009a, b). The isoprene oxidation scheme in TM5 is based on the CB05 chemical mechanism (Yarwood et al., 2005) with modifications made to both the oxidation rates of peroxides and the production efficiency of HO₂ from the OH-initiated oxidation of isoprene based on recommendations by Archibald et al. (2010). Our results here show that the rates of NMVOC oxidation are substantially faster in TM5 (shown in Table 4 and Fig. 20) than in the other models, and such faster NMVOCs oxidation rates are largely driven by the isoprene oxidation scheme in that model, which, together with the lower OH (shown in Table 6), lead to higher CO than in the other models.

HCHO budget terms from NIWA-UKCA and TM5 are listed in Table 5 (These terms were not saved in the other models). The surface emissions of HCHO are small compared to the in-situ chemical production and loss terms. The global total CP in NIWA-UKCA is slightly larger than in TM5 for both simulations, but the amounts are compa-

rable for the SH. Methane oxidation rates are higher in NIWA-UKCA for all regions due to the higher OH in that model, and NMVOC oxidation rates are significantly larger in TM5. Examining the chemical losses, HCHO loss through the reaction with OH is much higher in NIWA-UKCA; however, HCHO losses through photolysis are comparable between these two models. Together with the smaller burden of HCHO in TM5, this implies that HCHO photolysis rates are larger in TM5 than in NIWA-UKCA. (This diagnostic is not directly available for either model.) The much larger wet deposition of HCHO in TM5 (i.e., $\sim 10\%$ of the total loss terms), compared to that in NIWA-UKCA ($\sim 3\%$), could explain the lower HCHO burden/columns in TM5. A additional hydration of HCHO is applied in TM5 (but not in the other models), which further enhances the effective solubility of HCHO in aqueous solution (Huijnen et al., 2010). This may have resulted in an additional loss of HCHO to wet deposition in TM5 which is however still substantially smaller than the gas phase loss processes. The Henry's Law coefficients, governing gas- and liquid-phase partitioning of HCHO, applied in the other models are comparable.

6 Sensitivity of modelled SH CO and HCHO to uncertainties in biogenic emissions

In Sect. 3, we showed the model deviations in CO and HCHO columns from observed FTIR values at four remote SH sites using two different biogenic emissions inventories (for isoprene and monoterpenes), and found that modelled CO columns with LPJ-GUESS biogenic emissions are consistently lower and less representative of observed values than those produced using CLM-MEGANv2.1 emissions (Table 3). Here we further quantify the changes in CO and HCHO columns in response to changes in biogenic emissions at the hemispheric scale, and also highlight associated changes in the corresponding OH columns. Figures 21–23 display the monthly mean global distributions of relative differences in CO, HCHO, and OH columns between simulations using CLM-MEGANv2.1 and LPJ-GUESS, respectively, for January and July (averaged over 2004–2008). The differences calculated for all species are expressed as the percentage change relative to the CLM-MEGANv2.1

simulation. Here, we show results for the January and July months in order to contrast the seasonal features in oxidizing capacity. For all models, applying LPJ-GUESS emissions results in significant decreases in CO columns throughout the SH, with the largest decreases in the South American and Australian source regions (Fig. 21), in response to the smaller emission fluxes of isoprene and monoterpenes from LPJ-GUESS (the accumulated peak isoprene emissions in CLM-MEGANv2.1 are 25% higher than in LPJ-GUESS during the peak season of the austral summer months, shown in Fig. 1, and the biggest differences are in South America). Away from these source regions, the differences are largely homogeneous in the mid- to high latitudes. The models' responses to changes in biogenic emissions vary considerably, with TM5 having the largest sensitivity of CO columns change to changes in biogenic emissions, namely $\sim 35\,\%$ in January and $\sim 25\,\%$ in July in the source regions and 10–15% over the remote SH. GEOS-Chem has the lowest sensitivity with 15–20% changes in January and 10–15% in July in the source regions and less than 10% in remote regions in response to the same emission changes of isoprene and monoterpenes.

For corresponding changes in tropospheric HCHO columns (Fig. 22), substantial decreases (up to ~ 50 –60%) occur in the source regions of South America and Australia in response to smaller emission fluxes in LPJ-GUESS, relative to CLM-MEGANv2.1. These reductions in HCHO columns propagate to the sub-tropical remote oceans where the magnitude of the decreases is greatly reduced. There are some increases in HCHO columns over southern Africa, which are responses to the higher isoprene emissions in LPJ-GUESS. However, there is a consistent increase of up to 5% over large areas of the midto high latitudes which is apparently not directly caused by reduced biogenic emissions. We find that changes in both CO and HCHO are associated with changes in OH (Fig. 23); the tropospheric OH columns exhibit substantial increases in the source regions as a result of reduced isoprene and monoterpene emissions; qualitatively these effects follow the differences in the geographical distributions of the emissions, and are of opposite sign to both the CO and the HCHO columns changes there. OH increases in remote regions are largely positive, and are opposite in sign to the CO changes; i.e. reduced loss rates of CO

cause increases in OH. However, increases in OH columns away from the source regions correlate with HCHO changes; this implies that increases in HCHO in remote regions under LPJ-GUESS emissions are due to strengthened methane oxidation through increases in OH. The inter-model differences in HCHO changes are generally small in remote regions; TM5 shows the largest sensitivity over the source regions in both OH and HCHO, due primarily to the faster isoprene oxidation processes in that model. Note that the large relative differences in both HCHO and OH in July at high latitudes shown in CAM-chem are not significant because the background abundances of both species in the polar region are extremely small.

In table 7 we summarise hemispheric changes in chemical production and loss rates of tropospheric CO and HCHO, in response to the differences in biogenic emissions. Values expressed are percentage changes (i.e. LPJ-GUESS minus CLM-MEGANv2.1 relative to CLM-MEGANv2.1), and are given for both hemispheres to assess the hemispheric impact of biogenic emissions. In the SH, the changes in all terms are negative, except for the rates of chemical production of both CO and HCHO from methane oxidation; this is generally the result of increased OH in the LPJ-GUESS simulation, in response to reduced biogenic emissions in that inventory. For all models, relative reductions in NMVOC oxidation rates (-17.8 to -35.6%) are substantially larger than relative increases in CP from methane oxidation (3.2 to 8.0%), in response to changes in biogenic emissions. Therefore, NMVOC oxidation (mainly of isoprene) is the driving factor for model differences in in-situ CO and HCHO production. The burden changes are closely related to the changes in total CP, i.e. TM5 has the largest changes in both burden and the CP, and GEOS-Chem has the smallest terms for both. For all models, relative responses in the SH are much larger than in the NH, emphasizing the importance of biogenic emissions for CO and HCHO formation in the SH.

Complementing the comparison of columns, we here compare the seasonal differences in vertical profiles of CO mixing ratios between CLM-MEGANv2.1 and LPJ-GUESS simulations, averaged zonally and over 2004 to 2008. Fig. 24 shows large reductions in CO over

the SH tropics in all simulations using LPJ-GUESS emissions of isoprene and monoterpenes, relative to those using CLM-MEGANv2.1, and these reductions propagate to the upper SH tropical troposphere and spread throughout the middle and high latitudes. This shows that the CO column changes in the extratropics are dominated by the changes in the free and upper troposphere, where CO has a relatively long lifetime. Overall, the impact of biogenic emissions on CO are more significant in the SH than the NH. In the SH, throughout the depth of the troposphere, the LPJ-GUESS simulations have reduced CO, which is linked to much reduced CO in the tropics. This effect maximizes during austral winter and spring.

7 Summary and conclusions

We have compared modelled daily-mean CO and HCHO columns from a four-model ensemble with the observed daily-mean FTIR columns of these two species at SH sites including the tropical site Darwin, the mid-latitude sites Wollongong and Lauder, and the Antarctic site Arrival Heights for CO, and Wollongong and Lauder for HCHO. We use CLM-MEGANv2.1 biogenic emissions for the first set of simulations; for these simulations modelled and measured CO are in reasonable agreement, albeit with some low biases, in all models at most locations; annually averaged deviations relative to the observations are -3.2% at Arrival Heights, -8.6% at Lauder, -19.2% at Wollongong, and -6.9% at Darwin for the 4-model mean. The largest discrepancies between modelled and observed CO columns occur at Wollongong which is heavily influenced by local urban and industrial sources and episodic nearby bush fires that are most likely unaccounted for in the emission inventories. Large inter-model differences exist at all locations for all seasons with the exception of austral spring at Darwin where the local biomass burning sources dominate the CO columns. We also compare the modelled surface CO to observations; significant inter-model differences exist although the ensemble mean exhibits good agreement with the observed values for most sites. The inter-model differences for modelled surface CO are markedly larger than the differences between the ensemble mean and observed surface CO. In agreement with previous modelling studies of HCHO in the remote SH (?Jones et al., 2009; Vigouroux et al., 2009), the models significantly underestimate observed HCHO columns at Wollongong and Lauder by more than a factor of 2, and the largest discrepancy occurs during austral summer. We cannot reconcile such significant differences between the modelled and observed HCHO columns over the remote SH with our current understanding. We hypothesize that missing local sources and/or missing chemical processes are the most likely causes. The fact that model differences are much smaller than the differences between the models and the observations indicate that the cause of such a large discrepancy probably goes beyond what the differences in chemical mechanisms can explain.

To determine the sensitivity of CO and HCHO distributions to biogenic emissions, we perform a second set of simulations with emissions of isoprene and monoterpenes from the LPJ-GUESS dataset; results show that the LPJ-GUESS simulations exhibit systematically lower CO columns and lower surface CO than the CLM-MEGANv2.1 simulations, in response to an average of \sim 9% reduction in isoprene emissions globally and a \sim 17% reduction in the SH (monoterpene emissions are also substantially lower in LPJ-GUESS; see Table 1). Annually averaged relative differences between ensemble model mean and observed CO columns are -10.5% at Arrival Heights, -17.1% at Lauder, -27.5% at Wollongong, and -19.9% at Darwin. The differences in surface CO at remote monitoring sites between the two simulations are generally smaller than 5%. At neither Wollongong nor Lauder do we find that differences in biogenic emissions have any significant impact on modelled HCHO columns.

Examining the response of CO and HCHO columns to differences in biogenic emissions of isoprene and monoterpenes on the hemispheric scale, we show that both species exhibit large sensitivity to emissions in the source regions, with 30–40 % **reductions in** CO and HCHO columns, as a direct consequence of the mainly reduced emissions of isoprene and monoterpenes in the LPJ-GUESS inventory, relative to CLM-MEGAN v2.1 (i.e. \sim 37 % **reduction of isoprene emissions in Australia and Indonesia**, \sim 23 % **reduction in South America**, and \sim 13 % **overall increase in Africa**, with both increases and reductions

occurring in different regions), and these reductions in CO and HCHO are generally larger in summer than in winter. Away from the source regions and throughout the SH, decreases in CO columns are roughly half those occurring in the source regions, whereas there are moderate increases in HCHO columns ($\sim 5\,\%$) despite the significant decreases in and near the source regions for all models. We show that the increases in HCHO columns in the remote SH for LPJ-GUESS, relative to CLM-MEGANv2.1, are linked to the increases in OH columns through enhanced methane oxidation in the remote SH (see Tables 5 and 6). There are substantial increases in OH columns in the source regions in direct response to the reduced isoprene and monoterpene emissions in the LPJ-GUESS inventory, whereas the general increase (up to $\sim 5\,\%$ across the models) in the remote regions is the result of reductions in CO and possibly other longer-lived isoprene oxidation products.

Significant inter-model differences exist in modelled CO columns; we quantify these differences in three latitudinal regions (SH tropics, mid-, and high latitudes). The ratios of individual model columns to the ensemble mean columns (annually averaged and averaged across the three regions) are between 0.85 and 1.15 for the tropical region, and the range increases to between 0.7 and 1.2 at high latitudes. Using diagnostic tracers, we assess the impact of modelled transport (by CO₂₅), the contribution from primarilyemitted CO (by CO_{OH}), and CO produced and transported from secondary CO production $(CO_{sec} = CO - CO_{OH})$. The results reveal that the differences in transport are not sufficient to explain the differences in modelled CO columns. The modelled range of CO_{OH} corresponds much better to the modelled CO columns than CO₂₅ but still cannot fully explain the inter-model differences in modelled CO columns. The differences in secondary CO production, i.e. CO_{sec}, however, correspond well with those in modelled CO columns. TM5 exhibits the highest values in both variables, followed by GEOS-Chem, NIWA-UKCA, and CAMchem in magnitude. We calculate that CO_{sec} contributes around 65% to CO in the tropics and around 75% in the polar region in each model, and is responsible for two thirds of the inter-model differences in modelled CO columns overall. This suggests that the models' differences in secondary CO production from methane and NMVOCs oxidation play a major role in their ability to reproduce the CO columns in the SH, as also noted by Fisher et al. (2014).

We further quantify the models' differences in chemistry by examining the chemical production and loss terms of CO and HCHO in the models. Results show that large differences in chemical production between the models are largely attributed to differences in the rates of NMVOC oxidation, which are mainly driven by the differences in isoprene oxidation processes, which exhibit varying degrees of complexity in the models. We show the collective effects that different isoprene oxidation schemes have on the rates of chemical production of CO and HCHO but are not able to individually quantify which reactions/processes are responsible for the differences in modelled CO and HCHO. Among the four models, NIWA-UKCA has the highest total chemical production rates of CO, followed by GEOS-Chem, TM5, and CAM-chem which has the lowest chemical production rates. Methane oxidation rates are mainly driven by the OH abundance in the models with TM5 having the lowest OH hence the lowest methane oxidation rates. The fastest conversion rates from NMVOCs to CO occurs in TM5, and the slowest in CAM-chem, leading to respectively high and low CO in these two models. Modelled CO in NIWA-UKCA and GEOS-Chem both better matches the observations in general, irrespective of the different complexities of the isoprene oxidation schemes employed in these two models. Moreover, GEOS-Chem includes some recent advances in isoprene oxidation mechanisms, for example, OH formation of epoxide species which regenerate OH under low NO_x conditions (Paulot et al., 2009b). Epoxides are not included in other models. We have not specifically tested how recent experimental evidence on isoprene oxidation mechanisms, e.g., OH regeneration in a low-NO_x environment (Fuchs et al., 2013), will impact on modelled species. More detailed and targeted studies will be needed to clarify how individual approaches/processes making up isoprene oxidation schemes will impact chemical production of CO and HCHO in global models.

Production and loss terms of HCHO are assessed in NIWA-UKCA and TM5. We find that total chemical productions are comparable in the two models, with moderately

larger chemical production and loss rates in NIWA-UKCA. Again, the production of HCHO from the oxidation of NMVOCs is faster in TM5 although this is partly offset, for HCHO production, by the slower methane oxidation rates due to lower OH. The markedly lower HCHO in TM5 than in NIWA-UKCA could be due to the substantially larger wet deposition loss rate of HCHO, and a faster photo-dissociation rate of HCHO in TM5. Despite the differences in rates of HCHO formation and loss, we cannot, based on these differences alone, explain the substantial low bias in modelled HCHO in all models compared to the observed HCHO columns at Lauder and Wollongong. We therefore suspect that missing local sources and/or HCHO precursors might contribute to the differences between modelled and observed HCHO.

We conclude that the uncertainty in biogenic emissions remains a significant problem in modelling both long- and short-lived species throughout the SH. Understanding the differences between isoprene oxidation mechanisms and the resultant differences in modelled CO and HCHO is critical, and might result in an improvement in these mechanisms, allowing for a more robust use of HCHO and CO columns to constrain biogenic emissions and reduce this uncertainty. Given that the differences between the two biogenic emissions inventories used here are moderate compared to the much larger uncertainties existing in the current estimates of isoprene and monoterpene emissions, the resultant uncertainty in modelled CO could be much larger. Although the ensemble model mean satisfactorily compares to observed CO in the SH, the large inter-model differences add more uncertainties in modelled CO and in constraining biogenic emissions. Note that in this paper, we do not separately quantify the effect from changes in monoterpene emissions. The emissions from monoterpenes are around 30 and 10% of those of isoprenes in CLM-MEGANv2.1 and LPJ-GUESS inventories, respectively, which could have a significant impact on modelled CO. However, due to the large uncertainty in emissions and the varying degrees of complexity of the monoterpene degradation schemes included in each model, this will further complicate the interpretation of the impact from changing monoterpene emissions.

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Table 1. Emissions $(Tg yr^{-1})$ and initial conditions. Values in brackets are sums of SH emissions.

	2004	2005	2006	2007	2008
Total surface NO	96	98	99	101	99
Lightning NO*	7.1-13.3	7.5-13.9	7.5-13.7	7.3-13.9	7.7-13.9
Total Surface CO	1010	1037	1072	1037	985
Isoprene (M)	508 (269)	463 (246)	462 (243)	481 (254)	494 (265)
Isoprene (G)	442 (212)	450 (220)	433 (205)	439 (210)	431 (207)
Monoterpenes (M)	143 (72)	132 (66)	132 (68)	138 (70)	136 (68)
Monoterpenes (G)	35 (11)	36 (12)	34 (11)	35 (11)	34 (11)

M denotes CLM-MEGANv2.1 emissions; **G** denotes LPJ-GUESS emissions; * Individual model values for the year 2004 are: 13.3 (NIWA-UKCA), 12.9 (TM5), 12.4 (GEOS-Chem), and 7.1 (CAM-chem).

Table 2. Summary of model information.

	NIWA-UKCA	TM5	GEOS-Chem*	CAM-chem
Resolution lon/lat/lev	3.75°/2.5°/60	3.0°/2.0°/34	2.5°/2.0°/47	2.5°/1.9°/56
Meteorology	Driven by observed SSTs and sea ice	ERA-Interim	GEOS-5	Specified dynamics, MERRA reanalysis
Mean surface CH ₄ * [ppbv]	Global 1758, SH 1709	Global 1794, SH 1739	Global 1782, SH 1731	Global 1758, SH 1709
Chemistry	Tropospheric and stratospheric chemistry; 85 species (Morgenstern et al. 2013)	Modified CB05 chemical mechanism, 60 species , (Williams et al., 2013)	tropospheric chemistry, 121 speciess, 106 transported	150 species, MOZART scheme (Emmons et al., 2010)
Isoprene oxidation mechanism	Mainz Isoprene Mechanism (Pöschl et al., 2000) with update OH and NO initiation rates (Paulot et al., 2009a, b)	CB05 (Yarwood et al., 2005); Modified HO ₂ yields (Archibald et al., 2010)	Caltech Isoprene Mechanism (Paulot et al., 2009a, b)	MOZART scheme (Emmons et al., 2010)

^{*} GEOS-Chem version v9-01-03 is used in this study. ** Surface CH₄ mixing ratios shown here are for year 2004; NIWA-UKCA uses the same values for each year and the interannual variation is small in other models.

Table 3. Multi-annual averaged ensemble model mean deviations (%) from observed FTIR CO columns.

	CLM-MEGANv2.1	LPJ-GUESS
Arrival Heights	-3.2 %	-10.5%
Lauder	-8.6%	-17.1 %
Wollongong	-19.2%	-27.5%
Darwin	-6.9%	-19.9%

Table 4. Tropospheric CO budget for 2004. Units in $TgCO yr^{-1}$

	Global		S	Н	0-30S		30-60S		60-90S	
	M	G	M	G	М	G	М	G	М	G
			NIWA-U	JKCA						
Surface emission	1010	1010	306	306	288	288	18	18	0.3	0.0
Total CP	1887	1786	821	742	718	631	97	103	5.8	6.
CH ₄ oxidation*	1067	1086	437	451	353	367	78	76	5.0	5.
NMVOC oxidation**	820	700	384	291	365	264	19	27	8.0	1.
Chemical loss	2790	2668	1057	986	871	814	170	157	15	1
Dry deposition	101	98	27	25	25	23	2.0	1.6	0.0	0.
			TM	15						
Surface emission	1010	1010	306	306	288	288	18	18	0.3	0.
Total CP	1650	1535	743	663	654	565	86	94	3.9	4.
CH ₄ oxidation	865	927	361	390	284	312	73	74	3.7	3.
NMVOC oxidation	785	618	382	273	370	253	13	20	0.2	0.
Chemical loss	2516	2410	1016	941	801	748	201	180	14	1
Dry deposition	115	107	36	32	33	29	2.3	2.1	0.5	0.
			GEOS-	Chem						
Surface emission	1010	1010	306	306	288	288	18	18	0.3	0.
Total CP	1686	1670	770	729	646	600	116	121	8.0	8.
CH ₄ oxidation	1046	1072	455	470	360	375	89	90	5.6	5.
NMVOC oxidation	640	598	315	259	286	225	27	31	2.4	2.
Chemical loss	2749	2689	1142	1088	897	855	224	212	22	2
			CAM-c	chem						
Surface emission	1010	1010	306	306	288	288	18	18	0.3	0.
Total CP	1263	1210	552	504	465	415	82	84	4.8	4.
CH₄ oxidation	862	890	364	383	282	301	77	77	4.7	4.
NMVOC oxidation	401	320	188	121	183	114	5	7	0.1	0.
Chemical loss	2068	2021	804	771	632	612	159	147	13	1

M: CLM-MEGANv2.1 emissions; **G**: LPJ-GUESS emissions; CP: chemical production. Units in Tg yr $^{-1}$; * A conversion factor of 1.0 from methane oxidation is assumed here for diagnostic purposes; ** NMVOCs oxidation is derived from total chemical production and methane oxidation, i.e. $CP_{NMVOCs} = CP_{Total} - CP_{CH_4}$.

Table 5. Tropospheric HCHO budget for year 2004.

	Glo	bal	SH		0-3	80S	30-	60S	60-90S	
	М	G	М	G	М	G	М	G	М	G
			NIWA	-UKCA						
Total source	1839	1764	777	723	672	612	101	105	6.2	6.6
Surface emission	13	13	4.9	4.9	4.7	4.7	0.2	0.2	0.0	0.0
Total CP	1826	1751	772	718	667	607	101	103	6.2	6.6
CP from CH ₄	1137	1164	468	483	378	393	83	83	5.3	5.6
CP from NMVOC	689	587	304	235	289	214	17	20	0.9	1.0
Total sinks	1839	1764	780	726	672	612	102	107	6.3	6.7
OH + HCHO	507	519	190	193	170	172	20	20	0.5	0.5
HCHO + $h\nu$	1248	1170	553	502	471	415	76	81	5.5	6.0
Dry deposition	24	21	10	9	9	7	1.3	1.5	0.0	0.0
Wet deposition	58	53	26	23	22	18	4	4.4	0.1	0.4
			TI	M5						
Total source	1748	1647	777	704	674	593	99	106	4.1	4.4
Surface emission	13	13	4.9	4.9	4.7	4.7	0.2	0.2	0.0	0.0
Total CP*	1735	1634	772	699	670	588	99	106	4.1	4.4
CP from CH ₄	927	993	387	418	304	334	79	80	2.2	2.3
CP from NMVOC	808	641	385	281	366	254	20	25	1.9	2.1
Total sinks	1748	1647	777	704	674	593	99	106	4.1	4.4
OH + HCHO	293	317	106	115	93	101	13	14	0.2	0.2
HCHO + $h\nu$	1247	1156	577	515	499	430	75	81	3.4	3.6
Dry deposition	34	28	13	10	12	9	1	1	0.0	0.0
Wet deposition	174	146	81	64	70	53	10	10	0.5	0.6

 $^{^{\}star}$ Total chemical productions in TM5 are balanced by total sinks and surface emissions.

Table 6. Tropospheric CO, HCHO, and OH burden for year 2004.

	Glo	bal	S	Н	0-30S		30-60S		60-90S	
	М	G	М	G	М	G	М	G	М	G
			NIV	VA-UK	CA					
CO (Tg)	341	319	134	120	84	74	40	36	10	9.5
HCHO (Gg)	912	846	471	429	378	330	85	90	7	8
OH (Mg)	206	216	133	141	94	101	32	33	6.7	7.1
				TM5						
CO (Tg)	377	331	174	143	105	84	54	46	15	13
HCHO (Tg)	770	714	357	308	305	254	50	52	2.0	2.1
OH (Mg)	196	216	85	96	65	75	18	20	1.6	1.7
			GE	OS-Ch	em					
CO (Tg)	323	307	142	130	86	77	46	42	12	11
HCHO (Gg)	1052	1045	473	451	372	346	90	93	11	11
OH (Mg)	262	272	120	126	91	96	26	27	2.9	3.0
			CA	M-che	m					
CO (Tg)	264	246	113	100	72	63	33	30	8.2	7.6
HCHO (Gg)	733	700	320	291	258	227	57	59	4.6	4.7
OH (Mg)	207	221	92	101	69	77	21	21	2.3	2.3

Table 7. Relative differences (%) in tropospheric CO budget terms between CLM-MEGANv2.1 and LPJ-GUESS simulations for 2004.

	Tota	otal CP CP CH		CH ₄	CP NMVOCs		Total CL		Burden	
	NH	SH	NH	SH	NH	SH	NH	SH	NH	SH
	CO									
NIWA-UKCA	-2.1	-9.6	8.0	3.2	-6.2	-24.2	-2.9	-6.7	-3.9	-10.4
TM5	-3.9	-10.7	6.5	8.0	-14.4	-28.5	-2.1	-7.4	-7.4	-17.8
GEOS-Chem	2.7	-5.3	1.9	3.3	4.3	-17.8	-0.4	-4.7	-2.2	-8.5
CAM-chem	-0.7	-8.7	1.8	5.2	-6.6	-35.6	-1.1	-4.1	-3.3	-11.5
				НС	СНО					
NIWA-UKCA	-2.0	-7.0	1.8	3.2	-8.6	-22.7	-1.8	-6.5	-5.4	-8.9
TM5	-2.9	-9.5	6.5	8.0	-14.9	-27.5	-1.6	-7.8	-1.7	-13.7

Differences calculated as 100×(LPJ-GUESS minus CLM-MEGANv2.1)/CLM-MEGANv2.1.

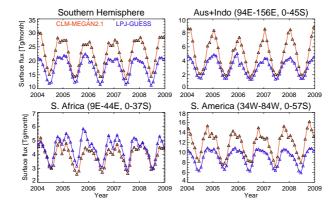


Figure 1. Regional emission fluxes for isoprene between 2004 and 2008 from the CLM-MEGANv2.1 and LPJ-GUESS emission inventories.

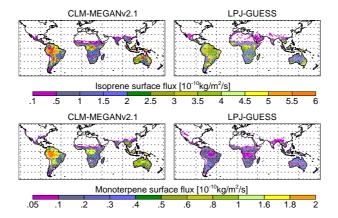


Figure 2. Isoprene and monoterpene emission distributions from CLM-MEGANv2.1 and LPJ-GUESS for January 2005.

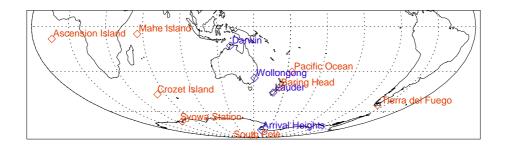


Figure 3. SH FTIR (blue) and NOAA GMD surface CO (red) measurement sites

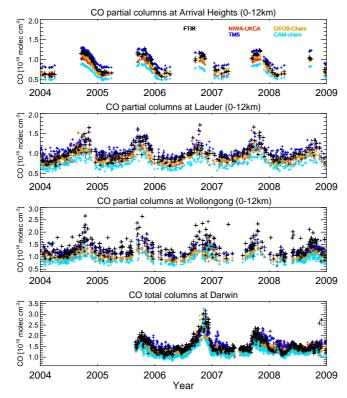


Figure 4. Modelled and observed daily mean FTIR CO columns at SH stations from 4 models. Simulations use CLM-MEGANv2.1 biogenic emissions.

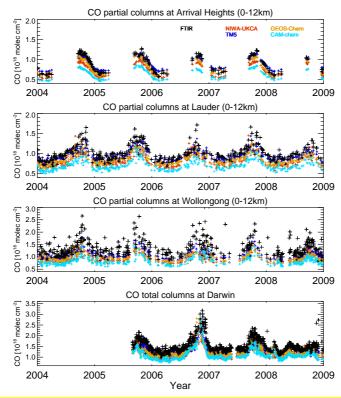


Figure 5. Modelled and observed daily mean FTIR CO columns at SH stations from 4 models. Simulations use LPJ-GUESS biogenic emissions.

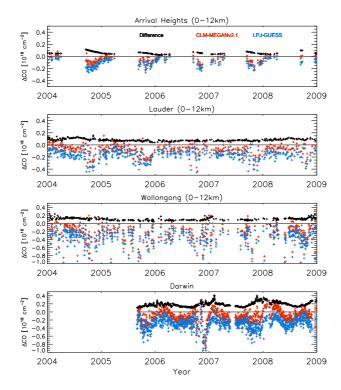


Figure 6. Deviations of model ensemble- and daily-mean CO columns from the observed FTIR CO columns with CLM-MEGANv2.1 simulation (red) and with LPJ-GUESS simulation (blue) respectively. The difference between the modelled CO columns from these two simulations are displayed in black symbols ($CO_{CLM-MEGANv2.1} - CO_{LPJ-GUESS}$).

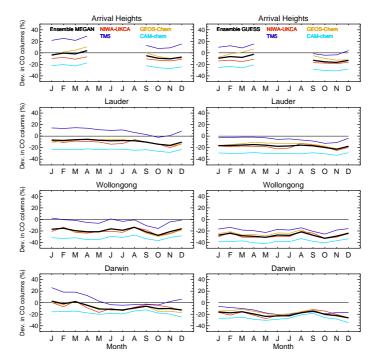


Figure 7. Percentage differences between modelled and observed multi-annual mean CO columns at Arrival Heights, Lauder, Wollongong, and Darwin from two simulations with CLM-MEGANv2.1 (left) and LPJ-GUESS emissions (right) respectively.

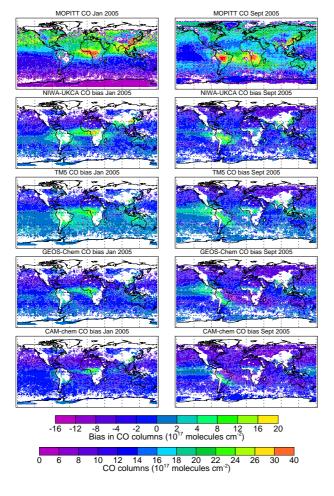


Figure 8. Monthly mean modelled and MOPITT CO columns for January and September 2005. Modelled data are convolved with MOPITT averaging kernels and a priori data. Simulations are with CLM-MEGANv2.1 biogenic emissions.

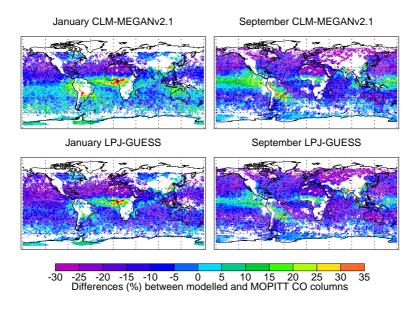


Figure 9. Percentage differences between ensemble model mean and MOPITT CO columns for January and Spetember 2005, from two simulations with CLM-MEGANv2.1 (top) and LPJ-GUESS (bottom) biogenic emissions, respectively.

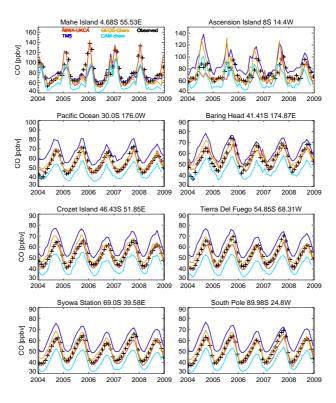


Figure 10. Modelled monthly mean surface CO with CLM-MEGANv2.1 emissions (coloured lines) and observed monthly mean surface CO at SH sites. Observations are from the NOAA GMD network (Novelli et al., 1998): http://www.esrl.noaa.gov/gmd/.

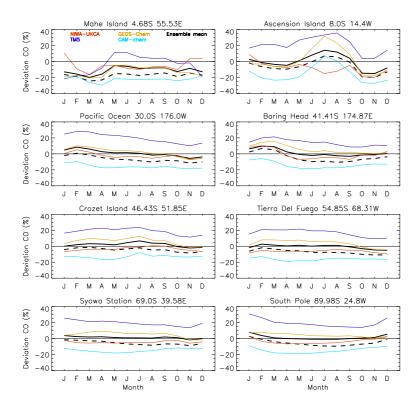


Figure 11. Percentage differences between monthly mean modelled and observed surface CO; Solid black lines for CLM-MEGANv2.1 ensemble and dashed black lines for LPJ-GUESS ensemble. Individual model deviations (colored lines) are from the CLM-MEGANv2.1 simulations only. Data are averaged over 2004-2008.

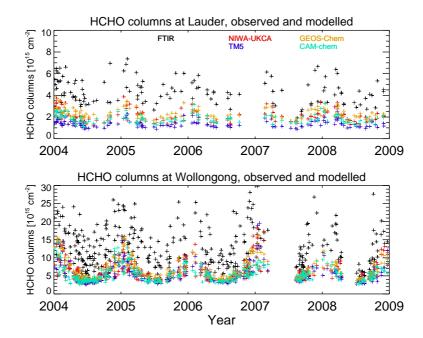


Figure 12. Modelled and observed daily mean FTIR HCHO columns at Lauder and Wollongong. Simulations use CLM-MEGANv2.1 biogenic emissions.

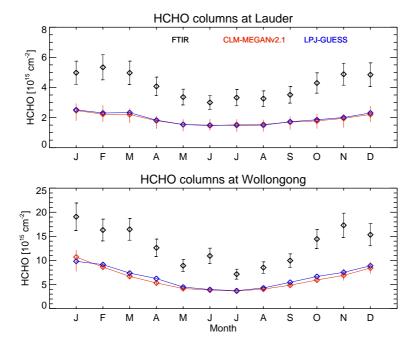


Figure 13. Observed (black symbols) FTIR multi-annual mean HCHO columns and ensemble and model mean from CLM-MEGANv2.1 (red) and LPJ-GUESS (blue) simulations. Measurement errors are shown by vertical bars (black). Model ranges from the CLM-MEGANv2.1 simulations are also given (coloured vertical bars).

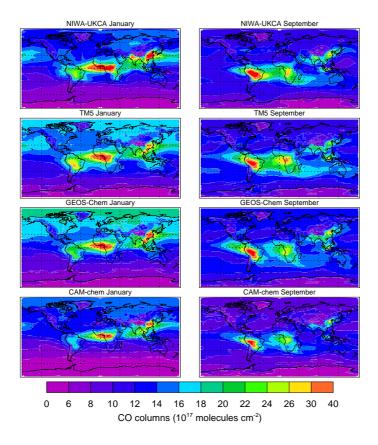


Figure 14. Tropospheric CO columns from the 4 models for January (left) and September (right) 2005, for the CLM-MEGANv2.1 simulations.

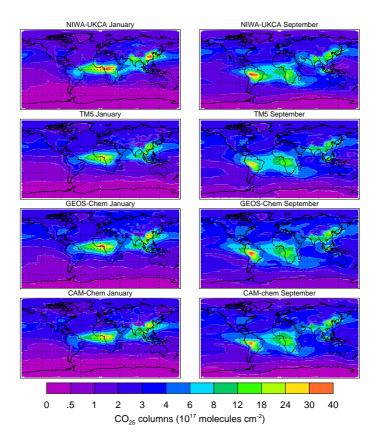


Figure 15. Tropospheric CO₂₅ tracer columns from 4 models for January (left) and September (right) 2005, for the CLM-MEGANv2.1 simulations.

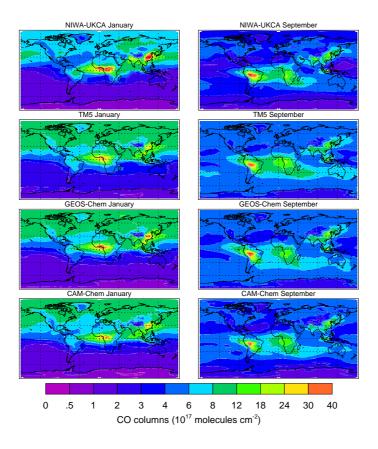


Figure 16. Tropospheric CO_{OH} tracer columns from 4 models for January (left) and September (right) 2005, for the CLM-MEGANv2.1 simulations.

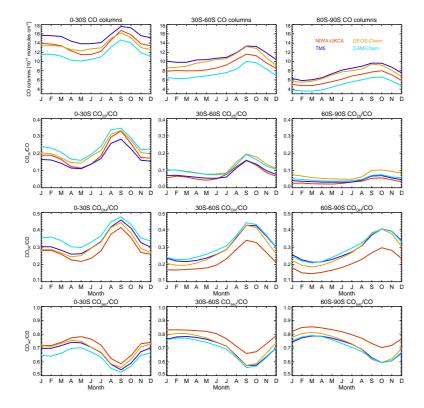


Figure 17. Monthly mean CO columns (top) and the ratio of CO_{25} to CO, CO_{OH} to CO, and CO_{sec} to CO columns averaged over three SH regions (0–30° S, 30–60° S, and 60–90° S) for CO, CO_{25} , CO_{OH} , and CO_{sec} . Data are for the year 2005.

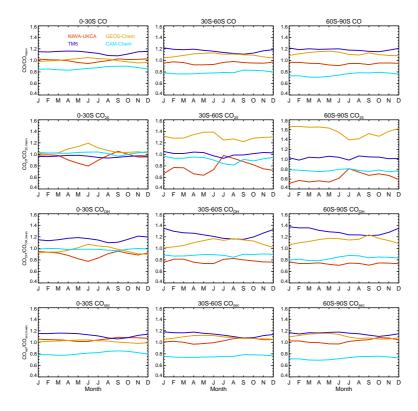


Figure 18. The ratio of individual models to the ensemble mean columns averaged over three SH regions (0–30 $^{\circ}$ S, 30–60 $^{\circ}$ S, and 60–90 $^{\circ}$ S) for CO, CO₂₅, CO_{OH}, and CO_{sec}. Data are for the year 2005.

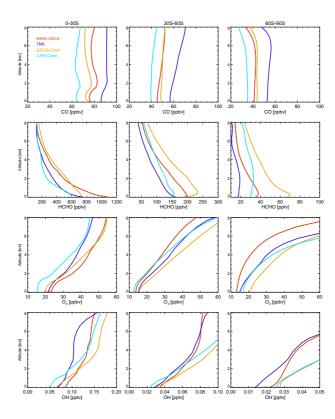


Figure 19. Monthly mean mixing ratios averaged over three SH regions $(0-30^{\circ} \text{ S}, 30-60^{\circ} \text{ S}, \text{ and } 60-90^{\circ} \text{ S})$ for CO, HCHO, O₃, and OH. Data are for January 2005.

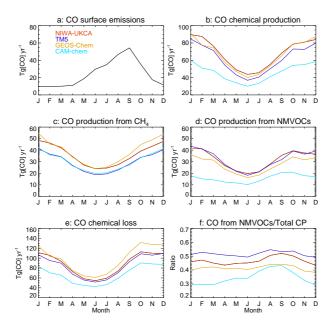


Figure 20. Monthly CO surface emissions, chemical production and loss terms, and the ratio of NMVOC oxidation to total chemical production in the SH. Data are for 2004.

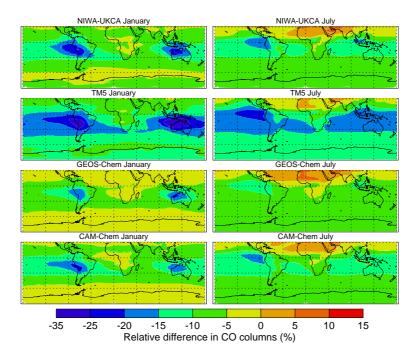


Figure 21. Relative differences (%) in modelled CO columns between the LPJ-GUESS and the CLM-MEGANv2.1 simulations from 4 models for January (left) and July (right). Results are expressed as " $100 \times (\text{CO}_{\text{LPJ-GUESS}} - \text{CO}_{\text{CLM-MEGANv2.1}})/\text{CO}_{\text{CLM-MEGANv2.1}}$ ". Data are averaged over 2004-2008.

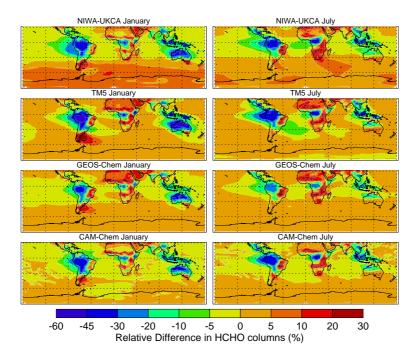


Figure 22. Same as Fig. 15, but for HCHO.

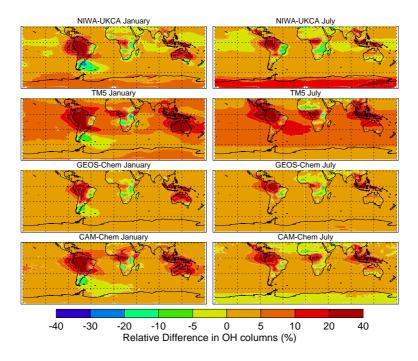


Figure 23. Same as Fig. 15, but for OH.

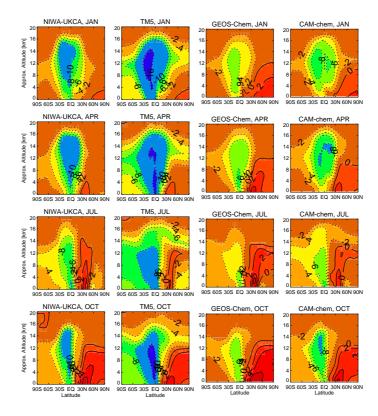


Figure 24. Zonal mean and monthly mean differences in CO [ppbv] between the LPJ-GUESS and the CLM-MEGANv2.1 simulations (expressed as "CO_{LPJ-GUESS}—CO_{CLM-MEGANv2.1}") for January, April, July, and October. Data are averaged over 2004–2008.