Evaluation of the Australian Community Climate and Earth-System Simulator Chemistry-Climate Model

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Abstract

Chemistry climate models are important tools for addressing interactions of composition and climate in the Earth System. In particular, they are used for assessing the combined roles of greenhouse gases and ozone in Southern Hemisphere climate and weather. Here we present an evaluation of the Australian Community Climate and Earth System Simulator-Chemistry Climate Model, focusing on the Southern Hemisphere and the Australian region. This model is used for the Australian contribution to the international Chemistry-Climate Model Initiative, which is soliciting hindcast, future projection and sensitivity simulations. The model simulates global total column ozone (TCO) distributions accurately, with a slight delay in the onset and recovery of springtime Antarctic ozone depletion, and consistently higher ozone values. However, October averaged Antarctic TCO from 1960 to 2010 show a similar amount of depletion compared to observations. A significant innovation is the evaluation of simulated vertical profiles of ozone and temperature with ozonesonde data from Australia, New Zealand and Antarctica from 38 to 90°S. Excess ozone concentrations (up to 26.4% at Davis during winter) and stratospheric cold biases (up to 10.1 K at the South Pole) outside the period of perturbed springtime ozone depletion are seen during all seasons compared to ozonesondes. A disparity in the vertical location of ozone depletion is seen: centered around 100 hPa in ozonesonde data compared to above 50 hPa in the model. Analysis of vertical chlorine monoxide profiles indicates that colder Antarctic stratospheric temperatures (possibly due to reduced mid-latitude heat flux) are artificially enhancing polar stratospheric cloud formation at high altitudes. The models inability to explicitly simulate supercooled ternary solution may also explain the lack of depletion at lower altitudes. The simulated Southern Annular Mode (SAM) index compares well with ERA-Interim data. Accompanying these modulations of the SAM, 50 hPa zonal wind differences between 2001–2010 and 1979–1998 show increasing zonal wind strength southward of 60°S during December for both the model simulations and ERA-Interim data. These model diagnostics shows that the model reason-
ably captures the stratospheric ozone driven chemistry-climate interactions important for Australian climate and weather while highlighting areas for future model development.

1 Introduction

Coupled chemistry-climate models are designed to address the interactions between atmospheric chemistry and the other components of the climate system. This involves the interactions between ozone, greenhouse gases (GHGs), and the dynamics of climate and weather. Improved understanding of these links is important for the Australian region due to the regular springtime Antarctic ozone depletion and its role in modulating Southern Hemispheric surface climate. The Australian region will be affected by these interactions over the course of this century due to ozone recovery as well as changes in GHGs. Thus, global collaborations, such as the currently ongoing Chemistry-Climate Model Initiative (CCMI) (Eyring et al., 2013b) and past chemistry climate modelling projects, will help shape our understanding of future Australian weather and climate.

The annual springtime depletion of Antarctic ozone is attributed to the anthropogenic emissions of ozone-depleting substances (ODSs), mostly chlorofluorocarbons (CFCs), the presence of the polar vortex, and the formation of polar stratospheric clouds (PSCs) within it (Solomon, 1999). In 1987, the Montreal Protocol was signed to phase out the production and release of ODSs into the atmosphere. This has been very effective in halting the build-up of halogens in the stratosphere, with ozone depletion presently not strengthening anymore, and peaking around the year 2000 (Dameris et al., 2014). This marks the first phase of ozone recovery, with the second phase being when ozone is consistently increasing. Antarctic ozone depletion over the previous half century has had a significant influence, equal to GHG increases, on Southern Hemisphere tropospheric climate during summer, mostly through the cooling of the stratosphere by ozone depletion affecting the Southern Annular Mode (SAM) in the late spring and summer, thus shifting surface wind patterns (Gillett and Thompson, 2003; Shindell and Schmidt, 2004; Arblaster and Meehl, 2006; Thompson et al., 2011; Canziani et al., 2014). Another obvious surface impact is an increase in ultraviolet (UV) radiation reaching the surface (World Meteorological Organization, WMO, 2011, 2014). Therefore future climate change in the Australian region is expected to be influenced both by stratospheric ozone recovery and by changes in GHG concentrations (Arblaster et al., 2011). Anthropogenic emissions of GHGs are also expected to influence stratospheric ozone concentrations, both through their dynamical and their chemical effects. GHG-induced cooling of the stratosphere is expected to contribute to an increase in the rate of ozone recovery by slowing gas-phase ozone loss reactions (Barnett et al., 1975; Jonsson et al., 2004). A warming troposphere and associated changes in wave activity propagation from the troposphere into the stratosphere are also predicted to speed up the Brewer–Dobson circulation (Butchart et al., 2006). Thus, the combined effects of a cooler stratosphere and a strengthening of the Brewer–Dobson circulation, causing a speedup of tropical stratospheric ozone advection to mid-latitudes, is expected to reduce the recovery rate in tropical stratospheric ozone, or even cause tropical ozone to decrease again later this century (Austin et al., 2010), and produce a larger recovery trend in the mid-latitudes (Shepherd, 2008; Li et al., 2009).

A simulation of these interacting processes is required to fully capture and assess the impact of future ozone recovery alongside increasing GHGs for many aspects of Australian climate, such as westerly winds and Southern Australian rainfall patterns. The Australian Community Climate and Earth-System Simulator-Chemistry Climate Model (ACCESS-CCM) is used to produce hindcast and future projections, as well as sensitivity simulations to help address these questions and contribute to the CCMI project. CCMI is designed to bring together the current generation of global chemistry models, which are coupled to an interactive ocean, to perform simulations to an agreed standard to help address questions relating to chemistry-climate interactions and inform future ozone assessments and Intergovernmental Panel on Climate Change (IPCC) reports. It also follows on from past chemistry climate modelling comparisons, such
as the Chemistry Climate Model Validation (CCMVal) activity (SPARC-CCMVal, 2010), the Atmospheric Chemistry and Climate Model Inter-comparison Project (ACCMIP) (Lamarque et al., 2013), and Atmospheric Chemistry and Climate Hindcast (AC&C Hindcast) simulations which informed the 5th Assessment Report of IPCC.

In this paper we describe the key components of the model we have used in our contribution to CCMI, which marks the first Australian contribution to an international chemistry-climate modelling project. We also describe the two main simulation setups used in this paper for the evaluation of the model. These include hindcast historical simulations and future projections. An evaluation of the model performance and an analysis of the simulation output, focusing on the Southern Hemisphere, are described. Emphasis is placed on diagnosing the model performance through analysis of ozone and temperature vertical profiles at Australian, New Zealand and Antarctic sites. Analysis of diagnostics related to climate impacts most relevant to the Australian region, such as shifting surface winds through analysis of the SAM metric and the stratospheric polar vortex are also included.

2 Model description

The model is based on New Zealand’s National Institute of Water and Atmospheric Research (NIWA) version of the United Kingdom Chemistry and Aerosols (UKCA) chemistry-climate model (NIWA-UKCA) (Morgenstern et al., 2009, 2014). It includes the HadGEM3 background climate model in the Global Atmosphere (GA) 2 configuration (Hewitt et al., 2011), with the UKCA module for the chemistry component (Morgenstern et al., 2013; O’Connor et al., 2014). It also incorporates the United Kingdom Meteorological Office’s (UKMO) Surface Exchange Scheme-II (MOSES-II). The model setup does not currently incorporate an interactive coupled ocean model; instead, prescribed time-evolving sea surface temperatures (SSTs) and sea ice concentrations (SICs) are used. The model is run at an N48 (3.75° longitude by 2.5° latitude) horizontal resolution and L60 (60 hybrid height levels) vertical resolution with a model top of 84 km.

HadGEM3 has a non-hydrostatic setup (Davies et al., 2005) and a semi-Lagrangian advection scheme (Priestley, 1993). Gravity wave drag is made up of both an orographic gravity wave drag component (Webster et al., 2003) and a parameterised spectral gravity wave drag component, representing the non-orographic components (Scaife et al., 2002). Radiation is described by Edwards and Slingo (1996) and has nine bands in the long-wave part of the spectrum ranging from 3.3 µm to 1.0 cm and six bands in the short-wave part of the spectrum ranging from 200 nm to 10 µm.

The UKCA module includes both stratospheric and tropospheric chemistry with 90 chemical species, including species involved in \( O_x \), \( NO_x \), \( HO_x \), \( BrO_x \), and \( ClO_x \) chemical family chemistry (Banerjee et al., 2014; Archibald et al., 2011). Appropriate species undergo dry and wet deposition. The chemical species undergo over 300 reactions, including bimolecular, termolecular, photolysis, and heterogeneous reactions on polar stratospheric clouds (PSCs). The model assumes two different kinds of PSCs, namely type II water ice and type Ia nitric acid trihydrate (NAT); which is assumed to be in equilibrium with gas phase nitric acid (HNO₃). Both undergo irreversible sedimentation, causing dehydration and denitrification of the polar vortex during winter (Morgenstern et al., 2009). Type 1b supercooled ternary solution of H₂SO₄-H₂O-HNO₃ (STS) PSCs are not explicitly simulated. However, reactions on the surface of liquid sulphuric acid are included. Photolysis reactions are calculated by the FASTJX scheme (Neu et al., 2007; Telford et al., 2013).

The model runs evaluated in this paper include the CCMI hindcast run, labeled REF-C1 from 1960–2010 and the historical part of a future projection run, labeled REF-C2 from 1960–2010 (Eyring et al., 2013b). For the REF-C1 run, SSTs and SICs are gridded fields based on observations from the Hadley Centre HadISST dataset (Rayner et al., 2003). GHGs are from Meinshausen et al. (2011) and Riahi et al. (2011) and follow the Representative Concentration Pathway 8.5 (RCP 8.5) after 2005. RCP 8.5 represents a greenhouse gas concentration pathway that will result in a mean pre-
dicted radiative forcing of 8.5 W m\(^{-2}\) at the top of the atmosphere by 2100 relative to pre-industrial values. ODSs follow the emission scenario that is balanced across all sources (A1B scenario) from World Meteorological Organization, WMO (2011). Anthropogenic and biofuel emissions follow Granier et al. (2011). Biomass burning emissions follow van der Werf et al. (2006); Schultz et al. (2008) and Lamarque et al. (2011). For the REF-C2 run, the only change before 2005 is that SSTs and SICs are climate model estimates taken from a HadGEM2-ES r1p1i1 CMIP5 model run (Jones et al., 2011), and after 2005, all forcings follow RCP 6.0.

3 Observational datasets

Evaluation of the model is undertaken by comparing output to different observation and model datasets, described below.

3.1 Total column ozone database

Simulated total column ozone (TCO) is evaluated against the monthly averaged TCO database (Bodeker et al., 2005). This database is assimilated from satellite observations and spans the period from 1979–2012, where offsets between datasets have been accounted for using Dobson and Brewer ground-based observations. It is important to note that it may not be prudent to directly compare Antarctic wintertime observations from this dataset to model data. This is because of the satellite-assimilated data only being available in sunlit hours, which is in clear deficiency during the Antarctic winter.

3.2 CCMVal-2

The CCMVal-2 project is described extensively in SPARC-CCMVal (2010), and was designed as a coordinated inter-comparison of eighteen chemistry climate models that performed hindcast historical, future projection, and sensitivity simulations. CCMI serves as the next iteration in this project, with improved chemistry climate models. We use the historical simulations from the CCMVal-2 dataset, from 1960 to 2005, labeled REF-B1, to evaluate time-series of Antarctic TCO, stratospheric temperature, and stratospheric winds from the REF-C1 and the historical part of the REF-C2 simulation.

3.3 CMIP5

The Coupled Model Inter-comparison Project Phase 5 (CMIP5) evaluates coupled ocean-atmosphere models (Taylor et al., 2012), and includes some chemistry climate models. We use the recent past (1960–2005) of the historical simulations from CMIP5 models that used prescribed ozone in the evaluation of the seasonal SAM index for the REF-C1 and the historical period of the REF-C2 simulations.

3.4 ERA-Interim

ERA-Interim re-analysis data, from the European Centre for Medium-Range Weather Forecasts (ECMWF), is used to compare stratospheric temperature and wind time series from the recent past with the REF-C1 and the recent past segment of the REF-C2 simulations. Observations in conjunction with a forecast model are used to create the dataset (Dee et al., 2011), which spans the period of 1979 to present.

3.5 Ozonesondes

Ozonesondes are balloon-borne instruments that measure the vertical structure of ozone, along with other parameters such as temperature, pressure and humidity over an observation site, typically up to an altitude of around 35 km. In this study we have used ozonesondes at five locations, namely: Melbourne (37.5° S, 145° E), Lauder, NZ (45° S, 169.7° E), Macquarie Island (54.6° S, 158.9° E), Davis (68.5° S, 79° E) and South Pole (90° S, 169° E).
3.6 Microwave Limb Sounder

The Microwave Limb Sounder (MLS) instrument onboard the Aura satellite is used to evaluate vertical profiles of chlorine monoxide (ClO) over the Antarctic region (Santee et al., 2008; Livesey et al., 2011). The Aura satellite orbits in a sun-synchronous orbit with an inclination of 98.2°. The MLS ClO measurements are scientifically useful within the vertical range of 147–1 hPa and comparison of the model data with the MLS ClO measurements has taken into account all data quality control considerations. The data covers the period from late 2004–present. Comparison with the model data has also taken into account the MLS ClO a priori profiles and retrieved averaging kernels to ensure that the two datasets are sampled consistently, this is done by adding the averaging kernel convolved model and a priori difference to the a priori (Livesey et al., 2011).

4 Model evaluation

To evaluate the performance of the model in the Southern Hemisphere and the Australian region, we have compared model data from the REF-C1 hindcast run and the historical part of the REF-C2 run to observations and ERA-Interim data. A map of global ozone, as well as time series of October averaged Antarctic TCO, stratospheric temperature, and stratospheric winds are used to investigate the model's performance in simulating springtime ozone depletion and its stratospheric drivers and consequences. To analyse the influences of dynamical transport and chemistry on the stratosphere, model-simulated ozone and temperature vertical profiles are compared to ozonesonde data from the five sites listed in Sect. 3.5. To analyse the difference in ozone vertical profiles over the Antarctic region, vertical ClO profiles from the MLS instrument are compared for the zonal area average of 67–70° S.

The model’s ability to simulate the influence of ozone depletion on the SAM was investigated by comparing the seasonal SAM index time series with CMIP5 models and ERA-Interim data, and by comparing stratospheric zonal wind differences with ERA-Interim data. The combination of these metrics and diagnostics gives a comprehensive description of the model’s improvements and differences from the CCMVal-2 ensemble and differences from observations, as well as the model’s capability to simulate important metrics for Australian climate and weather.

4.1 Global ozone

Figure 1 shows zonally averaged TCO over the 2001–2010 period for the REF-C1 hindcast simulation compared to observations from the Bodeker Scientific TCO database. The yearly zonal structure of TCO compares well to observations. However, there is consistently more ozone almost globally within the REF-C1 simulation. The onset of springtime Antarctic ozone depletion occurs a little later in the REF-C1 simulation compared to the observations. This is accompanied by the maximum in ozone depletion occurring later and the persistence of ozone depletion continuing later in the year for the simulation. Despite these temporal differences, the simulated amount of ozone destroyed during the ozone hole period is similar to what is observed.

4.2 Historical time series

Figure 2 compares observations and the CCMVal-2 ensemble with REF-C1 and REF-C2 simulations of Antarctic TCO averaged between 60–90° S for October. The latitude range of 60–90° S was chosen for the ozone comparison, as this area experiences the most significant springtime ozone depletion. The REF-C1 and REF-C2 simulations are consistently producing larger TCOs over the entire historical period examined compared to observations and the CCMVal-2 ensemble. However, the REF-C1 and REF-C2 simulations consistently lay inside the CCMVal-2 10th and 90th percentile, and the total amount of ozone depletion from 1960 to 2010 is similar. The inter-annual variability simulated by the model is not as large as in the observations. There are also slight differences between the REF-C1 and REF-C2 simulations for the historical pe-
period. This can be attributed to the different SST and SIC datasets used, marking the only difference between the REF-C1 and the historical part of the REF-C2 simulation before 2005.

Figure 3 similarly compares the REF-C1 and REF-C2 60–90° S averaged October temperature and 50–70° S average zonal winds to ERA-Interim and the CCMVal-2 ensemble for the stratospheric pressure levels: 100, 50 and 30 hPa. The latitude range between 50–70° S was chosen to examine the strong westerlies forming the polar vortex boundary.

At 100 hPa the REF-C1 and REF-C2 temperature simulations compare well to the ERA-Interim data, in contrast to the CCMVal-2 ensemble median, which shows a substantial cold bias of up to 6 K. The CCMVal-2 ensemble median captures a trend of decreasing temperature; consistent with colder stratospheric temperatures expected to accompany historical ozone depletion. This decreasing temperature is also seen in the REF-C1 and REF-C2 simulations, albeit to a lesser scale. The REF-C1 and REF-C2 zonal wind simulations at 100 hPa compare well with both ERA-Interim and the CCMVal-2 ensemble, with only slightly weaker zonal winds present in CCMVal-2 and the REF-C1 and REF-C2 simulations. This is surprising, as the cold bias present in the 100 hPa CCMVal-2 temperature is expected to be associated with more intense zonal wind. However, these inconsistencies are most likely due to similar temperature gradients between the poles and mid-latitudes seen in both ACCESS-CCM and the CCMVal-2 ensemble. The amount of variation in the REF-C1 and REF-C2 simulations is also similar to what is seen in the ERA-Interim data, and the ERA-Interim data lay entirely within the CCMVal-2 10th and 90th percentiles.

At 50 hPa a significant cold bias exists of around 5 K in the REF-C1 and REF-C2 model runs compared to ERA-Interim data. This is not as pronounced as the CCMVal-2 ensemble median, with ACCESS-CCM being consistently 3 K warmer after 1970. Note the ERA-Interim data still mostly lay within the 10th and 90th percentiles of the CCMVal-2 ensemble (illustrating large inter-model variability). The differences between the CCMVal-2 ensemble and the REF-C1 and REF-C2 simulations is likely a result of the larger ozone concentration present in the ACCESS-CCM model compared to the CCMVal-2 ensemble, as a higher ozone concentration warms the stratosphere through more absorption of UV radiation. A slight decreasing temperature trend is simulated over the historical period, which is not as pronounced as in the CCMVal-2 ensemble.

At 50 hPa there is an intensification of the polar vortex due to colder 50 hPa temperatures in the CCMVal-2 ensemble, however, the REF-C1 and REF-C2 simulations still agree well with ERA-Interim values. The differences between the CCMVal-2 ensemble median and the REF-C1 and REF-C2 simulations increase with time, reaching a maximum of 5 ms\(^{-1}\) at year 2000, and are reflective of the temperature differences.

At 30 hPa, the REF-C1 and REF-C2 simulations of temperature follow the CCMVal-2 ensemble median closely, with a large cold temperature bias relative to ERA-Interim, of 10–15 K. However, again the ERA-Interim mostly lay within CCMVal-2 inter-model variability (10th and 90th percentiles). This cold bias is accompanied by slightly stronger zonal winds in the REF-C1 and REF-C2 simulations compared to ERA-Interim. An even stronger zonal wind is associated with the CCMVal-2 ensemble, with a maximum difference of 5 ms\(^{-1}\). The increasing trend in the polar vortex strength seen in the CCMVal-2 models is not as pronounced in the REF-C1 and REF-C2 simulations.

### 4.3 Ozone, temperature and ClO profiles

Figure 4 shows vertical ozone profiles seasonally averaged over 2001–2010 for the REF-C1 simulation compared to ozonesonde observations averaged over 2003–2012 for five Southern Hemisphere sites and their nearest coincident model grid box. Similarly, Fig. 5 shows vertical temperature profiles averaged over the same time period and locations. To highlight the variability, shaded regions show one standard deviation of the monthly averaged model output for the REF-C1 profiles and one standard deviation divided by \(\sqrt{8}\) for the ozonesonde profiles. The ozonesonde standard deviations are divided by \(\sqrt{8}\) as we have presumed an average of one sounding per week. With the assumption of normal statistics, this will approximate the standard deviation of a monthly average, consistent with the model data used. The largest differences...
between the two datasets for both ozone concentration and temperature are provided in Table 1. Anomalies are visibly present in the upper levels of ozonesonde measurements, particularly in the temperature profiles. At these levels measurement sample size is severely reduced, resulting in possible skewed seasonal averages.

Figures 4 and 5 illustrate that there is general agreement in both ozone and temperature profiles between the ozonesondes and the REF-C1 simulation for Melbourne. The location of the peak in ozone concentration is consistent between REF-C1 and ozonesondes throughout summer, autumn and winter. There is a slight difference during spring, with the model simulating a slightly higher ozone peak altitude relative to ozonesondes. Consistently the model simulates excessive ozone peak concentrations between 20 and 25 km, as shown in Table 1. This is largest for autumn, with an excess of 8 % simulated by the model. The REF-C1 temperature profiles agree well with ozonesondes below 100 hPa. However, above 100 hPa there are consistent cold biases of up to 3.1 K that extend up to 10 hPa during all seasons.

The comparison at Lauder and Macquarie Island illustrates poorer agreement between the REF-C1 simulation and ozonesonde ozone observations. The ozone concentration peak altitudes are still consistent between the datasets, with the largest exception at Macquarie during summer, where the REF-C1 profile peak is situated slightly higher. Again, the model is predicting excess ozone concentration peaks during all seasons, with the largest at Lauder of 13.4 % during summer, and at Macquarie of 20.1 % seem during winter. The REF-C1 temperature profiles generally agree well with ozonesondes. However, there is still a cold bias present above 100 hPa, up to 4.5 K seen during summer at Lauder, and 5.6 K at Macquarie near the tropopause at 170 hPa.

Davis (located within the polar vortex collar region) comparisons of REF-C1 and ozonesonde profiles show very significant differences. During summer, spring, and autumn the simulated ozone maximum is at consistently higher altitudes compared to ozonesondes. The model is also simulating significantly more ozone during autumn and winter, with an excess of 26.4 % in maximum ozone concentration during winter. Simulated summer and to a lesser extent, autumn, temperature profiles also show a cold temperature bias, most noticeable between 250 and 30 hPa of up to 6.1 K. The winter simulated temperature profile agrees well with ozonesondes, in contrast to ozone concentrations, where there is a very large difference. Davis is located in an area that experiences perturbed springtime polar ozone depletion. Here, ozone depletion is captured in the simulated ozone profiles mostly between 50 and 20 hPa. This is in contrast to what is observed by ozonesonde profiles, where the majority of ozone depletion is seen at a lower altitude, below 50 hPa and centered around 100 hPa. This indicates a clear inadequacy of the model in capturing the springtime vertical ozone structure. The simulated temperature profiles at Davis also show a large cold bias above 70 hPa of up to 13.4 K, associated with the altitude of ozone depletion in the model. The variability, seen in the standard deviations is also much larger during spring for ozonesondes and REF-C1 compared to other seasons. This is due to the variable nature of springtime Antarctic ozone depletion, and the location of Davis, which is often in the collar region of the polar vortex.

Due to the dynamical variability experienced by Davis, with Davis being in the polar vortex edge region, comparisons of simulated and ozonesonde vertical ozone concentration and temperature for the South Pole were conducted. The South Pole shows very similar differences between ozonesondes and REF-C1 model simulations for both ozone concentrations and temperature to Davis. Therefore the disparity in the vertical location of springtime ozone depletion seen at Davis is not due to its potential location on the edge of the polar vortex. However, there are some differences. The amount of ozone depletion simulated during spring in the model is now enhanced greatly, with almost all ozone destroyed above 50 hPa. While ozonesondes only show slightly more ozone depletion. The discrepancy in the altitude of significant ozone depletion is still present, with the model simulating ozone depletion much higher than is observed. This produces a more pronounced cold bias in the model above 100 hPa with differences reaching 15.5 K at 30 hPa during spring.
A consistent ozone excess at all stations during seasons that are not perturbed by springtime ozone loss is seen in the vertical ozone profiles, increasing with increasing latitude (Fig. 4). This suggests possible problems with transport in the model. Also, as the model shows excess ozone globally, cold biases above 10 hPa may also be affecting gas phase ozone chemical cycles. On a global average scale, the stratospheric cold biases simulated by the model are likely due to incorrect concentrations and distributions of radiatively active gas or problems with the radiative scheme (SPARC-CCMVal, 2010). The two main radiative gases that are tied into the chemistry scheme in the stratosphere are ozone and water vapour. Global water vapour distributions of a previous iteration of this model where analysed in Morgenstern et al. (2009) and where shown to agree well with ERA-40 climatology.

The large differences seen in the vertical structure of perturbed springtime ozone between the REF-C1 simulation and ozonesondes is either chemical or dynamical in nature, or some combination of both. The slightly colder winter temperatures seen in the model over Antarctic regions can have implications for PSC formation and are likely a result of less pole-ward heat transport, analysed through comparison of 45–75° S heat flux with MERRA reanalysis (not shown). To investigate the links between the chemistry and dynamics of the problem, Fig. 6 shows a comparison of ClO volume mixing ratio, extracted for the zonal region of 67–70° S corresponding to Davis and temporally averaged between 2001–2010 for the REF-C1 simulation and 2005–2014 for MLS satellite observations. The altitude of large ClO volume mixing ratios is an indication of the altitude of where chemical cycles that are responsible for the destruction of ozone are occurring. ClO has a strong diurnal cycle, with concentrations peaking during sunlit hours. This may cause some disparity between the amount of ClO observed by the MLS instrument and simulated by the model. The MLS instrument only measures at any particular latitude at two different times: when travelling northwards or southwards, while the model monthly averages sample ClO at every model time step. Taking a seasonal average of the MLS zonal average data amply samples during both sunlit and dark hours during spring and autumn, providing a fair comparison to the monthly average model data used. However, while this type of comparison does not allow direct quantitative evaluation of the modelled ClO volume mixing ratios, it can give an indication of the vertical locations of the ClO volume mixing ratio peaks. This provides an indication of where the ozone loss chemical reactions are taking place.

During summer, the structure and peak of the simulated ClO profiles agrees well with MLS measurements. During autumn the model simulates a peak in ClO volume mixing ratio at around 20 hPa that is not captured by the observations. This is most likely due to stratospheric cold bias seen in the model causing late autumn polar stratospheric cloud formation. The winter profiles show very good agreement of the ClO peak location below 20 hPa. However, above 20 hPa there are large differences, with the ClO decline and minimum seen in the MLS measurements at 10 hPa being slightly lower in altitude and much less noticeable in the model. The peak in modelled ClO due to gas phase chemistry at higher altitudes is also slightly lower in the REF-C1 simulation compared to MLS observations. The large differences seen in the volume mixing ratios during winter can be attributed to the inability of the MLS instrument capturing the diurnal cycle at these latitudes. There is a large difference between the REF-C1 simulated ClO and that observed by MLS during spring. At 50 hPa and below, the MLS observed peak in ClO is not captured by the REF-C1 simulations. Above 50 hPa, the REF-C1 simulated ClO peak is also a little lower in altitude compared to MLS, similar to what is seen in winter. These ClO observations are consistent with the vertical structure of springtime ozone concentrations, and that the altitude of ozone depletion is misrepresented by our model over Davis and the South Pole.

These results suggest the colder Antarctic stratospheric temperatures above 100 hPa seen in the model are causing enhanced PSC formation at higher altitudes, and thus more heterogeneous reactions on the surface of PSCs. This is indeed the case through analysis of simulated nitric acid trihydrate (NAT) PSCs (not shown), which show persistent upper level (25 km and higher) PSCs throughout winter. Winter temperature profiles at the South Pole show a slight cold bias, agreeing well with the enhancement of PSCs at these levels, and perhaps indicating reduced sedimentation. This is
further highlighted by the disparity in MLS measured and modelled ClO springtime profiles, with MLS showing a peak due to heterogeneous reactions on PSCs at 50 hPa and below that is not captured well by the model. There is also absence of a defined minimum in the modelled ClO profile seen around 20 hPa in MLS measurements. This agrees well with the large differences seen in the vertical location of ozone depletion simulated for at Davis and the South Pole, consistent with the large springtime cold biases present in the model at 50 and 30 hPa. The lack of ozone depletion at lower altitudes compared to ozonesondes could possibly be explained by the lack of STS simulated by the model due to their higher effectiveness at lower altitudes (Solomon, 1999).

4.4 Southern annular mode

Figure 7 shows Southern Hemisphere seasonal SAM indices for REF-C1 and the historical part of REF-C2 compared to ERA-Interim data from 1979–2010 and the recent past section of the historical simulations from CMIP5 runs that used prescribed ozone (Eyring et al., 2013a). The seasonal SAM index was calculated following Morgenstern et al. (2014), using the seasonally averaged difference in area-averaged surface pressure between 38.75–61.25° S and 63.75–90° S. To be able to appropriately compare to ERA-Interim and CMIP5 data, this value was normalised by subtracting the 1979–2005 mean of the calculated SAM indices. The REF-C1, REF-C2 and ERA-Interim seasonal SAM indices are shown as both the yearly seasonal average (highlighting the year-to-year variability) and also as a ten-year running mean (highlighting the comparison to the CMIP5 ensemble). The CMIP5 time series shows the ensemble median and the 10th and 90th percentiles interval of the ensemble range.

During summer the CMIP5 ensemble captures a noticeable increase in the SAM index between 1960–2005, consistent with historical Antarctic ozone depletion. A large range in the ensemble data seen in the 10th and 90th percentiles accompanies this. The REF-C1 and REF-C2 data also agree well with the CMIP5 ensemble median, showing an increase in the simulated SAM index. There is a large amount of year-to-year variability in the REF-C1 and REF-C2 time-series, which mostly lay within the CMIP5 10th and 90th percentiles and very similar to what is seen in the ERA-Interim data. There are also noticeable differences between the REF-C1 and REF-C2 data, mostly before 1985. This can be mostly attributed to different SSTs and the SICs used between the two model runs, or random climate fluctuations. The differences in temporal Antarctic stratospheric ozone depletion between the REF-C1 and REF-C2 would also be an important influence. The increasing SAM index is representative of a southward shift of the westerly winds and precipitation regimes, and is attributed to both decreasing Antarctic stratospheric ozone concentrations and increasing GHGs. An increasing summer SAM index simulated by the model not only agrees with CMIP5 data and ERA-Interim re-analysis, it also complements conclusions from Keeble et al. (2014), which show significant increases in SAM attributed to lower stratospheric ozone depletion within a similar model environment.

Autumn also shows an increase in the SAM index in the CMIP5 ensemble, albeit on a smaller scale to that seen in summer. The REF-C1 and REF-C2 time-series agree well with the CMIP5 data and especially well with the ERA-Interim data. An increase in the SAM index over time is consistent with the CMIP5 ensemble, and the year-to-year variability of the REF-C1 time-series is consistently within the CMIP5 10th and 90th percentiles. However, the REF-C2 seasonal variation shows a frequent low SAM index values outside of the CMIP5 variability, most frequently before 1980. The cause of the positive SAM trend observed during autumn is currently not well understood (Canziani et al., 2014). The seasonal variation seen in the REF-C1 and REF-C2 time-series is also similar to that seen in the ERA-Interim data. The differences between the REF-C1 and REF-C2 time-series are much less pronounced, especially after 1980 where they follow each other closely. The differences before 1980 can be attributed to the different SSTs and SICs used or random climate fluctuations, and less likely due to the differences in stratospheric ozone.

The winter and spring SAM indexes are consistent between all datasets over the entire time-series. There is no noticeable long-term change in the CMIP5 ensemble,
with the REF-C1 and REF-C2 time series agreeing well. The largest excursion from the CMIP5 ensemble median is seen in the REF-C2 time-series centered around 1970 during winter, where a positive SAM index is seen consistently over 3 years. A noticeable difference between the REF-C1 and REF-C2 winter and spring SAM indexes is a strong decadal correlation during spring, in contrast to the winter comparison.

### 4.5 Zonal wind anomalies

Figure 8 shows 50 hPa average zonal winds of 1979–1988 minus the 2001–2010 average for REF-C1, REF-C2, and ERA-Interim data for the months of August, October and December. The ten-year averages represent the earliest time available in the ERA-Interim and the latest time available in the historical simulations, while also being able to represent important phases in stratospheric springtime Antarctic ozone depletion, with 1979–1988 representing the onset of ozone depletion while 2001–2010 representing the maximum springtime ozone depletion. The months of August, October and December were chosen to represent different stages of the annually forming ozone hole. The ozone hole typically begins forming in late August, reaching a maximum by the end of October, and closing by mid-December.

August shows some small-scale differences between the REF-C1 and REF-C2 relative to ERA-Interim, most likely caused by differences in decadal variations between the model and observations. October shows some larger differences, with an opposite dipole in the western hemisphere when comparing REF-C1 and REF-C2 with ERA-Interim. Again, this can be attributed to decadal differences in the variations, and possible differences in the maximum location in zonal wind, which is more pole-ward in ERA-Interim compared to the model simulations. The December differences are very consistent across the REF-C1, REF-C2, and ERA-Interim data, with increasing zonal wind seen south of 60°S. This is an indication of the strengthening of the polar vortex due to Antarctic ozone depletion, and is consistent with the increasing summertime SAM index seen in the ERA-Interim and model simulations.

### 5 Conclusions

The ACCESS-CCM model presented here is able to confidently provide an initial contribution from Australia to the international community via the Chemistry-Climate Model Initiative (CCMI). It simulates slightly larger October total column ozone values compared to observations and the CCMVal-2 ensemble, however simulates a similar ozone decline over the historical period (1960 to 2010). A cold bias compared to ERA-Interim of up to 5 K at 50 hPa and 10–15 K at 30 hPa is present during October. This is an improvement from the CCMVal-2 ensemble, which shows colder temperatures compared to ACCESS-CCM at 100 and 50 hPa of up to of 5 and 3 K respectively. Our model simulates polar vortex strength above 100 hPa closer to ERA-Interim compared to the CCMVal-2 ensemble median.

Model-simulated seasonal averaged vertical profiles of ozone and temperature compared to Southern Hemisphere ozonesondes show very good agreement in ozone vertical distribution, concentration and seasonal variation for Melbourne. However, there is less agreement at higher latitudes sites, with peak ozone concentrations in excess of observed values. The largest difference is seen at Davis during winter, with ACCESS-CCM simulating 26.4% excess. A stratospheric cold bias is also present, mostly above 100 hPa and noticeably over polar latitudes during summer of up to 15.3 K. The majority of springtime ozone depletion at Davis and the South Pole is occurring above 50 hPa in ACCESS-CCM compared to being centered near 100 hPa in ozonesondes. This also induces a significant cold bias in the stratosphere during spring at the altitudes of ozone depletion in the model.

The altitude differences of springtime polar ozone loss can be attributed to differences in simulated ClO profiles during spring, pointing to a modelling deficiency in simulating heterogeneous chlorine release. The MLS instrument shows a peak in ClO at and below 50 hPa. ACCESS-CCM instead shows increasing ClO above 50 hPa. This can be explained by the simulation of colder stratospheric temperatures, possibly caused by reduced mid-latitude heat flux, enhancing PSC formation at these alti-
tudes, and thus providing a mechanism for increased ozone loss at higher altitudes. The deficiency in modelling a large springtime ClO peak at lower altitudes, explains the relatively small simulated ozone loss at these altitudes relative to ozonesonde observations, and could possibly be due to the models inability simulating supercooled ternary solution polar stratospheric clouds.

The large model-ozonesonde differences in the ozone profiles during summer, autumn and winter, seasons outside perturbed polar springtime ozone loss conditions, is consistent with the excess ozone seen in the global total column ozone map (Fig. 1), and time series (Fig. 2). This could possibly be due to too much transport in the model, and cold biases above 10 hPa affecting the gas-phase ozone chemical cycles. The drivers of the cold biases and excessive transport within the ACCESS-CCM are unclear, however, mid-latitude cold biases are likely influenced by incorrect radiatively active gases such as ozone and water vapour or inaccuracies in the radiation scheme. Whereas lower simulated mid-latitude heat flux is likely a driver of the high latitude cold biases.

The SAM index for ACCESS-CCM agrees well with ERA-Interim and CMIP5 ensemble. All show an increasing SAM index during summer and to a lesser extent autumn, indicating a southward shift of mid-latitude winds and storm tracks. Zonal wind differences of 1979–1988 average minus 2001–2010 average at 50 hPa during December show increasing high south latitude wind strength, consistent with the simulated increase in the SAM during summer.

Future versions of this model will follow the UKCA release candidates, with a major goal of obtaining a fully coupled chemistry-climate-ocean model.

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References


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Table 1. Vertical profile maximum differences.

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<th>Macquarie</th>
<th>Davis</th>
<th>South Pole</th>
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Figure 1. Zonally 2001–2010 averaged TCO for REF-C1 hindcast simulation compared to observations from the Bodeker Scientific total column ozone database.

Figure 2. Time series of REF-C1 and REF-C2 TCO averaged between 60–90° S compared with the Bodeker Scientific total column ozone database observations and the CCMVal-2 ensemble. Dashed lines show the October average, while solid lines have undergone a 10 year running mean of October averages. The shaded region shows 10th and 90th percentiles of the CCMVal-2 ensemble.
Figure 3. Time series of REF-C1 and REF-C2 temperature at (a) 100 hPa, (b) 50 hPa, and (c) 30 hPa averaged between 60–90° S and zonal wind at (d) 100 hPa, (e) 50 hPa, and (f) 30 hPa averaged between 50–70° S compared with ERA-Interim and the CCMVal-2 ensemble. The shaded region shows 10th and 90th percentiles of the CCMVal-2 ensemble.

Figure 4. Seasonal average REF-C1 ozone profiles compared to ozonesondes for Melbourne, Lauder, Macquarie Island, Davis and the South Pole. REF-C1 data is averaged between 2001–2010, while ozonesonde data is averaged between 2003–2012. Shaded regions show one standard deviation for REF-C1 and one standard deviation divided by \( \sqrt{8} \) for the ozonesonde data. This is done for statistical consistency as monthly averaged output was used for the REF-C1 data (see Sect. 4.3 in the main text). Altitude values are approximate.
Figure 5. Same as Fig. 4, except for temperature.

Figure 6. Comparison of seasonal average vertical profiles of ClO zonally averaged between 67–70° S. Seasonal average data is from 2001–2010 for REF-C1 and from 2005–2014 for MLS. Shaded regions show one standard deviation. Altitude values are approximate.
Figure 7. Seasonal SAM indexes for REF-C1 and REF-C2 simulations compared to ERA-Interim data and the CMIP5 ensemble. Dashed lines show seasonal averages, while the solid lines have undergone a 10 year running mean of seasonal averages. Shaded regions show the 10th and 90th percentiles of the CMIP5 ensemble.

Figure 8. 2001–2010 minus 1979–1988 50 hPa zonal wind anomaly maps for REF-C1 and REF-C2 simulations compared to ERA-Interim data.
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