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Key Points:

- The ozone-mediated impact of GHGs on the summer SAM offsets the direct effect
- In climate simulations prescribed ozone needs to be consistent with GHG forcing

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Direct and ozone-mediated forcing of the Southern Annular Mode by greenhouse gases

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Abstract We assess the roles of long-lived greenhouse gases and ozone depletion in driving meridional surface pressure gradients in the southern extratropics; these gradients are a defining feature of the Southern Annular Mode. Stratospheric ozone depletion is thought to have caused a strengthening of this mode during summer, with increasing long-lived greenhouse gases playing a secondary role. Using a coupled atmosphere–ocean chemistry–climate model, we show that there is cancellation between the direct, radiative effect of increasing greenhouse gases by the also substantial indirect—chemical and dynamical—feedbacks that greenhouse gases have via their impact on ozone. This sensitivity of the mode to greenhouse gas-induced ozone changes suggests that a consistent implementation of ozone changes due to long-lived greenhouse gases in climate models benefits the simulation of this important aspect of Southern Hemisphere climate.

1. Introduction

Variations in the Southern Annular Mode (SAM), as defined by the hemispheric pressure gradient between polar and extrapolar regions, affect the strength and position of the midlatitude wind maximum (the “Roaring Forties”), and temperature and precipitation fields throughout the southern extratropics. A considerable increase in the frequency of occurrence of the positive phase of the SAM and associated blocking events [O’Kane *et al.*, 2013] during summer (December to February), observed during recent decades, thus means increased wind speeds over the Southern Ocean [Thompson *et al.*, 2011; Abram *et al.*, 2014; World Meteorological Organization (WMO), 2014]. In a mechanism understood to involve deep coupling of the stratosphere and troposphere, stratospheric ozone changes have influenced Southern Hemisphere climate, from the tropics to the pole, causing an expansion of the tropical Hadley Cell and southward shifts of climate regimes [Kang *et al.*, 2011]. For the remainder of this century, models project some cancellation of these impacts of ozone depletion, set to reduce under ozone recovery, with those of further increases in long-lived greenhouse gases (GHGs), chiefly CO₂, CH₄, and N₂O [Intergovernmental Panel on Climate Change (IPCC), 2013; Zheng *et al.*, 2013]. However, these GHGs do not just affect climate by trapping heat and influencing the equator-to-pole temperature gradient; they also have diverse impacts on ozone and its recovery. Increases in any of them, particularly CO₂, cool the stratosphere. This slows down gas phase ozone-depleting reactions and generally leads to an acceleration of stratospheric overturning (the “Brewer-Dobson Circulation”), enhancing poleward transport of ozone [WMO, 2011, 2014]. Within the polar vortices, the cooling increases the prevalence of polar stratospheric clouds [Roscoe and Lee, 2001], which under contemporary, high-chlorine conditions causes more polar ozone depletion. In combination, the GHG-induced cooling and accelerated overturning cause ozone to increase in midlatitudes and to decrease in the tropics and seasonally over the poles [Denman *et al.*, 2007]. In addition to these dynamical and temperature effects, CH₄ and N₂O also affect chemistry. Increases in CH₄ reduce the rate of chlorine-catalyzed ozone loss by promoting the return of chlorine to its reservoir form, hydrogen chloride (HCl), and also directly cause increased ozone production in the stratosphere in a catalytic process involving nitrogen oxides. N₂O increases, by contrast, induce ozone depletion by enhancing nitrogen-catalyzed ozone loss reactions [Revell *et al.*, 2012]. Anthropogenic emissions of N₂O now cause more ozone depletion than those of any of the ozone-depleting substances (ODSs) regulated under the Montreal Protocol [Ravishankara *et al.*, 2009]. Despite these known chemical and dynamical effects

Table 1. NIWA-UKCA Simulations Used Here^a

Experiment	REF-C2	SEN-C2-fGHG	SEN-C2-fODS	fOZONE	NoInder
Description	all forcings	fixed GHGs	fixed ODSs	fixed ozone	No GHG effect on ozone
No. of simulations	5	1	2	5	5
Ozone	Interactive	Interactive	Interactive	1960-1969 REF-C2 mean	SEN-C2-fGHG smoothed
Non-ODS GHGs	RCP 6.0	1960 abundances	RCP 6.0	RCP 6.0	RCP 6.0
ODSs	WMO [2011]	WMO [2011]	1960 abundances	1960 abundances	WMO [2011]
Effects present	OD, dGHG, iGHG	OD	dGHG, iGHG	dGHG	OD, dGHG

^aThe first three experiments are as defined for the Chemistry-Climate Modelling Initiative (CCMI) [Eyring et al., 2013b]. RCP = Representative Concentration Pathway. OD = anthropogenic ozone depletion. dGHG = direct radiative effect of changing GHGs. iGHG = indirect, ozone-mediated effect of changing GHGs. Differencing REF-C2 and NoInder, or SEN-C2-fODS and fOZONE, quantifies the iGHG effect. Note that all ozone climatologies used here are zonally variable.

of the leading GHGs onto ozone, only a minority of models participating in the Fifth Climate Model Intercomparison Project (CMIP5) calculated ozone interactively [Eyring et al., 2013a], partly because including interactive stratospheric ozone chemistry has not been clearly shown to benefit the quality of climate simulations for nonchemical model fields [Son et al., 2010], and because explicit ozone chemistry is a computationally expensive addition to climate models.

Here we investigate the roles of ozone depletion and long-lived GHGs in driving the SAM, differentiating between the above-discussed direct and indirect, ozone-mediated effects of GHG changes. For this we use a climate model, alternatively using interactive or prescribed ozone. The model (National Institute of Water and Atmospheric Research-United Kingdom Chemistry and Aerosols, “NIWA-UKCA”) is an atmosphere-ocean model, based on an early version of the HadGEM3-AO climate model, optionally including an explicit stratosphere-troposphere chemistry scheme [Morgenstern et al., 2009, 2013]. Apart from the interactive chemistry, the model is similar to the version described by Hewitt et al. [2011] but at a lower resolution in both the atmosphere (3.75° × 2.5°, 60 levels to 84 km) and the ocean (~ 2° × 1°, 31 levels). Relative to the model used by Nowack et al. [2014], there are only some differences in chemistry.

2. Experiments

Table 1 summarizes the experiments used here. The simulations all cover 1960–2100. The experiments using interactive ozone (REF-C2: all forcings; SEN-C2-fODS: ODSs fixed to their 1960 abundances; and SEN-C2-fGHG: GHGs fixed to their 1960 abundances) comply with corresponding definitions of the Chemistry-Climate Model Initiative (CCMI; <http://www.met.reading.ac.uk/ccmi>) [Eyring et al., 2013b]. These simulations all use the same surface emissions of tropospheric ozone and aerosol precursors, and do not consider volcanic eruptions and solar variability.

In the SEN-C2-fODS simulations, where ODSs are held constant, we expect trends in ozone due to only the chemical and dynamical influences of the changing long-lived GHGs as discussed above, which may drive trends in the SAM. In the SEN-C2-fGHG simulation, where the long-lived GHGs are held constant, we expect only ozone depletion to influence the SAM. fOZONE uses a prescribed, annually periodic ozone field generated by ensemble- and time-averaging monthly mean ozone fields from the first 10 years of the five REF-C2 simulations. As such it will not contain changes in ozone, and any trend in the SAM can be attributed to only the radiative effects of the GHGs. NoInder uses a prescribed ozone field generated by smoothing in time the monthly mean ozone field produced by SEN-C2-fGHG, in order to reduce the interannual variability in the field while retaining the annual cycle and the decadal-scale influence of ozone depletion. Thus, in this experiment, trends in the SAM can be caused by ozone depletion or the direct radiative forcing of greenhouse gases, but the indirect chemical and dynamical effects of changing greenhouse gases on ozone are absent. This suite of five experiments allows us to isolate the contributions not only of just changing GHGs or just changing ODSs but also the benefits of including interactive chemistry. Hence, both the direct and indirect (ozone-mediated) effects of GHG changes are present in the REF-C2 and SEN-C2-fODS experiments, but the indirect effect is suppressed in NoInder and fOZONE. Therefore, differencing the REF-C2 and NoInder experiments isolates this indirect effect.

Figure 1 shows effective equivalent stratospheric chlorine (EESC; a measure of the combined effect of chlorine and bromine onto ozone) [Daniel et al., 2007] in the ODS scenario considered here [WMO, 2011]

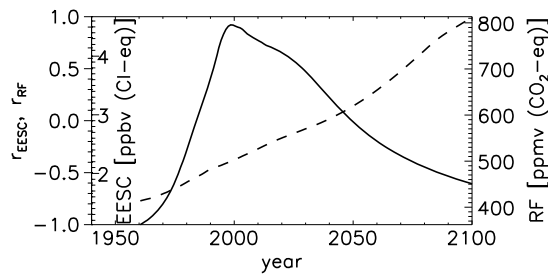


Figure 1. Regression functions used in the attribution analysis. Solid: EESC [Daniel et al., 2007] indicating an increase from 1.1 to 4.5 parts per billion by volume (ppbv) of chlorine, followed by a decrease during the 21st century to 1.8 ppbv. Dashed: Combined CO₂-equivalent RF due to all long-lived GHGs in parts per million by volume (ppmv) of CO₂, indicating a doubling during the course of the simulations. The nondimensional scale (left) is used in the regression analysis for both regression functions. On this scale, both regression functions satisfy $\bar{r} = 0$ and $\max |r| = 1$.

and total CO₂-equivalent radiative forcing (RF; a measure of the combined effect of anthropogenic greenhouse gases onto climate) [Forster et al., 2007] in the Representative Concentration Pathway (RCP) 6.0 [Meinshausen et al., 2011]. These forcing functions are used in this analysis to represent in a simplified way the anthropogenic forcings (ozone depletion and long-lived GHGs) applied in the model simulations. A regression analysis, detailed below, using these functions as basis functions can isolate the relative contribution of each to any trend in the SAM.

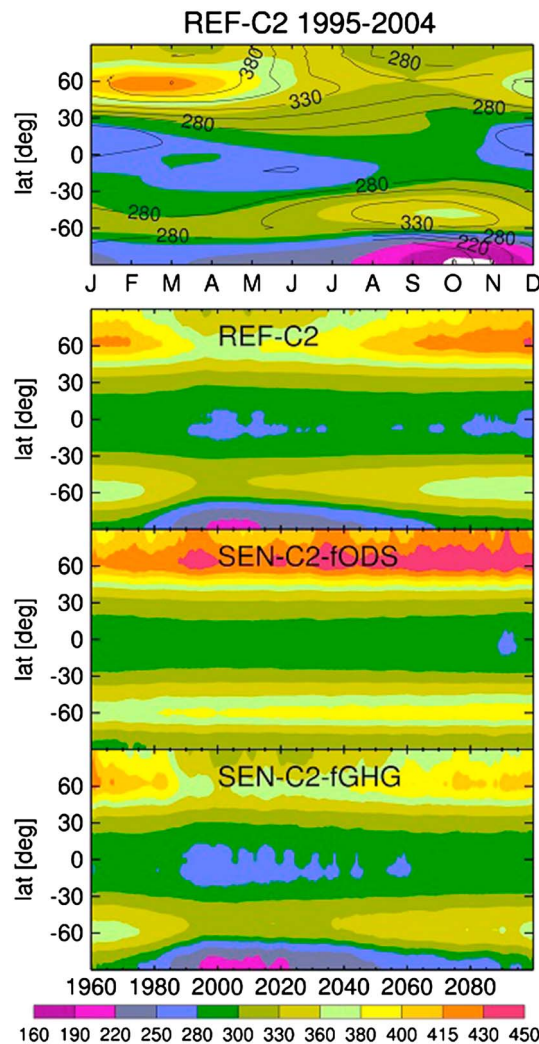


Figure 2. Zonal mean total-column ozone in the NIWA-UKCA simulations, in Dobson Units (DU). (first panel) Mean annual cycle for 1995–2004, from the REF-C2 ensemble average. Contours: NIWA-assimilated satellite data [Bodeker et al., 2005]. (second panel) REF-C2 annual mean. (third panel) SEN-C2-fODS annual mean. (fourth panel) SEN-C2-fGHG annual mean.

3. Results

An inspection of total-column ozone simulated by NIWA-UKCA shows that basically, the model produces a good representation of total-column ozone, with an annual cycle similar to observations (Figure 2). A small positive bias is within the usual range characterizing chemistry-climate models [Austin et al., 2010]. The model produces ozone decline and subsequent recovery, caused by anthropogenic ozone depletion, some ozone superrecovery (i.e., late 21st century ozone abundances exceeding their 1960s values) in the Northern Hemisphere, caused by the impacts of increasing GHGs, and more intense ozone depletion in the fixed-GHG simulation (SEN-C2-fGHG) than in the reference experiment (REF-C2), suggesting that the model captures the partial cancelation of ozone depletion with the impact of increasing GHGs discussed above.

Next, we assess surface pressure in these simulations. The SAM is an approximately ring-shaped feature with (for a positive SAM index) negative pressure anomalies over the South Pole and positive ones in a belt around midlatitudes [e.g., Gillett and Thompson, 2003]. Hence, here we define the SAM index as the difference in seasonal mean, area mean, and ensemble mean surface pressure between midlatitudes

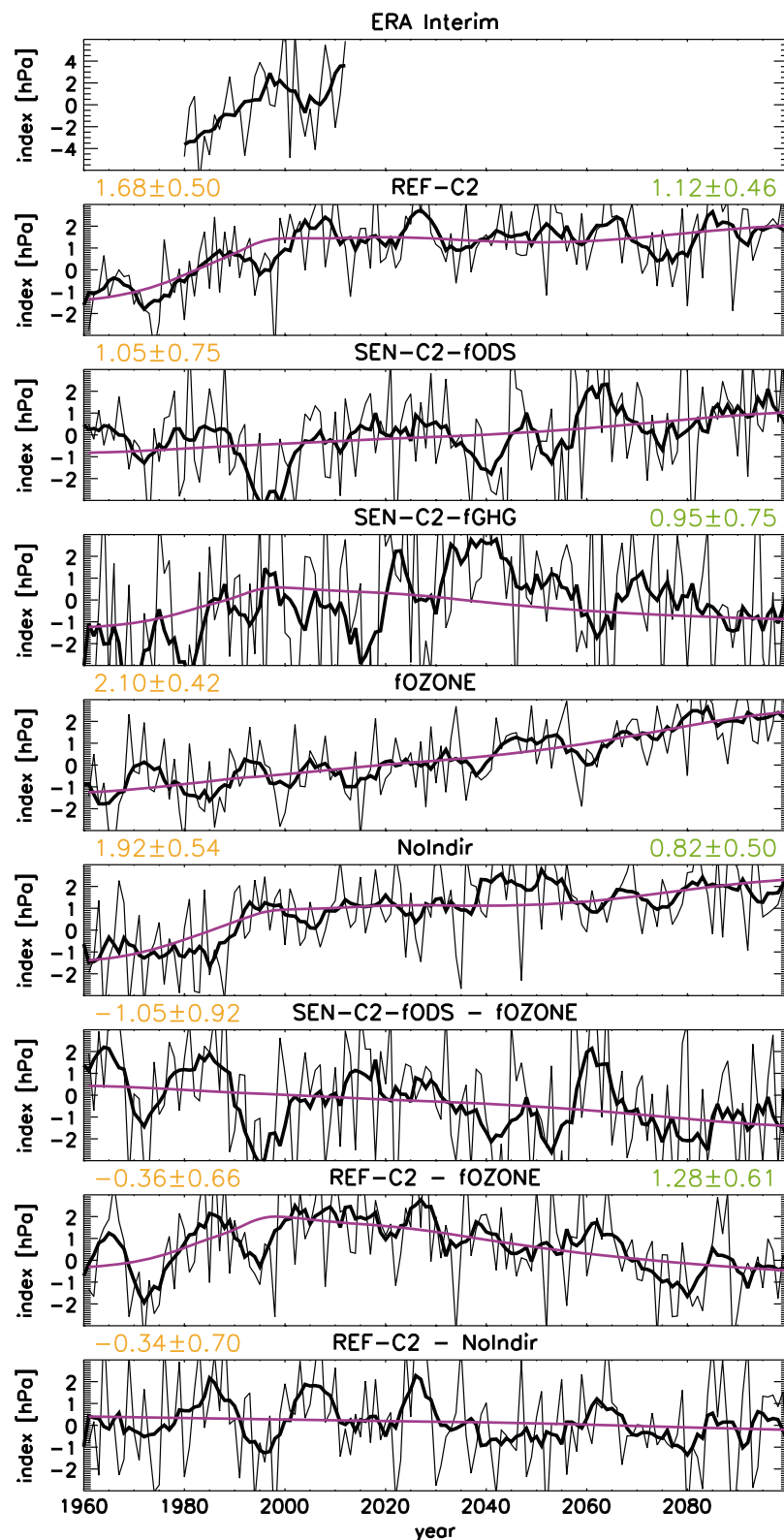


Figure 3. Seasonal SAM indices (hPa) for December–February. Ensemble-average (thin black). Ensemble-average (thick black), smoothed with 7 year boxcar filter. Least squares regression fits using the EESC and RF explanatory functions (violet). Regression coefficients against the nondimensional radiative-forcing regression function (hPa; Figure 1) (orange). Same but for the EESC regression function (green). Uncertainties refer to the 95% confidence level. Note the different scale applied to the ERA-Interim panel.

(38.75°S–61.25°S) and polar latitudes (63.75°S–90°S), minus the 1979–2012 mean of this quantity. This simple definition avoids complications to do with climatological drifts which affect other definitions based on empirical orthogonal function analysis [Morgenstern *et al.*, 2010].

Figure 3 shows the resulting SAM indices for the summer season. Similar to observations, for the period of 1960–2000, the REF-C2 ensemble mean SAM index increases by about 3 hPa during this period. Relative to the ERA-Interim meteorological reanalyses [Dee *et al.*, 2011], the magnitude of the trend is likely underestimated. For the period after 2000, the REF-C2 simulations indicate little trend during summer, consistent with other modeling experiments that show some cancelation of the impacts of ozone recovery and increasing GHGs [e.g., Gillett and Fyfe, 2013]. If underestimation of the 1960–2000 trend in REF-C2 is not a coincidence—our ensemble exhibits some considerable variability of this trend—it is consistent with an imperfect simulation of the SAM in our model. This might be the consequence of a cloud radiative forcing bias over the Southern Ocean which is common in climate models [IPCC, 2013, and references therein]. A further investigation of this aspect is however beyond the scope of this paper.

We now perform a multiple linear regression analysis on the SAM indices, using as basis functions the large-scale forcing functions shown in Figure 1. Details of the analysis are in Appendix A. The resulting least squares regression fits are depicted as violet lines in figure 3. First, we note that in the experiments subject to both global warming and ozone depletion, both influences are of comparable magnitude (e.g., 1.68 ± 0.5 and 1.12 ± 0.46 hPa for REF-C2; the regression coefficients are given in Figure 3). In the three experiments subject to ozone depletion (REF-C2, SEN-C2-fGHG, and NoIndir), the regression coefficients representing ozone depletion (given in green) are consistent with each other (namely 1.12 ± 0.46 , 0.95 ± 0.75 , and 0.82 ± 0.5 hPa), suggesting a reproducible trend in the SAM due to ODSs. The fixed-ozone (fOZONE) experiment produces a significant strengthening of the SAM (i. e., at 2.1 ± 0.42 hPa a positive regression coefficient with RF), which is significantly stronger than 1.05 ± 0.75 hPa which is found for the fixed-ODS (SEN-C2-fODS) experiment. Together, these two experiments suggest that the indirect, ozone-mediated influence of increasing GHGs onto the summer SAM offsets around half of the direct, radiative-forcing effect of increasing GHGs. (Note that the direct effect on the SAM of GHGs is present in both experiments, but their ozone-mediated, indirect effect is only present in SEN-C2-fODS, see Table 1). Hence, as a best estimate based on this analysis, we postulate that there is a 50% cancelation of the direct effect of GHG changes by the impact of ozone changes in SEN-C2-fODS.

We test this hypothesis using the other simulations. To assess whether the influence of the long-lived GHGs on the SAM is demonstrably larger in NoIndir than in REF-C2, we perform the regression on the difference in the SAM index between these two experiments. At -0.34 ± 0.70 hPa, with the uncertainty interval reflecting the 95% confidence limits, the resulting regression coefficient is less than 0 at about 83% confidence. We obtain numerically almost the same result if we replace NoIndir with fOZONE in this analysis (-0.36 ± 0.66 hPa). Both results suggest that about 20% of the direct effect of increasing GHGs onto the SAM (as inferred from the fOZONE experiment) is offset by their impact on ozone. While this appears to be a smaller effect than the above-derived half cancelation, both results are within each others' respective 95% confidence intervals. There is considerable variability within each of the different ensembles for the regression coefficients, which makes it impossible to attribute this discrepancy to anything other than random variations.

4. Discussion

The analysis suggests that all three influences on the SAM in REF-C2 appreciably contribute toward its long-term trend, and that there is a sizeable cancelation between the direct and indirect effects of increasing GHGs onto the SAM. A caveat pertaining to this analysis is that this is just a single-model study which should be expanded into a multimodel analysis. Some of the simulations used here form part of a multimodel intercomparison (CCMI) which will in time facilitate such an analysis. The magnitude of the effect is scenario dependent; we have only studied one scenario [Meinshausen *et al.*, 2011; WMO, 2011] here.

However, if the above results are found to be robust, they would have important implications for large climate-model intercomparisons such as CMIP5 which informed IPCC 2013. There, the majority of simulations used prescribed ozone climatologies [Eyring *et al.*, 2013a]. Considering the different sensitivities of stratospheric ozone to changes in CO₂, CH₄, and N₂O [WMO, 2014], the resulting trends in the SAM could depend on not just the combined radiative forcing of these gases, which is essentially what a climate

model forced with prescribed ozone responds to, but also the specific assumptions made about the future evolutions of CH₄ and N₂O, increases of which have opposing chemical effects on ozone. This means scenario-specific ozone climatologies are required which are consistent with the GHG evolutions assumed in these scenarios [Nowack *et al.*, 2014]. Such consistency may be model dependent, making it difficult to establish consistency across a range of different models. The findings presented here may increase the motivation to enter successors to CMIP5 (such as CMIP6) with coupled chemistry-climate models which by construction generate internally consistent ozone fields.

Appendix A: Details of Regression Analysis

We stipulate that in REF-C2, the ensemble average SAM index is the sum of five contributions:

$$t = \alpha r_{\text{EESC}} + (\beta_d + \beta_i) r_{\text{RF}} + \gamma + \rho. \tag{A1}$$

Here t is the ensemble mean SAM index for one of our five experiments, r_{EESC} is the dimensionless EESC regression function, r_{RF} is the dimensionless RF regression function (Figure 1), α and $\beta_d + \beta_i$ are regression coefficients, quantifying the influences of ozone depletion and the direct and indirect GHG impacts, respectively, γ is the generally nonzero average of t , and $\rho = \mathcal{O}(n^{-1/2})$ is the remainder not explained by the other terms, which scales with $n^{-1/2}$ (n = ensemble size). We calculate the coefficients performing a linear least squares regression. Uncertainties are given relative to the 95% confidence interval. β_d and β_i are individually quantified by differencing the regression coefficients of pairs of experiments that include and exclude, respectively, the impact of changing GHGs on ozone, such as SEN-C2-fODS and fOZONE. To reduce uncertainty in the regression coefficients, in experiments in which either GHGs do not change or anthropogenic ozone depletion does not occur, the corresponding regression coefficients are assumed 0. For example, $\alpha = 0$ for SEN-C2-fODS. Such zero coefficients are omitted in Figure 3.

The error calculation, which forms part of the above analysis, relies on two assumptions about the SAM indices: Residuals ρ need to be normally distributed, and autocorrelation needs to be small. Autocorrelation is addressed by applying the Durbin-Watson test [Montgomery *et al.*, 2001] to ρ :

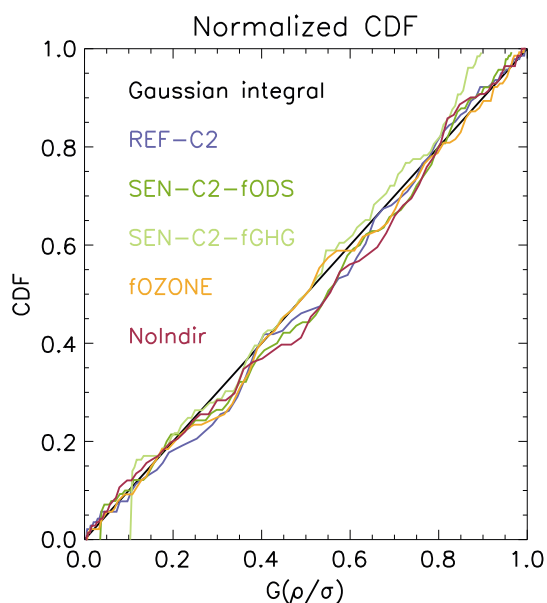


Figure A1. Cumulative distribution functions for the residuals ρ , displayed as $G(\rho/\sigma)$, where σ is the standard deviation of ρ and $G(x) = \frac{1}{\sqrt{\pi}} \int_{-\infty}^x \exp(-t^2) dt$ is the Gaussian integral. Perfect normal distribution (black). The five experiments (colors).

$$d = \frac{\sum_{t=2}^T (\rho_t - \rho_{t-1})^2}{\sum_{t=1}^T \rho_t^2}, \tag{A2}$$

where $T = 140$ is the number of complete summer seasons in the time series for an experiment and ρ is the residual in the regression fit (equation (A1)). No autocorrelation is indicated by $d = 2$, whereas $d \approx 1$ would usually indicate significant autocorrelation. For the eight time series shown in Figure 3 we find $1.85 < d < 2.06$. Hence, we conclude that autocorrelation does not significantly affect the regression analysis. Normality is assessed using a graphical approach: Figure A1 shows the cumulative distribution functions (CDF) for the five different experiments of the residuals ρ , expressed in multiples of their standard deviations σ . Residuals in all five ensemble averages satisfy essentially the same CDF which differs only slightly from a perfect Gaussian integral. Hence, we can assume that the residuals are normally distributed.

The paper compares simulations with interactive ozone with other simulations driven with prescribed ozone. This opens up the question of whether a model driven with interactive ozone behaves differently from an equivalent model in which ozone is prescribed. We have conducted an additional ensemble of four simulations which are forced by the same GHG and ODS scenario as REF-C2 but ozone is not calculated interactively. Instead, it is prescribed as the smoothed ensemble mean and monthly mean ozone field of the REF-C2 experiment. Note that this ozone climatology is zonally varying. We have applied the analysis to these simulations as well. The resulting regression coefficients do not differ significantly from REF-C2, suggesting that whether ozone is interactive or prescribed does not—within the limits of statistical analysis inherent here—affect the slowly varying components of the SAM discussed in this work.

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