Assessing the sensitivity of the hydroxyl radical to model biases in composition and temperature using a single-column photochemical model for Lauder, New Zealand

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Abstract. We assess the major factors contributing to local biases in the hydroxyl radical (OH) as simulated by a global chemistry–climate model, using a single-column photochemical model (SCM) analysis. The SCM has been constructed to represent atmospheric chemistry at Lauder, New Zealand, which is representative of the background atmosphere of the Southern Hemisphere (SH) mid-latitudes. We use long-term observations of variables essential to tropospheric OH chemistry, i.e. ozone (O3), water vapour (H2O), methane (CH4), carbon monoxide (CO), and temperature, and assess how using these measurements affect OH calculated in the SCM, relative to a reference simulation only using modelled fields. The analysis spans 1994 to 2010. Results show that OH responds approximately linearly to correcting biases in O3, H2O, CO, CH4, and temperature. The biggest impact on OH is due to correcting an overestimation by approximately 20 to 60 % of H2O, using radiosonde observations. Correcting this moist bias leads to a reduction of OH by around 5 to 35 %. This is followed by correcting predominantly overestimated O3. In the troposphere, the model biases are mostly in the range of −10 to 30 %. The impact of changing O3 on OH is due to two pathways; the OH responses to both are of similar magnitude but different seasonality: correcting in situ tropospheric ozone leads to changes in OH in the range −14 to 4 %, whereas correcting the photolysis rate of O3 in accordance with overhead column ozone changes leads to increases of OH of 8 to 16 %. The OH sensitivities to correcting CH4, CO, and temperature biases are all minor effects. The work demonstrates the feasibility of quantitatively assessing OH sensitivity to biases in longer-lived species, which can help explain differences in simulated OH between global chemistry models and relative to observations. In addition to clear-sky simulations, we have performed idealized sensitivity simulations to assess the impact of clouds (ice and liquid) on OH. The results indicate that the impacts on the ozone photolysis rate and OH are substantial, with a general decrease of OH below the clouds of up to 30 % relative to the clear-skies situation, and an increase of up to 15 % above. Using the SCM simulation we calculate recent OH trends at Lauder. For the period of 1994 to 2010, all trends are insignificant, in agreement with previous studies. For example, the trend in total-column OH is 0.5 ± 1.3 % over this period.

1 Introduction

The hydroxyl radical (OH) is essential to atmospheric chemistry as the leading oxidizing agent. It acts as a “detergent”, reacting with numerous, mostly organic pollutants (Levy, 1971; Logan et al., 1981; Thompson, 1992; Lelieveld et al., 2004; Naik et al., 2013) and controls the lifetimes of many trace gases containing carbon–hydrogen bonds, particularly methane (CH4), because reaction with OH is their dominant removal mechanism. It is also responsible for oxidizing at-
mospheric trace gases such as carbon monoxide (CO), non-methane volatile organic compounds (NMVOCs), and also some ozone-depleting substances such as hydrochlorofluorocarbons (HCFCs) (DeMore, 1996). Therefore, the oxidizing capacity of the atmosphere is largely defined by the abundance of OH. Tropospheric ozone (O₃), an air pollutant and greenhouse gas (GHG), is the primary source of OH in the troposphere. Although it only accounts for 10% of the total atmospheric O₃ abundance, it plays an essential role in photochemical processes controlling tropospheric composition. It forms OH via O₃ photolysis yielding excited oxygen (O(¹D)) and a subsequent reaction of O(¹D) with water vapour (H₂O). CH₄ and CO oxidation by OH, and other oxidation processes involving NMVOCs, lead to formation of tropospheric O₃ in the presence of NOₓ (Logan et al., 1981; Thompson, 1992; Lelieveld and Dentener, 2000). In low-NOₓ atmospheric environments, such as in much of the SH, downward transport of O₃ from the stratosphere is the main source of tropospheric O₃, followed by O₃ transport from other regions where it is chemically produced (Zeng et al., 2010). Stratospheric O₃ also plays an important role through its impact on the O₃ photolysis rate j(0D) which is affected by the overhead O₃ column. For instance, stratospheric O₃ depletion produces increased UV penetration to the troposphere. This affects the production of tropospheric OH.

The most widely used method for field measurements of OH is the Fluorescence Assay by Gas Expansion (FAGE) technique and is based on the measurement of OH and other species concentrations through ultra-violet (UV) laser-induced fluorescence spectroscopy. OH measurements using the FAGE technique have been conducted in a large variety of atmospheric environments, ranging from polluted (Ren et al., 2003; Dusanter et al., 2009) to clean (Creasey et al., 2003; Bloss et al., 2007) atmospheres. However, due to its very short lifetime (the global lifetime is estimated to be ~1 s, Prinn, 2001; Elshorbany et al., 2012) and large variability, such in situ measurements of OH do not sufficiently capture its global abundance, which makes it difficult to sufficiently constrain global OH abundances with in situ measurements (Heard and Pilling, 2003). For that reason, modelling is an essential tool to study global OH. OH is routinely included in global models of tropospheric chemistry, but the complexity of the tropospheric chemical system and the sensitivity of OH to a variety of environmental factors mean that there is considerable disagreement among global chemistry-transport and chemistry–climate models regarding the global OH abundance; this is often expressed in terms of the CH₄ lifetime (e.g. Stevenson et al., 2006; Naik et al., 2013; Voulgarakis et al., 2013). Several model studies have examined changes in OH abundance and the CH₄ lifetime since pre-industrial times. Chemistry-transport models (which use off-line, precalculated meteorology) generally simulate decreases in OH and increases in the CH₄ lifetime, ranging from 6 to 25% during the 21st century (Thompson, 1992; Lelieveld et al., 1998; Wild and Palmer, 2008). These results differ from those produced by chemistry–climate models which account for changes in both emissions and climate (Stevenson et al., 2000; Johnson et al., 2001; Shindell et al., 2006; Zeng et al., 2010; John et al., 2012). All of them project a reduction in the CH₄ lifetime and an increase in OH. In particular, Shindell et al. (2006) and Zeng et al. (2010) obtain a ~10% decrease in the CH₄ lifetime using different emission scenarios in their simulations. More recent and comprehensive studies compare present-day and future results for OH and the CH₄ lifetime among models participating in the Atmospheric Chemistry and Climate Model Intercomparison Project (ACCMIP, Naik et al., 2013; Voulgarakis et al., 2013). Naik et al. (2013) analyse the evolution of the CH₄ lifetime and OH in ACCMIP models since pre-industrial times (1850–2000). They point out large variations in the sign and magnitude of OH changes (from ~12.7% to 14.6%) amongst ACCMIP models, reflecting uncertainties in natural CO, NOₓ, and NMVOC emissions as well as roles of the diverse chemical mechanisms included in the models. For present-day (year 2000) simulations of OH and the CH₄ lifetime, Voulgarakis et al. (2013) suggest that diversity in photolysis schemes and NMVOC emissions might cause large variations in simulated OH and the CH₄ lifetime. Trends in OH between 2000 and 2100 are mainly attributed to stratospheric O₃ changes and trends in modelled temperature fields.

A useful indirect method for constraining global OH is based on tracking the abundance of long-lived, well mixed chemicals for which oxidation by OH is the dominant sink and which have a well quantified, industrial source. The most widely used such species is methyl chloroform (CH₃CCl₃) (Prinn et al., 2005; Bousquet et al., 2005; Manning et al., 2005; Krol et al., 2008). Montzka et al. (2011) use CH₃CCl₃ measurements to infer only a small interannual variability in OH for 1998–2007. The global multi-model mean OH inferred from the ACCMIP ensemble (Naik et al., 2013) increases slightly (3.5±2.2%) over the period 1980–2000. This result largely agrees with Montzka et al. (2011) and with other models (Dentener et al., 2003; Hess and Mahowald, 2009; John et al., 2012; Holmes et al., 2013), but disagree with other studies of CH₃CCl₃ observations that find a decrease in OH over that period (Prinn, 2001; Bousquet et al., 2005). For the year 2000, Naik et al. (2013) underestimate the CH₃CCl₃ lifetime (and thus overestimate OH) by 5 to 10% relative to observations. CH₃CCl₃ is controlled under the Montreal Protocol, meaning its abundance in the atmosphere is approaching the detection limit and it will no longer be a useful constraint on OH in decades to come.

A further indirect method to address OH is to measure ¹⁴CO. Manning et al. (2005) find some considerable variability but no long-term trend using this method. According to Krol et al. (2008), this method is considerably more sensitive to high-latitude than low-latitude OH, in contrast to the CH₃CCl₃ method which is mostly sensitive to tropical OH.
Therefore, a step forward in addressing the uncertainty in modelling OH in global models is to quantitatively assess the contributions of biases in long-lived species that are central to OH. This sometimes involves juxtaposing global models to local-scale (box or single-column) models constrained as much as possible by observations and incorporating only fast photochemical processes. For example, Emmerson et al. (2005, 2007) develop a box model to assess the sensitivity of OH and HO$_2$ to biases in long-lived species, and compare the model results to observations. However, their analyses only pertain to polluted environments not representative of much of the global atmosphere and only take in episodic and surface measurements. Single-column models have been applied to modelling the atmospheric boundary layer (Mihailovic et al., 2005; Cuxart et al., 2006), diabatic processes (Randall et al., 2003; Bergman and Sardeshmukh, 2004), clouds and aerosols (Kylling et al., 2005; Lebassi-Habtezion and Caldwell, 2015; Dal Gesso et al., 2015), the impacts of GHGs on climate change (Vupputuri et al., 1995), and the chemistry of halogen compounds (Piot and von Glasow, 2008; Joyce et al., 2014). Tropospheric OH chemistry of the remote atmosphere has not been assessed in a single-column model framework before.

In the present paper, we introduce and evaluate a single-column model (SCM) constrained with available long-term observations at Lauder, New Zealand (45° S, 170° E, 370 m above sea level), to investigate how chemistry–climate model biases in long-lived chemical species and temperature affect OH. Lauder is known for its clean air (Stedman and McEwan, 1983; McKenzie et al., 2008) and large diversity of available measurements (it is part of the Network for the Detection of Atmospheric Composition Change (NDACC), Bados et al., 2007; McKenzie et al., 2008; WMO, 2011). Observations made at Lauder include UV radiation and surface, profile, and/or total columns of O$_3$ and several other species. The O$_3$, H$_2$O, and temperature records produced by ozone sondes cover 1986 to the present. Lauder therefore is ideal for this kind of study. The SCM is built around a medium-complexity stratosphere–troposphere chemistry scheme. The model is forced with Lauder observations and/or output from a chemistry–climate model that uses the same scheme (see below). In Sect. 2, we describe the set-up of the SCM, the construction of time series of key species and meteorological parameters that drive the SCM, and the simulations. In Sect. 3, we present results of simulated OH concentrations and trends from the SCM and analyse the sensitivity of OH to various forcings. Conclusions are gathered in Sect. 4.

## 2 Models and simulations

### 2.1 The single-column photochemical model (SCM)

The single-column photochemical model is a stand-alone version of the stratosphere–troposphere chemistry mechanism used by the National Institute of Water and Atmospheric Research (NIWA) and United Kingdom Chemistry and Aerosol (UKCA) model (NIWA–UKCA), which comprises gas-phase photochemical reactions relevant to the troposphere and stratosphere (Morgenstern et al., 2009, 2013; Telford et al., 2013; O’Connor et al., 2014; Morgenstern et al., 2016). For consistency with NIWA–UKCA, the SCM uses the same chemical mechanism. We used a more complex mechanism (which the SCM approach lends itself to), then a direct comparison with the NIWA–UKCA output would no longer be possible, and also the results would be less relevant to other global CCMs characterized by relatively simple chemical mechanisms. The 60 vertical levels of the SCM are the same as in NIWA–UKCA, extending to 84 km. We do not use horizontal interpolation and take profiles of atmospheric properties from the grid point closest to Lauder (45° S, 168.75° E). Unlike NIWA–UKCA, the SCM excludes all non-chemistry processes, such as transport, dynamics, the boundary-layer scheme, radiation, emissions, etc. This means the model is only suitable for assessing fast photochemistry. Forcing data for the SCM are mostly interpolated from 10-daily instantaneous outputs from a NIWA–UKCA simulation (see below), except for those fields for which observational data are used.

Morgenstern et al. (2013) and O’Connor et al. (2014) describe the chemistry scheme included in the SCM. The SCM includes an isoprene oxidation scheme (Pöschl et al., 2000; Zeng et al., 2008; Morgenstern et al., 2016) not included in the NIWA–UKCA model version used by Morgenstern et al. (2013). In addition to CH$_4$ and CO, the model includes a number of primary NMVOC source gases, i.e. ethane (C$_2$H$_6$), propane (C$_3$H$_8$), acetone (CH$_3$COCH$_3$), formaldehyde (HCHO), acetaldehyde (CH$_3$CHO), and isoprene (C$_5$H$_8$). As noted above, emission and deposition of species are not considered in the SCM. The SCM includes a comprehensive formulation of stratospheric chemistry (Morgenstern et al., 2009) comprising bromine and chlorine chemistry and heterogeneous processes on liquid sulfate aerosols. Overall, the model represents 86 chemical species and 291 reactions including 59 photolysis and 5 heterogeneous reactions. The FAST-JX interactive photolysis scheme (Neu et al., 2007; Telford et al., 2013) has been implemented in the SCM; this scheme solves a radiative transfer equation accounting for absorption by ozone. The chemical integration is organized through a self-contained atmospheric chemistry package (Carver et al., 1997), and the differential equations describing chemical kinetics are solved using a Newton–Raphson solver (Morgenstern et al., 2009).

### 2.2 Construction of vertical profiles of forcing species and meteorological parameters

Time series of O$_3$, H$_2$O, CO, and temperature profiles are produced using mainly long-term measurements from Lauder, supplemented with NIWA–UKCA results as de-
Figure 1. (a) Time series of O$_3$ profiles constructed by ozone sonde measurements spliced with MOPI1 measurements. (b) Time series of H$_2$O profiles constructed by radiosonde measurements spliced with NIWA–UKCA H$_2$O (time series of ERAI – NIWA–UKCA H$_2$O is not displayed here). (c) Time series of CH$_4$ profiles constructed by rescaling the NIWA–UKCA CH$_4$ to surface CH$_4$ measurements from Cape Grim (Tasmania). (d) Time series of CO profiles constructed by rescaling the NIWA–UKCA CO to CO measurements from the FTIR spectrometer. (e) Time series of temperature profiles constructed by radiosonde measurements (up to 25 km) merged with NCEP/NCAR reanalyses up to the stratopause (50 km) and a mesospheric climatology based on local lidar measurements. Above 25 km these data are as used in the retrieval of O$_3$ from MOPI1 measurements. The areas within black boxes were filled using a Fourier series gap filling method.
Table 1. Simulations performed with the SCM to assess the contribution of changes in the key forcings to OH chemistry at Lauder under clear-sky conditions. The table includes the type of measurement/dataset used to prescribe the key forcings. The time period of simulation is between 1994 and 2010.

<table>
<thead>
<tr>
<th>Forcing</th>
<th>Data used</th>
</tr>
</thead>
</table>
| O_3     | 1. Kinetics effect: O_3 changes $\rightarrow$ ozone sondes (0–25 km) + MOPI1 (26–84 km) NIWA–UKCA data for other species and temperature  
          2. Photolysis effect: $j_{O_1D}$ changes according to O_3 changes NIWA–UKCA data for all species and temperature  
          3. Kinetics + photolysis effects: O_3 changes $\rightarrow$ ozone sondes + MOPI1 $j_{O_1D}$ changes according to O_3 changes NIWA–UKCA data for other species and temperature |
| H_2O    | 1. Changes in H_2O $\rightarrow$ radiosondes (0–8 km) + NIWA–UKCA H_2O (9–84 km) NIWA–UKCA data for other species and temperature  
          2. Changes in H_2O $\rightarrow$ ERAI (0–8 km) + NIWA–UKCA H_2O (9–84 km) NIWA–UKCA data for other species and temperature |
| CH_4    | Changes in CH_4 $\rightarrow$ rescaled NIWA–UKCA CH_4 to Cape Grim surface CH_4 NIWA–UKCA data for other species and temperature |
| CO      | Changes in CO $\rightarrow$ rescaled NIWA–UKCA CO profiles to FTIR CO NIWA–UKCA data for other species and temperature |
| T       | 1. Kinetics effect: temperature changes $\rightarrow$ NCEP/NCAR reanalyses (26–50 km) + lidar climatology (50–84 km) NIWA–UKCA data for all species  
          2. Photolysis effect: $j_{O_1D}$ changes according to temperature changes NIWA–UKCA data for all species and temperature  
          3. Kinetics + photolysis effects: temperature changes $\rightarrow$ NCEP/NCAR (26–50 km) + lidar climatology (50–84 km) $j_{O_1D}$ changes according to temperature changes NIWA–UKCA data for all species |
| O_3, H_2O, CH_4, CO, T | For H_2O, radiosonde (0–8 km) + NIWA–UKCA (9–84 km) data are used. |

Reference NIWA–UKCA data for all species and temperature

tailed below. Lauder is a member of several international networks, including the NDACC (http://www.ndsc.ncep.noaa.gov), the Global Reference Upper Air Network (GRUAN; http://www.gruan.org), and Global Atmosphere Watch (GAW; http://www.wmo.int/pages/prog/arep/gaw/gaw_home_en.html), where these data are archived and made available. The networks coordinate long-term observations of O_3, various other constituents, and meteorological parameters. Here we briefly describe the procedure of constructing forcing data, using Lauder observations, to be used to constrain the SCM. The resulting profiles are shown in Fig. 1.

O_3 profiles used here are a combination of ozone sonde time series (from the surface to 25 km, Bodeker et al., 1998) combined with the Microwave Ozone Profiler Instrument 1 (MOPI1) time series for altitudes above 25 km (Boyd et al., 2007; Nedoluha et al., 2015), covering 1994 to 2010 (Fig. 1a). The ozone sondes have been launched approximately weekly; this defines the temporal coverage of the forcing data used in the SCM calculations. Microwave measurements used here come as several profiles a day at a variable spacing; we interpolate them to the ozone sonde launch times. Any missing data (during the two periods when the microwave instrument was out of service) are filled using a Fourier series gap-filling method. We compare the two datasets in the height region usefully covered by both (20 to 30 km). The differences between the two measurements range between $-2$ and $+6\%$, and a mean bias that is less than $5\%$. O_3 profiles are linearly interpolated onto the SCM’s grid. Total column ozone calculated by integrating the observed O_3 profiles is also compared to total-column O_3 measured by the Lauder Dobson instrument; the difference is about $5\%$ on average (López Comí, 2016). Lauder ozone measurements have been used in various World Meteo-
H$_2$O profiles have been constructed using the weekly radiosonde measurements of H$_2$O vapour below 8 km (the same soundings that also measure ozone) and NIWA–UKCA model output data above. For validation, we use the monthly National Atmospheric and Oceanic Administration (NOAA) Frost Point Hygrometer (FPH) H$_2$O vapour measurements (Vömel et al., 2007; Hall et al., 2016), which start in 2003. FPHs are more accurate compared to radiosonde hygrometers, particularly for stratospheric conditions. However, due to the later start of the FPH time series and the lower measurement frequency, radiosonde measurements are used in this study. The comparison of FPH and radiosonde H$_2$O reveals differences that are generally less than $\pm 5\%$ in the lower and middle troposphere but generally increase in and above the tropopause region ($\sim 11$ km, López Comí, 2016).

The radiosonde hygrometers have some known problems with measuring low humidity (Miloshevich et al., 2001). This is reflected in the large differences observed particularly at these altitudes (up to $30\%$) and, to a lesser degree, below them (Fig. 2a). In a comparison of NIWA–UKCA output with FPH H$_2$O, larger discrepancies are found throughout the whole troposphere and tropopause region (Fig. 2b), as can be expected from a low-resolution model unconstrained by observations and subject to problems with modelling H$_2$O. Given the consistency of FPH and radiosonde H$_2$O below 8 km found before, here we use radiosonde data up to 8 km of altitude merged, in the absence of a more suitable dataset, with NIWA–UKCA output above that.

We use surface in situ measurements from Cape Grim, Tasmania (Cunnold et al., 2002) to rescale NIWA–UKCA model profiles, producing CH$_4$ profiles that coincide with the ozone sonde launches. The NIWA–UKCA model simulation had been constrained with historical global-mean surface CH$_4$ values, resulting in an overestimation relative to the Cape Grim data by $\sim 2\%$ (not shown), and both data show a $\sim 5\%$ increase in CH$_4$ at the surface over the period between 1994 and 2010. Cape Grim CH$_4$ is a good surrogate for the Lauder measurements because CH$_4$ is a long-lived, well-mixed atmospheric constituent.

The time series of CO profile over the period of 1994–2010 has been constructed using the NIWA–UKCA CO profiles, rescaled such that the total columns that obtained from the mid-infrared Fourier Transform Spectrometer (FTS) at Lauder (Rinsland et al., 1998; Zeng et al., 2012; Morgenstern et al., 2012). Gaps in the total-column FTS series, such as the period between 1994 and 1996 when the FTS measurements had not started yet, are filled using a regression fit accounting for the mean annual cycle (modelled as a 6-term harmonic series) and the linear trend.

The time series of temperature profiles are constructed following the same procedure as used in the construction of O$_3$ profiles, comprising the radiosonde temperature profiles (from the surface to 25 km) merged with NCEP/NCAR reanalyses (Kalnay et al., 1996) temperatures used in the retrieval of MOPITT ozone (above 25 km) for the period of 1994–2010. From near the stratopause upwards the NCEP/NCAR temperatures are merged with a mesospheric climatology based on local lidar measurements. There are some warm anomalies occurring in the data at 40–60 km during winter months (e.g. in 1996); these may reflect planetary wave breaking in the upper stratosphere.

## 2.3 Simulations

We perform SCM simulations covering the period of 1994–2010, summarized in Table 1. The forcing data needed by the SCM consist of profile series of temperature, pressure, optional cloud liquid and ice mass mixing ratios, and the mixing ratios of 86 chemical compounds. With the exceptions detailed below, these fields and species are taken from a NIWA–UKCA simulation for the period of 1994–2010 inter-
In this section, we present sensitivity studies to assess the contribution of biases in known factors (O3, H2O, CH4, CO, and temperature, or all of these simultaneously, are replaced with the time series of the profiles that are constructed using long-term observational data as described above. For species other than those five fields, in all cases we use NIWA–UKCA REF-C1 forcings. We evaluate the SCM only for those times, spaced roughly weekly, for which ozone sonde data are available. With the exceptions of those simulations assessing cloud influences, simulations are conducted assuming clear-sky conditions.

3 OH sensitivity to correcting chemistry–climate model biases

In this section, we present sensitivity studies to assess the contribution of biases in known factors (O3, H2O, CH4, CO, and temperature) affecting OH photochemistry at Lauder.

Figure 3. Multi-annual and monthly-mean OH responses to O3 biases between observations and the reference simulation. (a) Difference in O3 (%) between ozone sonde and NIWA–UKCA ozone, relative to NIWA–UKCA ozone as prescribed in the reference simulation. (b) OH difference (%) relative to the reference simulation accounting only for the kinetics effects of O3 differences (e.g. with jO1D unchanged). (c) Difference in jO1D (%) relative to the reference simulation. (d) OH difference (%) relative to the reference simulation accounting only for jO1D differences (e.g. with O3 unchanged). (e) OH differences relative to the reference simulation considering the combined kinetics and photolysis effects. (f) Sum of (b) and (d).
The response of OH to changes in each forcing is assessed individually and also in combination.

3.1 OH sensitivity to O3 biases

Three sensitivity simulations are conducted to quantify the impact of O3 biases (defined as differences between observed O3 and NIWA–UKCA simulated O3) on OH at Lauder.

As discussed above, the rate of production \( P \) of HOx via O(\(^1\)D) + H2O can be expressed as

\[
P(\text{HO}_x) \approx \frac{2k_1j_{\text{O}}(1)\text{[O}_3]\text{[H}_2\text{O]}}{k_2\text{[O}_2] + k_3\text{[N}_2] + k_4\text{[H}_2\text{O}]}.
\]

where \( k_1 \) is the rate coefficient for O(\(^1\)D) + H2O, \( j_{\text{O}}(1) \) is the rate of O3 photolysis producing O(\(^1\)D), and \( k_2 \) and \( k_3 \) are the rate coefficients of quenching of O(\(^1\)D) with O2 and N2, respectively (Liu and Trainer, 1988; Thompson et al., 1989; Madronich and Granier, 1992; Fuglestvedt et al., 1994). Accordingly, \( P(\text{HO}_x) \) is affected by ozone changes principally in two different ways: Either locally through a change in [O3] or non-locally through a change in \( j_{\text{O}}(1) \) caused by changes in the overhead total-column ozone (TCO). To separate the effects, we conduct three simulations with the SCM: The first simulation targets the local kinetics effect by applying changes in O3 concentrations but keeping all photolysis rates unchanged vs. the reference simulation. A second simulation involves applying changes in \( j_{\text{O}}(1) \) according to changes in O3 (keeping the rest of photolysis rates unchanged), but considering a fixed O3 concentration, i.e., using the O3 concentrations of the reference simulation. The \( j_{\text{O}}(1) \) calculation consistently takes into account absorption and scattering by stratospheric and tropospheric O3. A third simulation includes both effects simultaneously.

The results of these three sensitivity runs are displayed in Fig. 3. As expected, the pattern of O3 differences between observed O3 and modelled O3 (Fig. 3a) is reflected in the pattern of OH differences produced by the SCM, considering only the “kinetics” effect and assuming no changes in the photolysis rates (Fig. 3b), with increases of ozone in spring and decreases in autumn, relative to the reference simulation, resulting in changes of the same sign in OH. However, there is a height dependence to this relationship.

In summer and autumn, O3 biases range between −5 and −45%, meaning that the reference simulation overestimates the observations. Such a bias in O3 results in up to 12% reductions in OH for these seasons when the bias is corrected. In spring between 2 and 6 km, observed O3 is larger than in the reference simulation by up to 10% at 4 km in October. Consequently, this results in an increase of OH at around the same altitudes and times of up to 5%.

Regarding the sensitivity simulation considering the photolysis effects, \( j_{\text{O}}(1) \) exhibits differences relative to the reference simulation ranging from ∼ 14 to ∼ 30%. The corrections are positive everywhere, in accordance with the overestimation of TCO in the NIWA–UKCA model with respect to observations (Morgenstern et al., 2013; Stone et al., 2016). In accordance with eq. 1, such an intensification of \( j_{\text{O}}(1) \) causes OH to increase (Fig. 3c). The relative OH response is approximately 50% of the \( j_{\text{O}}(1) \) relative difference. However, Fig. 3c and d suggest that the magnitudes of the kinetics and the photolysis effects, for the O3 bias found at Lauder, are comparable, but the seasonalties differ. For example, the kinetics effect maximizes in spring at 5% and minimizes in summer/early autumn at −15% (in the upper troposphere) whereas the photolysis effect on OH maximizes in summer at 16 to 20% and minimizes in spring (Fig. 3b and d).

OH resulting from the combined kinetics and photolysis effects is displayed in Fig. 3e. OH responds approximately linearly to the two effects combined, compared to the sum of their individual impacts (Fig. 3f), despite some small differences between Fig. 3e and f.

Next, we examine the relationship between the slant column of O3 (SCO), \( j_{\text{O}}(1) \), and OH. Figure 4a shows that there is an approximately exponential relationship between \( j_{\text{O}}(1) \) and the SCO at 6 km of altitude (this effect also exists at other altitudes). The small curvature may be the result of inaccurately diagnosing the SCO (ignoring the curvature of the Earth). Another reason could be that the cross section of O3 is wavelength-dependent, and consequently the actinic flux spectrum moves towards longer wavelengths with increasing SCO. Under Lambert–Beer’s Law, a perfectly exponential relationship would be expected for a monochromatic UV light source and an isothermal atmosphere. \( j_{\text{O}}(1) \) and the OH concentration exhibit an approximately exponential relationship (Eq. 1, Fig. 4b). Combining these results, we derive an approximately exponential relationship between the SCO and the OH concentration (Fig. 4c). The fit parameters are stated in Fig. 4. Due to the compact relationship between \( j_{\text{O}}(1) \) and the SCO, and to account for the curvature, we fit a quadratic relationship between the SCO and log(\( j_{\text{O}}(1) \)).

To determine a simple coefficient that describes the quantitative contribution of O3 to OH, a linear regression between differences in OH and O3 relative to the reference was conducted through the following expression (note that this equation is also used to derive the linear contributions of the other key species to OH chemistry at Lauder):

\[
\frac{\Delta[\text{OH}]}{[\text{OH}]_{\text{ref}}} = A_i \frac{\Delta X_i}{X_{i, \text{ref}}},
\]

where \( X_i \) is the perturbation variable (in this case [O3]), \( A_i \) is the slope of the linear regression, \( \Delta[\text{OH}] \) is the absolute difference between the OH concentrations in the reference and perturbation simulations, \( \Delta X \) is the absolute difference in concentrations of the perturbation variable X between the observations and the reference, \([\text{OH}]_{\text{ref}}\) is the OH concentration obtained from the reference simulation, and \( X_{i, \text{ref}} \) is the value of \( X_i \) in the reference simulation. The regression coefficients \( A_i \) represent the sensitivity of OH to changes in each individual variable for the troposphere at Lauder. The regression coefficients are depicted in Fig. 5. Reverting to
The sensitivity coefficients of OH to the kinetics and photolysis effects of O₃ (SCO) at 6 km of altitude. (b) Same, but for j₀₁D and OH. (c) Same, but for the SCO and OH. The results shown in this figure are those obtained from the combined simulation (kinetics and photolysis effects). Red lines denote least-squares fits between the variable pairs. The best fits are stated in the panels, with [OH] in units of 10^6 molec cm⁻³, j₀₁D in units of 10⁻² s⁻¹, and the SCO in Dobson Units.

\[ A_i = \frac{\partial \ln[OH]}{\partial \ln X_i}. \]  

(3)

The sensitivity coefficients of OH to the kinetics and photolysis effects of O₃ are shown in Fig. 5a. Coefficient A₁, which represent the kinetics effect, ranges from 0 to 0.25 (meaning the relative response of OH is up to a quarter of the relative difference in O₃). The sensitivity to photolysis (A′ᵢ) is > 0.5 throughout much of the troposphere (meaning the relative response in OH is over half the relative change in j₀₁D).

3.2 OH sensitivity to H₂O biases

A perturbation simulation was performed using combined radiosonde and NIWA–UKCA H₂O (Sect. 2.2). The OH response to correcting H₂O biases (Fig. 6) shows an approximately linear response with respect to the relative changes in H₂O, i.e. decreases in H₂O generally lead to a reduction of OH concentrations (Eq. 1). Note that NIWA–UKCA substantially overestimates the radiosonde-observed H₂O vapour by up to 60 % between 2 and 6 km (Fig. 6a); this translates into an overestimation of OH by up to ~ 40 % in the reference simulation (Fig. 6c) in the same region. The sensitivity of OH to changes in H₂O (Eq. 2) range from 5 to 0.5 in the troposphere (Figs. 6e and 5b coefficient A₂), with high sensitivity in the lower and free troposphere and reduced sensitivity in the tropopause region.

It is known that large uncertainties are associated with H₂O vapour measurements. To illustrate this, we repeat the above simulation but now using European Centre for Medium-Range Weather Forecasts (ECMWF) ERA-Interim reanalysis (hereafter ERAI) H₂O (Dee et al., 2011). Irrespective of the large differences and the opposite signs in H₂O biases between Lauder radiosonde and ERAI data, the OH response to biases in H₂O shows approximately the same linear relationship in both cases (Fig. 6). Likewise, the sensitivity of OH to changes in H₂O using ERAI data (Figs. 6f and 5b, coefficient A₁) resembles the sensitivity simulation using radiosonde H₂O.

3.3 OH sensitivity to CH₄ and CO biases

The effect of CH₄ changes on OH is displayed in Fig. 7a, c, e. The CH₄ biases are generally small, up to only ~ 2 %, and are assumed to be vertically uniform, with some seasonal variations. Decreases in CH₄ lead to increases in OH due to reduced loss of HO₂ by CH₄ + OH. The response of OH to CH₄ changes maximizes at 0.6 % around 2 km, and decreases at higher altitudes. The seasonal variation of the OH response to CH₄ biases maximizes in March/April (Fig. 7c), which coincides with the maximum absolute bias in CH₄ (Fig. 7a) in the same months. The sensitivity coefficient describing the dependence of OH on CH₄ changes (denoted as A₂ in Fig. 5c) ranges from −0.17 at the surface to −0.32 at ~ 2 km of altitude, and then decreases to −0.15 at 10 km.

The CO bias and the resulting differences in OH are displayed in Fig. 7b, d, f. The relative difference of OH with respect to the reference simulation is less than ±5 % for all seasons (Fig. 7d), showing that decreases in CO generally lead to increases in OH through the reduced loss of OH through OH + CO. Note that during austral spring NIWA–UKCA overestimates CO, presumably due to exaggerated tropical
Figure 5. Sensitivity coefficients $A''_k$ between OH and each perturbation variable: In the calculation, multi-annual mean relative differences in OH and in the forcing are ratioed. (a) Sensitivity of OH to changes in $O_3$ levels (kinetics effect) denoted by $A_1$ (solid line) and to changes in $iO^1D$ due to changes in $O_3$ (photolysis effect) denoted by $A''_1$ (dashed line); (b) sensitivity of OH to changes in radiosonde – NIWA–UKCA CCM H$_2$O ($A_2$ solid line) and to changes in ERAI – NIWA–UKCA H$_2$O ($A_3$ dashed line); (c) sensitivity of OH to changes in CH$_4$ ($A_4$); (d) sensitivity of OH to changes in CO ($A_5$); (e) sensitivity of OH to changes in temperature (kinetics effect) denoted by $A_6$; (f) sensitivity of OH to changes in $iO^1D$ due to changes in temperature (photolysis effect) denoted by $A''_6$.

3.4 OH sensitivity to temperature biases

To assess the effects of changes in temperature on OH, we apply the same procedure as for $O_3$, for which the effects of temperature have been decomposed into kinetics and photolysis effects. We perform three simulations: In the first simulation, we only apply the same procedure as for $O_3$; for which the effects of temperature are weakly, temperature-dependent, making CH$_4$ + OH (which is much more sensitive to temperature) the leading factor in causing this small OH reduction. The rate coefficient for this reaction in NIWA–UKCA and the SCM is $k_{OH+CH_4} = 1.85 \times 10^{-12}\exp(-1690K/T)$; at 290 K the sensitivity of $k_{OH+CH_4}$ to temperature changes evaluates to 10% (Fig. 8b). The sensitivity of OH to changes in CO varies from $-0.3$ to $-0.5$ and in absolute terms increases with altitude (the white band shown in October is the result of CO differences being close to zero).

The sensitivities of OH to CH$_4$ and CO show comparable values at the surface, but the OH sensitivity to CO increases with height whereas its sensitivity to CH$_4$ decreases. Note that the CH$_4$ + OH reaction rate is strongly temperature-dependent, which may contribute to the lower sensitivity of OH to CH$_4$ changes at altitude than to CO. However, further investigation will need to investigate how these ratios change in different chemical regimes, and to assess whether the relative sensitivity of OH to CO and to CH$_4$ are specific to the clean SH environment.

At Lauder, the reference simulation is generally cold-biased (i.e. the temperature correction is positive; Fig. 8a). This is particularly the case in the lowest 2 km and throughout the troposphere in the autumn–winter season. The kinetic effect leads to a reduction of OH by up to 2% (Fig. 8b).
about 2% K⁻¹. However, OH is well buffered by other reactions, so its sensitivity is considerably smaller than that. The photolysis effect is often somewhat larger than the kinetics effect but peaks in spring (Fig. 8c). This translates into a slight OH reduction comparable in magnitude to the kinetics effect (Fig. 8d). Both effects add nearly linearly in the combined simulation (Fig. 8e, f).

We calculate sensitivity coefficients $A_k$ and $A_{\text{OH}}$ that define the OH responses to both effects (Fig. 5e, f). Coefficient $A_k$ represents the kinetics effect and varies from 0 to −1.75 (i.e. in absolute terms, the relative OH response can be larger than the relative difference in $T$). The sensitivity coefficient that describes the sensitivity of OH to changes in photolysis ($A_{\text{OH}}$) ranges from 0.6 at the surface to 0 at 10 km of altitude. Figure 5e and f show sensitivity coefficients for both effects ($A_k$ and $A_{\text{OH}}$). OH changes due to both effects are small (up to 2.5%) and comparable in magnitude.

Several sensitivity studies have been conducted previously to elucidate the impact of temperature on OH (Stevenson et al., 2000; Wild, 2007; O’Connor et al., 2009). None of these studies separately assessed the impacts of the kinetics and photolysis effects of temperature on OH. Wild (2007) applied a globally uniform temperature rise of 5 K that led to a larger OH abundance and an around 10% decrease in the CH₄ lifetime. O’Connor et al. (2009) showed a small impact on global OH abundances due to temperature biases; this may be because either the temperature biases in their model were both positive and negative, in different regions, leading to some cancellation of the impact on global OH, or to low OH sensitivity to temperature biases. Here, bias-correcting temperature is shown to also have only a small impact on OH abundance (Fig. 8e); this result broadly corroborates that of O’Connor et al. (2009).
3.5 Linearity of OH sensitivity to biases in all forcings

Here, we assess the effect of changing all forcings (O$_3$, H$_2$O, CH$_4$, CO, and temperature) simultaneously on OH at Lauder. Figure 9a shows the responses of OH to changing all forcings. A comparison with Fig. 6 suggests that H$_2$O changes dominate the total response of OH to changes in these forcings. At Lauder, NIWA–UKCA is too moist (relative to radiosonde water vapour); this translates into a large OH overestimation of up to $\sim$ 40% in the reference simulation (Fig. 9a). This is consistent with the underestimated CH$_4$ lifetime by the NIWA–UKCA model (Morgenstern et al., 2013; Telford et al., 2013), assuming that the NIWA–UKCA model is also too moist in other regions. (In the NIWA–UKCA reference simulation used here, the global CH$_4$ lifetime, disregarding dry deposition, is 7.2 years, whereas a recent best estimate is 9.8 years, with an uncertainty range of 7.6–14 years; SPARC, 2013.) In general, in the SCM OH responds approximately linearly to the combined changes in major forcings that play an important role in OH chemistry (Fig. 9).

To examine the linearity of OH responses to simultaneous changes in key forcings defined in this study, the combination of all individual contributions, i.e. O$_3$ (kinetics and photolysis effects), H$_2$O, CH$_4$, CO, and temperature (kinetics and photolysis effects) to OH, was compared to the OH response to all forcings combined in the SCM through Eq. (4):

$$\frac{\Delta[OH]}{[OH]_{ref}} \approx A'' \frac{\Delta [O_3]}{[O_3]_{ref}} + A' \frac{\Delta J_0^O D_2}{J_0^{O, D_2}_{ref}} + A_2 \frac{\Delta [H_2O]}{[H_2O]_{ref}} + A_4 \frac{\Delta [CH_4]}{[CH_4]_{ref}} + A_5 \frac{\Delta [CO]}{[CO]_{ref}} + A_6 \frac{\Delta T}{T_{ref}} + A'_6 \frac{\Delta J_0^O D_T}{J_0^{O, D_T}_{ref}}, \tag{4}$$
where $\Delta[OH]/[OH]_{ref}$ is the relative difference in the OH concentration obtained with the SCM with respect to the reference simulation, using all forcings combined. The forcings comprise the kinetics and photolysis effects of O$_3$ ($A_1$ and $A_1''$), radiosonde H$_2$O ($A_2$), CH$_4$ ($A_4$), CO ($A_5$), and the kinetics and photolysis effects of temperature ($A_6$ and $A_6''$). Equation (4) expresses a working hypothesis that the model responds linearly to the applied forcings; we will test this assumption in the following paragraph.

Figure 9a and b indicate that the model responds approximately linearly to the combinations of all forcings, with OH responses in the all-forcings simulation correlating at 0.9 with the sum of the OH responses in the individual-forcing simulations. Figure 9c however also suggests that there is some notable non-linearity in the chemistry of the troposphere at Lauder. Chemical feedbacks between the impacts of correcting water vapour and ozone may contribute to this non-linearity; for example, a change in the water vapour abundance may impact the sensitivity of OH to changing O$_3$.

3.6 Trends in OH

We examine variability and trends in OH using the SCM simulation including all key forcings separately for different altitude bins. The results (Fig. 10) indicate that there are no significant long-term trends in OH throughout the troposphere for the period of the simulation (1994–2010) We find trends of $-2.1 \pm 4.8\%$ at 0–2.5 km, $0.9 \pm 2.3\%$ at 2.5–5 km, $2.6 \pm 3.5\%$ at 5–7.5 km, and $3.6 \pm 4.1\%$ at 7.5–10 km over the period of 1994–2010), but there is evidence of interannual variations at all altitudes (e.g. Manning et al., 2005; Montzka et al., 2011).

In addition, we explore variability and trends in the OH column at Lauder to be compared with other estimates of OH...
Figure 9. (a) Multi-annual and monthly-mean percentage difference in OH between a simulation with bias correction applied to all five fields and the reference simulation. Radiosonde H$_2$O is assumed below 8 km. (b) Summation of all the single forcing contributions as expressed by the right hand side of Eq. (4). Radiosonde H$_2$O is assumed below 8 km. (c) Scatter plot of the response of OH to the combination of all forcings (vertical axis, denoted as $\Delta[OH]$) vs. the summation of the OH response to individual forcings (horizontal axis) as expressed by the right hand side of Eq. (4) (denoted by $\sum_i \Delta[OH]_i$). The red solid line denotes an orthogonal fit. The black dashed line is the diagonal.

global OH. As expected from the results of OH trends at different altitude bins, we find no significant long-term trend in the OH column (0.5 ± 1.3 %) (Fig. 11). However, there is evidence of short-term variations (5–10 %), in agreement with other studies that used observations to infer global OH concentrations (e.g. Manning et al., 2005; Montzka et al., 2011).

3.7 OH sensitivity to the presence of clouds

We have assessed the OH sensitivity to correcting biases in key forcings assuming clear skies. Here we explore the impact of simulated clouds on OH, recognizing that this process is associated with large uncertainties due to difficulties with representing clouds in models. Measurements of cloud profiles do not exist at Lauder, hence a bias correction like that performed with the composition and temperature fields is not possible. Therefore, here we only examine the impact of clouds simulated by NIWA–UKCA on $j_{O1D}$ and OH at Lauder, relative to the clear-sky reference simulation used before. The impacts of liquid water clouds (LWCs) and ice clouds (ICs) were assessed separately and in combination. Three simulations are defined here, i.e. (1) including only ICs, (2) including only LWCs, and (3) considering both combined (ICs).

Figure 12a, c and e show the response of $j_{O1D}$ and OH to the presence of the ICs. $j_{O1D}$ and OH are generally reduced below the ICs, relative to the cloud-free situation. The maximum reduction in OH is 10 to 15 % in winter below 2 km, coinciding with the maximum reduction in $j_{O1D}$. There are increases in both fields (up to ∼8 %) above the ICs in austral spring, associated with the seasonal peak in IC occurrence at the same time. In general, $j_{O1D}$ and OH impacts vary strongly with season, with the maximum reduction occurring in winter close to the surface, and the maximum increase in spring above the ICs.

LWCs are mostly present between 1 and 4 km with the seasonal peak in austral spring (Fig. 12b). Similarly to ICs, $j_{O1D}$ and OH are enhanced above and throughout much of the cloud layer, and reduced in the lowest 1 km above the surface (Fig. 12e, g). The enhancement in $j_{O1D}$ and OH peaks at 12 % between 2 and 4 km of altitude, coinciding with the spring maximum in liquid water content at 1–2 km. Conversely, the reduction in $j_{O1D}$ and OH with respect to the clear-sky condition is ∼10 % and is produced below the clouds.

The simulation with the combined effect of ICs and LWCs (LICs) produces a reduction in $j_{O1D}$ and OH that ranges between 0 and 20 % below the transition of ICs to LWCs at around 2 km, since LWCs are as much as twice as optically dense as ICs (Fig. 12g). An enhancement is produced above this altitude of up to 18 %. The magnitudes of changes in $j_{O1D}$ and OH are similar when either ICs or LWCs are considered in the SCM. Furthermore, their effects add up slightly less than linearly when both are present in the simulations (Fig. 12h).

The results shown here indicate that lower clouds generally produce an enhancement in $j_{O1D}$ (Fig. 12d), but higher clouds generally produce a reduction in $j_{O1D}$ in the free troposphere (Fig. 12b; Tang et al., 2003; Tie et al., 2003; Liu et al., 2009). Furthermore, the vertically and seasonally av-
eraged enhancement and reduction in $j_{O(1D)}$ are about 2 and 6% respectively for the LWCs, similar to the response for the ICs’ condition; this suggests that the cloud vertical distribution has a bigger effect on photolysis than the change in cloud water content (Tie et al., 2003).

Figure 10. Variability and trends of the OH anomalies at different altitudes: (a) 0–2.5, (b) 2.5–5, (c) 5–7.5, and (d) 7.5–10 km. The red solid line is the time series of the reference simulation and the blue solid line is the combined forcings simulation considering radiosonde – NIWA–UKCA H$_2$O.

Figure 11. Variability and trend of the OH column anomaly. The red solid line is the time series of the reference simulation and the blue solid line is the combined forcings simulation considering radiosonde – NIWA–UKCA H$_2$O.

4 Conclusions

The sensitivity of the OH abundance at Lauder to NIWA–UKCA model biases in key forcing variables (O$_3$, H$_2$O, CH$_4$, CO, and temperature) have been quantified for clear-
Figure 12. Multi-annual and monthly-mean OH responses to the presence of clouds. Multi-annual and monthly mean (a) ice content (10^{-5} \text{kg kg}^{-1}), (b) liquid water content (10^{-5} \text{kg kg}^{-1}). (c) Response of j(O^1D) (%) to the presence of ICs relative to the cloud-free reference simulation. (d) Response of j(O^1D) (%) to the presence of LWCs relative to the cloud-free reference simulation. (e) Response of OH (%) to the presence of ICs relative to the cloud-free reference simulation. (f) Response of OH (%) to the presence of LWCs relative to the cloud-free reference simulation. (g) Response of OH (%) to the presence of both LWCs and ICs relative to the cloud-free reference simulation. (h) Sum of (e) and (f).

sky conditions, using a single-column model. Only fast photochemistry is represented in the SCM; slow chemistry (i.e. timescales similar to or longer than the 1 h chemical timestep), transport, and other physical processes are thus not considered. The bias-corrected profiles of the key forcing variables have been constructed largely using long-term Lauder measurements, combined with NIWA–UKCA output. A few other sources of data (Cape Grim methane measurements, ERA-Interim water vapour) have also been used.

The results show that OH responds approximately linearly to correcting biases in O_3, H_2O, CH_4, CO, and temperature. We have decomposed the OH response to O_3 changes into the kinetic effect (i.e. local impacts on the chemical steady state of changing [O_3]) and the photolysis effect (as mediated...
by changes in the overhead O$_3$ column affecting photolysis rates). We find that the kinetic effect of correcting positive biases in modelled O$_3$ causes a reduction in OH during austral summer and autumn (by up to 20% at 7 km), and an increase in the free troposphere in austral spring (of > 5% in October at 3 km); such changes in OH are nearly linearly related to the corresponding ozone biases. NIWA–UKCA generally overestimates the ozone column. Correcting this bias causes $j_{O1D}$ to increase by 15–30% below 10 km, causing general OH increases which maximize at around 16% between 2 and 6 km in summer. The model responds approximately linearly to the combined effects of photolysis and kinetics.

NIWA–UKCA considerably overestimates the H$_2$O vapour concentration by up to ∼50% compared to radiosonde measurements. Correcting this moist bias leads to > 34% reductions in OH in the free troposphere during the austral summer. The sensitivity coefficient of OH to biases in H$_2$O vapour is relatively large in the lower troposphere but decreases with altitude. Assuming this moist bias is not restricted to Lauder (which we do not assess here), this is thus a leading explanation for NIWA–UKCA to produce an underestimated CH$_4$ lifetime (Morgenstern et al., 2013; Telford et al., 2013), relative to literature estimates (Naik et al., 2013; Voulgarakis et al., 2013; SPARC, 2013).

The bias in modelled CH$_4$ is small since surface CH$_4$ in the SCM reference simulation is constrained to follow globally averaged surface observations. The Southern Hemisphere generally has a slightly smaller CH$_4$ burden than the Northern Hemisphere. Correcting the resulting positive bias at Lauder causes increases in OH throughout the troposphere, with a seasonal peak in March/April. OH is most sensitive to CH$_4$ changes in winter, though. In the analysis of the OH sensitivity to CH$_4$, the impact of subsequent changes in CH$_4$ oxidation products which also affect OH could not be addressed within the constraints of an SCM. Inclusion of this effect could change the sensitivity coefficient for CH$_4$ (Spivakovski et al., 2000).

Except for October–December, NIWA–UKCA has a tendency to underestimate CO. As with CH$_4$, the sensitivity of OH to changes in CO is negative throughout the troposphere, reflecting that CO + OH is an important sink for OH.

We show that OH responds linearly to temperature biases. These effects cause a reduction in OH due to the strong dependence of OH + CH$_4$ on temperature (Eq. 1). However, the impact of this reaction on OH is buffered by other less temperature-dependent reactions, causing only a small sensitivity of OH to temperature. This is in agreement with O’Connor et al. (2009).

The results of the simulation considering simultaneous changes in all the key forcings indicate that OH responds approximately linearly to all the major forcings that contribute to the oxidizing capacity of the atmosphere. We find that biases in O$_3$, H$_2$O, CH$_4$, CO, and temperature all affect the oxidizing capacity of the atmosphere at Lauder, with H$_2$O and O$_3$ biases dominating. We find no significant trend in OH over Lauder over the period 1994–2010.

The SCM approach can be applied to other parts of the globe where reliable long-term observations of O$_3$ and H$_2$O exist. In situ observations of CH$_4$ and CO are not that critical; CH$_4$ can be estimated from non-local measurements, and relatively reliable satellite measurements of total-column CO exist (e.g. Pan et al., 1995; Morgenstern et al., 2012). However, in polluted regions, such as in much of the Northern Hemisphere, NO$_x$ and NMVOC levels are elevated relative to Lauder and affect in situ ozone production. This means that these constituents might need to be bias-corrected if the SCM is applied in such regions. This might affect the suitability of our approach under these conditions.

Having determined the contributions of the major forcings to the chemistry of OH at Lauder under clear-sky conditions, a step forward would be to assess the impact of clouds on photolysis and thus OH, which could be substantial. Due to a lack of suitable observations to constrain the SCM model with cloud profiles at Lauder, we only assessed how the presence of modelled cloud affects OH, relative to the clear-sky situation. The results show that OH response to cloud strongly depends on the vertical distribution of the clouds, not just the total amount. Both liquid and ice clouds lead to increases in OH above and to some extent inside the cloud, particularly in the spring season when this effect maximizes. Considering that clouds are amongst the most difficult aspects of the climate system to model adequately, we stipulate that observational profiles of cloud properties would be highly desirable to use for a future continuation of this line of research.

In summary, we conclude that at Lauder, OH modelled in NIWA–UKCA is most sensitive to issues with representing water vapour and ozone. This points to the need to improve representations of the hydrological cycle and of tropospheric and stratospheric ozone chemistry in NIWA–UKCA and possibly other, similar chemistry–climate models. Water vapour is coupled to clouds in NIWA–UKCA; it is well known that clouds are difficult to represent adequately in global low-resolution climate models. The biases in ozone may well be partly caused by the moist bias in NIWA–UKCA; this is a subject of ongoing research.

Progress with the simulation of the hydrological cycle in present-generation Earth system models should improve the simulated water vapour product. Simulating an accurate hydrological cycle has been a long-standing issue in climate models, and progress has been slow. If errors in the simulation of moisture cannot be avoided, perhaps their impact on OH can be corrected for using an approach similar to that which we have presented but using global water vapour measurements. Such a “correction” of modelled OH might result in a reduction in the inter-model spread of the OH abundance and consequently a more accurate quantification of the methane lifetime. For this, tropical radiosonde data would be particularly valuable – most OH is located in the tropics.
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