



# Effects of ozone–vegetation coupling on surface ozone air quality via biogeochemical and meteorological feedbacks

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**Abstract.** Tropospheric ozone is one of the most hazardous air pollutants as it harms both human health and plant productivity. Foliage uptake of ozone via dry deposition damages photosynthesis and causes stomatal closure. These foliage changes could lead to a cascade of biogeochemical and biogeophysical effects that not only modulate the carbon cycle, regional hydrometeorology and climate, but also cause feedbacks onto surface ozone concentration itself. In this study, we implement a semi-empirical parameterization of ozone damage on vegetation in the Community Earth System Model to enable online ozone–vegetation coupling, so that for the first time ecosystem structure and ozone concentration can coevolve in fully coupled land–atmosphere simulations. With ozone–vegetation coupling, present-day surface ozone is simulated to be higher by up to 4–6 ppbv over Europe, North America and China. Reduced dry deposition velocity following ozone damage contributes to ~40–100% of those increases, constituting a significant positive biogeochemical feedback on ozone air quality. Enhanced biogenic isoprene emission is found to contribute to most of the remaining increases, and is driven mainly by higher vegetation temperature that results from lower transpiration rate. This isoprene-driven pathway represents an indirect, positive meteorological feedback. The reduction in both dry deposition and transpiration is mostly associated with reduced stomatal conductance following ozone damage, whereas the modification of photosynthesis and further changes in ecosystem productivity are found to play a smaller role in contributing to the ozone–vegetation feedbacks. Our results highlight

the need to consider two-way ozone–vegetation coupling in Earth system models to derive a more complete understanding and yield more reliable future predictions of ozone air quality.

## 1 Introduction

Tropospheric ozone is one of the air pollutants of the greatest concern due to its significant harm to human respiratory health. Increases of ozone since preindustrial times have been associated with a global annual burden of  $0.7 \pm 0.3$  million respiratory mortalities (Anenberg et al., 2010). Decades of observational records have also demonstrated the damaging effect of surface ozone on vegetation and crop productivity (Ainsworth et al., 2012). The phytotoxicity of ozone is shown to induce stomatal closure and reduce primary production, with ramifications for climate through the modification of surface energy and water fluxes and a decrease in the land carbon sink (Sitch et al., 2007; Wittig et al., 2007; Lombardozzi et al., 2015). Meanwhile, vegetation helps reduce ambient ozone concentration through stomatal deposition (e.g., Kroeger et al., 2014). However, the effect of such ozone-induced vegetation damage on ozone concentration itself, which thereby completes the ozone–vegetation feedback loop, has not been examined before but is potentially significant in modulating tropospheric ozone. This work uses a fully coupled land–atmosphere model to, for the first time, quantify the impacts of ozone–vegetation coupling on sur-

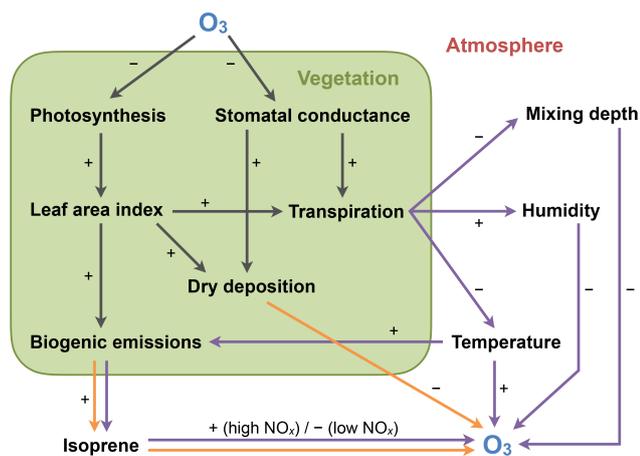
face ozone, and diagnoses the contributions from various feedback pathways in terrestrial ecosystems.

Tropospheric ozone is mainly produced from the photochemical oxidation of carbon monoxide (CO), methane (CH<sub>4</sub>) and non-methane volatile organic compounds (VOCs) by hydroxyl radical (OH) in the presence of nitrogen oxides (NO<sub>x</sub> ≡ NO + NO<sub>2</sub>). Vegetation plays various significant roles modulating surface ozone concentration. Precursor gases of ozone have large anthropogenic and natural sources, including vegetation and soil microbes for CH<sub>4</sub> and other VOCs. The most abundant single non-methane VOC species emitted by vegetation is isoprene (C<sub>5</sub>H<sub>8</sub>), which acts as a major precursor for ozone formation in polluted, high-NO<sub>x</sub> regions, but eliminates ozone by direct ozonolysis or by sequestering NO<sub>x</sub> as isoprene nitrate in more pristine environments (Fiore et al., 2011). The major sinks for tropospheric ozone include photolysis in the presence of water vapor and uptake by vegetation (i.e., dry deposition, mainly through the leaf stomata). Vegetation, therefore, plays a significant role in modulating ozone biogeochemically through dry deposition and biogenic VOC emissions. Meanwhile, transpiration from vegetation can affect ozone by regulating the overlying hydrometeorological environment. For instance, transpiration influences near-surface water vapor content, which affects the chemical loss rate of ozone. Transpiration also controls surface temperature and mixing depth, which can all influence the formation and dilution of ozone in the atmospheric boundary layer (Jacob and Winner, 2009).

Vegetation not only affects but is also affected by surface ozone. Stomatal uptake of ozone by leaves damages internal plant tissues, leading to severe damage to forest, grassland and agricultural productivity (Ashmore, 2005; Karnosky et al., 2007; Ainsworth et al., 2012). Elevated ozone since the industrial revolution is suggested to have reduced light-saturated photosynthetic rate and stomatal conductance by 11 and 13 %, respectively (Wittig et al., 2007). Modeling studies have also suggested that elevated ozone could decrease gross primary production (GPP) by 4–8 % in the eastern USA and more severely so (11–17 %) in several hot spots there (Yue and Unger, 2014), and decrease transpiration rate globally by 2–2.4 % (Lombardozzi et al., 2015), with significant implications for climate. For instance, the ozone-induced reduction in the global land carbon sink by 2100 is shown to have an indirect radiative forcing of +0.62–1.09 W m<sup>-2</sup>, which is comparable to the direct radiative forcing of ozone as a greenhouse gas (0.89 W m<sup>-2</sup>) and contributes to more pronounced warming (Sitch et al., 2007). Changes in stomatal conductance also modify the land–atmosphere exchange of water and energy and thus regional hydrometeorology (Bernacchi et al., 2011; Lombardozzi et al., 2015). In view of the important roles vegetation plays in shaping tropospheric ozone, the above biogeochemical and biogeophysical effects induced by ozone damage would affect not only weather and climate but would also constitute important feedbacks that ultimately affect ozone air quality itself.

In many land surface models, photosynthetic rate and stomatal conductance are highly coupled through the computation within the Farquhar–Ball–Berry model (Farquhar et al., 1980; Ball et al., 1987; Bonan et al., 2011). In global modeling studies on ozone-mediated vegetation changes and climate (Sitch et al., 2007; Collins et al., 2010; Yue and Unger, 2014), the effects of ozone damage on photosynthesis and stomata are thus strongly coupled to each other. Ozone uptake is assumed to directly affect photosynthetic rate, which in turn affects stomatal conductance via changes in internal CO<sub>2</sub> concentration. However, recent studies have suggested that separate modification of photosynthetic rate and stomatal conductance by cumulative ozone uptake in the Community Land Model (CLM) leads to better representation of plant responses to ozone exposure (Lombardozzi et al., 2012). This decoupling of ozone effects on photosynthesis and stomata is shown to decrease water use efficiency of affected plants, but leads to an overall smaller impact of ozone on transpiration and GPP than previously predicted.

Many climate–chemistry–biosphere modeling studies performed to date have demonstrated the importance of the co-evolution of climate, land cover and terrestrial ecosystems in air quality simulations and predictions (Wu et al., 2012; Tai et al., 2013; Pacifico et al., 2015), but they have not taken into account the potentially strong feedbacks arising from ozone damage on vegetation. For instance, ozone exposure can reduce stomatal conductance and thus transpiration rate, which may modify the partition between latent and sensible heat fluxes and lead to a cascade of meteorological changes: lower humidity that reduces the chemical loss rate of ozone; a thicker boundary layer that dilutes all pollutants, but may enhance entrainment, which either increases or decreases surface ozone depending on the vertical ozone profile (Super et al., 2015); and higher temperature that enhances ozone mainly through increased biogenic emissions and higher abundance of NO<sub>x</sub> (Jacob and Winner, 2009). These transpiration-mediated pathways can be characterized as biogeophysical feedbacks, as they are commonly known in the context of climate change, but here we prefer to call them hydrometeorological or simply “meteorological feedbacks” to emphasize that they are effected through ozone-induced changes in the hydrometeorological variables that ultimately affect ozone. On the other hand, reduced dry deposition caused by lower stomatal conductance and a possible decline in leaf area index (LAI) following ozone exposure can potentially increase ozone. The short-term impact of ozone on foliage-level isoprene emission is still under debate (Fares et al., 2006; Calfapietra et al., 2007), but as foliage density (e.g., represented by LAI) declines due to chronic ozone exposure (Yue and Unger, 2014), isoprene emission would likely decrease in the long term. These pathways directly involving plant biogeochemistry and atmospheric chemistry can be collectively termed “biogeochemical feedbacks”. Figure 1 summarizes the potentially important biogeochemical and meteorological feedbacks on surface ozone concentration, which



**Figure 1.** Possible pathways of ozone–vegetation coupling and feedbacks. The sign on each arrow indicates the sign of correlation or effect of one variable with or on another variable; the product of all signs along a given pathway indicates the overall sign of feedback. Orange arrows indicate biogeochemical feedbacks (i.e., via modulating atmospheric chemistry directly); purple arrows indicate meteorological feedbacks (i.e., via modifying the hydrometeorological environment). We focus only on processes that directly affect ozone; meteorological feedbacks on photosynthesis and stomatal conductance are included in the model but not emphasized in this figure.

are expected to have ramifications for simulations and future projections of ozone air quality. Such feedbacks may further alter atmospheric composition (e.g., aerosol and oxidant concentrations) and climate at large but remain poorly characterized in an Earth system modeling framework.

In this study, we adopt and implement a semi-empirical scheme for ozone-induced vegetation damage (Lombardo et al., 2015) into a coupled land–atmosphere model with fully interactive atmospheric chemistry and biogeochemical cycles, and examine the resulting impacts on present-day simulations of tropospheric ozone air quality with respect to observations. We perform sensitivity simulations to quantify the relative importance of different biogeochemical and meteorological feedback pathways, elucidate the larger sources of uncertainties, and make specific suggestions regarding Earth system model development.

## 2 Methods

### 2.1 Model description

This study investigates the impacts of ozone–vegetation coupling on ozone concentrations using the Community Earth System Model (CESM), which includes several different model components representing the atmosphere, land, ocean, and sea ice to be run independently or in various coupled configurations (Oleson et al., 2010; Lamarque et

al., 2012; Neale et al., 2013). We employ CESM version 1.2.2 with fully interactive atmosphere and land components, but with prescribed ocean and sea ice consistent with the scenarios of concern. For the atmosphere component, we use the Community Atmosphere Model version 4 (CAM4) (Neale et al., 2013) fully coupled with an atmospheric chemistry scheme (i.e., CAM-Chem) that contains full tropospheric  $O_3$ – $NO_x$ – $CO$ – $VOC$ –aerosol chemistry based on the MOZART-4 chemical transport model (CTM) (Emmons et al., 2010; Lamarque et al., 2012). This version of CAM-Chem simulates the concentrations of 56 atmospheric chemical species at a horizontal resolution of  $1.9^\circ \times 2.5^\circ$  latitude–longitude and 26 vertical layers for the atmosphere up to around 40 km.

For the land component, we use the Community Land Model version 4 (CLM4) (Oleson et al., 2010) with active carbon–nitrogen biogeochemistry (CLM4CN), which contains prognostic treatment of terrestrial carbon and nitrogen cycles (Lawrence et al., 2011). In CLM4, the Model of Emissions of Gases and Aerosols from Nature (MEGAN) version 2.1 is used to compute biogenic emissions online as functions of changing LAI, vegetation temperature, soil moisture and other environmental conditions (Guenther et al., 2012). For dry deposition of gases and aerosols we use the resistance-in-series scheme in CLM4 as described in Lamarque et al. (2012) with a further update of optimized coupling of stomatal resistance to LAI (Val Martin et al., 2014). Evapotranspiration is calculated based on the Monin–Obukhov similarity theory and the diffusive flux-resistance model with dependence on vegetation, ground and surface temperature, specific humidity, and an ensemble of resistances that are functions of meteorological and land surface conditions (Oleson et al., 2010; Lawrence et al., 2011; Bonan et al., 2011). Evapotranspiration is partitioned into transpiration, ground evaporation and canopy evaporation, with updates from Lawrence et al. (2011), and is linked to photosynthesis via the computation of stomatal resistance, as described below.

### 2.2 Photosynthesis–stomatal conductance model and ozone damage parameterization

The Farquhar–Ball–Berry model is used in CLM4CN to compute leaf-level photosynthetic rate and stomatal conductance under different environmental conditions (Farquhar et al., 1980; Ball et al., 1987). Leaf photosynthetic rate,  $A$  ( $\mu\text{mol CO}_2 \text{ m}^{-2} \text{ s}^{-1}$ ), is calculated as

$$A = \min(W_c, W_j, W_e), \quad (1)$$

where  $W_c$  is the Ribulose-1,5-bisphosphate carboxylase (RuBisCO)-limited rate of carboxylation,  $W_j$  is the light-limited rate, and  $W_e$  is the export-limited rate. Photosynthesis and stomatal conductance ( $g_s$ ) are related by

$$g_s = \frac{1}{r_s} = m \frac{A}{c_s} \frac{e_s}{e_1} P_{\text{atm}} + b, \quad (2)$$

where  $g_s$  is the leaf stomatal conductance;  $r_s$  is the leaf stomatal resistance ( $\text{s m}^2 \mu\text{mol}^{-1}$ );  $m$  is the slope of the conductance–photosynthesis relationship with values ranging from 5 to 9;  $c_s$  is the  $\text{CO}_2$  partial pressure at leaf surface (Pa);  $e_s$  is the vapor pressure at leaf surface (Pa);  $e_i$  is the saturation vapor pressure inside the leaf (Pa);  $P_{\text{atm}}$  is the atmospheric pressure (Pa); and  $b$  is the minimum stomatal conductance when  $A = 0$ , and is set to give a maximum stomatal resistance of  $20\,000 \text{ s m}^{-1}$  in CLM4 (Oleson et al., 2010).

Parameterization for the impact of ozone exposure on photosynthesis and stomatal conductance follows the work of Lombardozi et al. (2015), who tested the sensitivity of global ecosystem productivity and hydrometeorology to ozone damage on vegetation using satellite phenology (i.e., prescribed LAI, canopy height, etc.) and present-day ozone concentrations. The scheme uses two sets of ozone impact factors, one for modifying photosynthetic rate and another for stomatal conductance independently. These factors account for different plant groups, and are calculated based on the cumulative uptake of ozone (CUO) under different levels of chronic ozone exposure (Lombardozi et al., 2013). CUO ( $\text{mmol m}^{-2}$ ) integrates ozone flux into leaves over the growing season as

$$\text{CUO} = 10^{-6} \sum \frac{[\text{O}_3]}{k_{\text{O}_3} r_s + r_a} \Delta t, \quad (3)$$

where  $[\text{O}_3]$  is the instantaneous surface ozone concentration ( $\text{nmol m}^{-3}$ ) computed from CAM-Chem at a given model time step  $\Delta t$  ( $\Delta t = 1800 \text{ s}$  here);  $k_{\text{O}_3} = 1.67$  is the ratio of leaf resistance to ozone to leaf resistance to water,  $r_s$  is the stomatal resistance ( $\text{s m}^{-1}$ ), and  $r_a$  is the boundary layer and aerodynamic resistance between leaf surface and reference level ( $\text{s m}^{-1}$ ) (Sitch et al., 2007). Ozone uptake is only cumulated over time steps during the growing season when vegetation is most vulnerable to air pollution episodes; growing season is defined as the period in which total leaf area index (TLAI)  $> 0.5$  (Lombardozi et al., 2012). Ozone uptake only cumulates when the ozone flux is above an instantaneous critical threshold,  $0.8 \text{ nmol O}_3 \text{ m}^{-2} \text{ s}^{-1}$ , to account for ozone detoxification by vegetation at lower ozone levels (Lombardozi et al., 2015). Three different plant groups are accounted for: evergreen, deciduous, and crops or grasses. We also include a leaf-turnover ozone decay rate for evergreen plants so that accumulated ozone damage does not accrue beyond the average foliar lifetime. The ozone impact factors have empirical linear relationships with CUO such that

$$F_{\text{pO}_3} = a_p \times \text{CUO} + b_p, \quad (4)$$

$$F_{\text{cO}_3} = a_c \times \text{CUO} + b_c, \quad (5)$$

where  $F_{\text{pO}_3}$  is the ozone damage factor multiplied to the photosynthesis rate ( $A$ ), and  $a_p$  and  $b_p$  are slope and intercept, respectively, from empirical and experimental studies (listed in Table 1);  $F_{\text{cO}_3}$  is the ozone damage factor multiplied with

**Table 1.** Slopes (per  $\text{mmol m}^{-2}$ ) and intercepts (unitless) used to calculate ozone impact factors in Eqs. (4) and (5), following Lombardozi et al. (2015).

Plant group	Photosynthesis		Conductance	
	Slope ( $a_p$ )	Intercept ( $b_p$ )	Slope ( $a_c$ )	Intercept ( $b_c$ )
Broadleaf	0	0.8752	0	0.9125
Needleleaf	0	0.839	0.0048	0.7823
Crops and grasses	−0.0009	0.8021	0	0.7511

stomatal conductance ( $g_s$ ), and  $a_c$  and  $b_c$  are the corresponding slope and intercept (Table 1). The ozone damage is applied to the optimal photosynthesis and stomatal conductance values, which are calculated iteratively at first without ozone damage, to allow the damage to be applied independently.

### 2.3 Model experiments

Incorporating the ozone–vegetation parameterization above into CLM4CN and coupling it with CAM-Chem, we allow, for the first time, ecosystem structure (e.g., in terms of LAI and canopy height) to evolve in response to ozone exposure but at the same time allow ozone concentration to evolve in response to such ecosystem changes. Therefore, previously discussed feedbacks are mostly included. We conduct four sets of fully coupled land–atmosphere simulations: (1) a control case without ozone damage on vegetation ([CTR]); (2) a simulation with both photosynthetic rate and stomatal conductance modified by ozone impact factors (independently) ([PHT + COND]), following the approach of Lombardozi et al. (2015); (3) a simulation where we apply the ozone impact factor to photosynthetic rate only ([PHT]), but stomatal conductance is calculated using the intact, optimal photosynthetic rate; and (4) simulation where we apply the ozone impact factor to stomatal conductance only ([COND]), but photosynthetic rate is calculated using the intact stomatal conductance. Simulations [PHT] and [COND], when compared with [PHT + COND], allow us to quantify the relative contribution from each pathway. To determine the relative contribution of those pathways involving biogenic emissions toward the overall ozone–vegetation feedback, we conduct an additional set of sensitivity simulations with prescribed isoprene emission and MEGAN turned off: a control case with no MEGAN (CTR\_nM), and a simulation with modified photosynthesis and stomatal conductance but with no MEGAN ([PHT + COND\_nM]). To determine the relative contribution of pathways involving dry deposition vs. transpiration, we compare simulated results with that of Val Martin et al. (2014) who have used the similar CAM-Chem–CLM framework but without ozone–vegetation coupling to

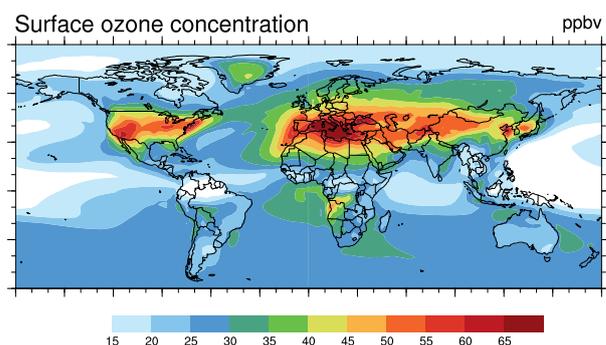
test the sensitivity of ozone to perturbations in dry deposition velocity.

All simulations are conducted for 20 years using year 2000 initial conditions and the corresponding land cover data (e.g., land cover and land use types, satellite LAI, etc.). The first 5 years of outputs are treated as spin-up and thus discarded in the analysis. We observe that the annual averages of key aboveground ecosystem parameters such as LAI and ozone concentration come into a relatively steady state after 5 years. We focus on changes in the 15-year northern summertime (June, July and August: JJA) averages for most of the variables in the rest of this paper because this is the period when the growing season of the majority of global vegetation overlaps most significantly with the high-ozone season, especially in the northern midlatitudes.

### 3 Simulated ozone with and without ozone–vegetation coupling

Figure 2 shows the 15-year mean summertime surface ozone concentration from the [PHT + COND] simulation. The corresponding CUO used to affect vegetation is shown in Supplement Fig. S1. Simulated ozone is generally higher in the northern midlatitudes than elsewhere, and is the highest over the Mediterranean where solar radiation is particularly strong. CUO also has high values in Europe, but the overall distribution does not exactly follow that of surface ozone concentration because CUO also depends on the length of the growing season and stomatal conductance. CUO ranges between 20 and 70 mmol m<sup>-2</sup> over regions with both high summertime ozone and high productivity. The simulated CUO is comparable in both magnitude and spatial distribution with Lombardozzi et al. (2015), who used prescribed meteorology, ozone and vegetation phenology with no active carbon–nitrogen cycle or atmospheric coupling, as opposed to this study. This suggests that online ozone–vegetation coupling, which can modify ozone concentration substantially depending on the region, leads to a similar pattern of ozone uptake by vegetation to the case using prescribed ozone due to the compensation between higher (lower) concentration and higher (lower) stomatal resistance, as reflected in Eq. (3). During the growing season, CUO is used to calculate the ozone impact factors that modify photosynthetic rate and stomatal conductance according to Eqs. (4) and (5) and parameter values listed in Table 1.

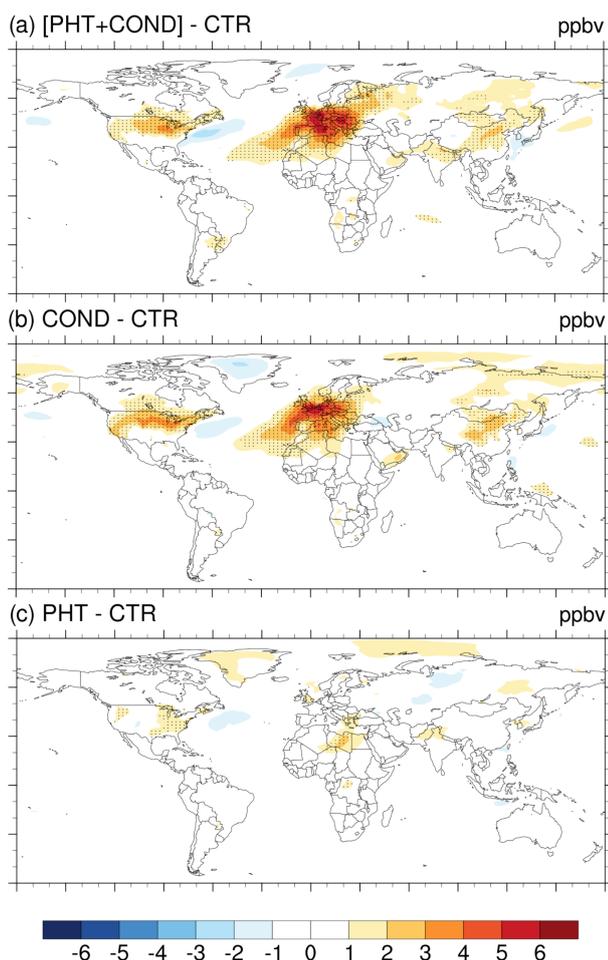
Figure 3 shows the differences in surface ozone concentration in different simulations from the control case (corresponding relative changes shown in Fig. S2). Implementing ozone–vegetation coupling that includes simultaneous modification of photosynthetic rate and stomatal conductance by ozone exposure (the [PHT + COND] case) increases mean surface ozone globally, and significant increases by up to 4–6 ppbv are found over China, North America and Europe (Fig. 3a). Ozone exposure is thus found to constitute a pos-



**Figure 2.** Mean summertime (JJA) surface ozone concentration from the [PHT + COND] case, where ozone uptake simultaneously modifies both photosynthetic rate and stomatal conductance. Results are averaged over the last 15 years of simulations.

itive feedback loop via vegetation that ultimately enhances surface ozone levels when ozone–vegetation coupling is accounted for.

The simulated increases in ozone levels due to ozone–vegetation coupling are significant when compared with the possible impacts of 2000–2050 climate and land cover changes on surface ozone, which are in the range of +1–10 ppbv (Jacob and Winner, 2009; Tai et al., 2013; Val Martin et al., 2015). This coupling effect is smaller than the potential ozone changes driven by anthropogenic emissions (up to +30 ppbv), but it more likely reflects compensation among various pathways (e.g., Ganzeveld et al., 2010). These simulated increases, however, slightly worsen the performance of CAM-Chem in reproducing ozone concentrations against observations, as seen in Fig. 4, which shows the model–observation comparison for the control case (standard CAM-Chem-CLM with dry deposition improvement of Val Martin et al., 2014) and the [PHT + COND] case. The high biases in CESM-simulated summertime surface ozone concentrations in North America and Europe are a commonly acknowledged issue with CAM-Chem (Lamarque et al., 2012) and other global and regional models (Lapina et al., 2014; Parrish et al., 2014). Uncertain emissions, coarse resolution (Lamarque et al., 2012), misrepresentation of dry deposition process and overestimation of stomatal resistance (Val Martin et al., 2014) are all likely factors contributing to these high biases. Inclusion of ozone–vegetation coupling in the model further increases the normalized mean biases of the modeled results against three sets of observational data: the Clean Air Status and Trends Network (CASTNET) (1999–2001), Air Quality System (AQS) (1999–2001) and European Monitoring and Evaluation Programme (EMEP) (1999–2001), from 18 to 22 %, 31 to 35 % and 14 to 22 %, respectively. Although there remains considerable uncertainty in the parameterization of ozone–vegetation coupling and in ozone simulations by Earth system models, we show that including ozone damage in a coupled climate–chemistry–biosphere



**Figure 3.** Changes in summertime surface ozone concentrations in different simulations: (a) the case where both photosynthetic rate and stomatal conductance are modified by ozone uptake; (b) modified photosynthetic rate only; and (c) modified stomatal conductance only, all relative to the control case (CTR). Stippling with dots indicates significant changes at 90 % confidence from Student's *t* test.

framework can have a potentially significant impact on surface ozone simulations.

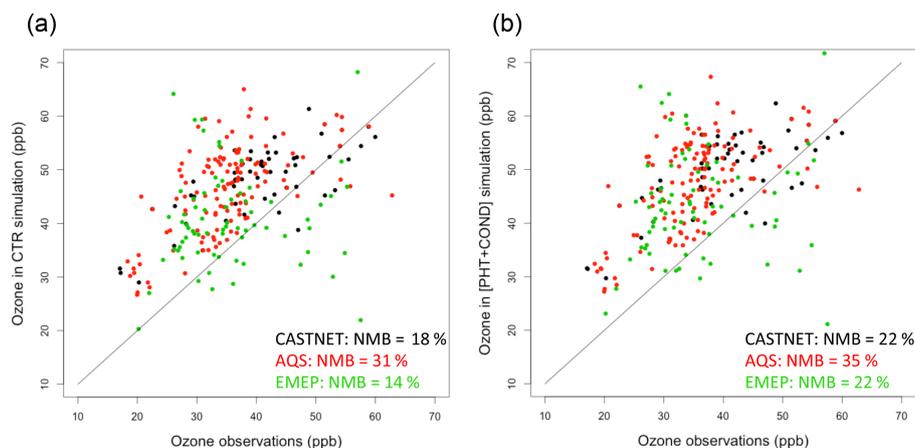
#### 4 Attribution to different biogeochemical and meteorological feedback pathways

Figure 3b and c show the differences in ozone for the cases where ozone damages stomatal conductance alone and photosynthesis alone, respectively, noting that each of them is calculated using the undamaged, intact values of the other variable. Comparison of Fig. 3a with b–c shows that the modification of stomatal conductance by ozone uptake contributes more dominantly to the overall effect of ozone–vegetation coupling (Fig. 3a). This suggests that, among the various feedback pathways that may influence surface ozone

(Fig. 1), those triggered by changes in stomatal conductance are generally more important than those associated with photosynthesis or the associated changes in ecosystem production and structure including LAI, at least in the modeling framework of this study. This is also supported by sensitivity simulations performed under the same modeling framework but without ozone damage, in which a 50 % increase in LAI decreases summertime surface ozone by on average 3 ppb, which is relatively small in comparison with the changes following optimization of stomatal resistance (Val Martin et al., 2014). Indeed, the effect of modifying stomatal conductance alone ([COND]; Fig. 3b) is slightly larger than the case of [PHT + COND] (Fig. 3a), where the additional effect of modifying photosynthesis together with stomatal conductance would slightly offset the overall positive feedback on ozone. It is noteworthy that this additional effect is, however, not consistent with the effect of modifying photosynthesis alone ([PHT]; Fig. 3c), reflecting nonlinear interactions between photosynthesis and stomatal conductance.

Figure 5 shows the differences in dry deposition velocity, transpiration rate and biogenic isoprene emission between the [PHT + COND] and [CTR] simulations (relative changes shown in Fig. S3). Over China, Europe and North America, ozone dry deposition velocity is lower (by up to  $\sim 20\%$ ) in [PHT + COND]. In these same regions but especially in the eastern USA, southern Europe and southern China, isoprene emission is significantly higher (by up to  $\sim 50\%$ ). In addition, in similar regions but especially in central North America, the transpiration rate is reduced by ozone exposure (by up to  $\sim 20\%$ ), which would reduce boundary-layer humidity, increase surface temperature, enhance dry convection and thicken the boundary layer. In view of Fig. 1, all of these pathways may add to or offset each other, leading to the overall ozone changes seen in Fig. 3a. The sensitivity simulations and comparison with Val Martin et al. (2014), which examined the sensitivity of simulated ozone to differences in dry deposition schemes under essentially the same modeling framework, allow us to quantify more precisely which of these pathways are more important, as we discuss next.

Figure 6a shows the changes in surface ozone in the [PHT + COND\_nM] minus CTR\_nM simulations, where we use prescribed biogenic emissions from the original control case (CTR) to drive ozone chemistry so that we essentially shut down any feedback pathways involving biogenic emissions. A comparison between Figs. 6a and 3a shows that the changes in biogenic VOC emissions account for  $\sim 0\text{--}60\%$  of the ozone increases over Europe, North America and China, while dry deposition and/or transpiration-driven meteorological changes (excluding the temperature effect on isoprene emission) account for remaining  $\sim 40\text{--}100\%$ . We further show in Fig. 6b the theoretical changes in surface ozone, by multiplying the dry deposition changes in Fig. 5a by the change in ozone concentration per unit change in dry deposition velocity from the study of Val Martin et al. (2014), which provided an approximate sensitivity of simulated ozone to



**Figure 4.** Scatter plots of simulated summertime ozone concentration in (a) the control case (CTR) and (b) the case where both photosynthesis and conductance are modified by ozone uptake ([PHT + COND]), vs. observed average values from the Clean Air Status and Trends Network (CASTNET) (1999–2001), Air Quality System (AQS) (1999–2001) and European Monitoring and Evaluation Programme (EMEP) (1999–2001). Normalized mean biases (NMB) are also shown.

perturbed dry deposition velocity only to separate this impact from that due to hydrometeorological changes associated with changing stomatal conductance, e.g., changes in mixing depth. We find that the ozone changes in Fig. 6a and b are similar in magnitude, suggesting that globally most of the non-isoprene-driven differences in ozone are driven by dry deposition. Notable exceptions include the US Midwest and southeastern Europe, where higher mixing depth following reduced transpiration might have partly offset the ozone positive feedback, whereas in western Europe the lower chemical loss rate following reduced transpired water might have further enhanced the positive feedback.

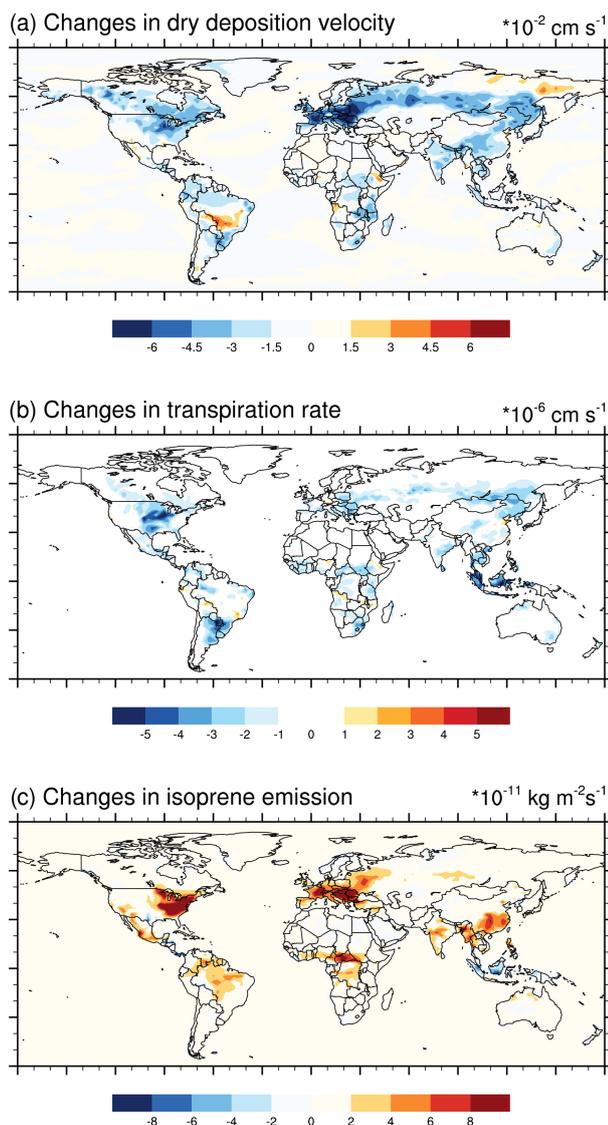
The simulated general reduction in dry deposition velocity and transpiration rate (Fig. 5a and b) is mostly due to increased stomatal resistance (Fig. 7a), i.e., reduced stomatal conductance, a direct response to CUO. The reduced dry deposition velocity represents a positive biogeochemical feedback on ozone (orange arrows in Fig. 1). The simulated increase in biogenic isoprene emission (Fig. 5c) is found to be mostly driven by higher surface (thus vegetation) temperature (Fig. 7b) that results from lower transpiration rate and latent heat flux (Fig. 7c). Therefore, this feedback loop involving biogenic emissions is indeed an indirect, meteorological feedback that is also initiated by stomatal and transpiration changes (purple arrows in Fig. 1). Relative changes in variables shown in Fig. 7 are included in Fig. S4.

By including immediate ozone–vegetation coupling, we find a larger decline in transpiration rate (6.4 % globally) than in the offline, uncoupled land model results (2.0–2.4 %) estimated by Lombardozzi et al. (2015). On the other hand, although reduced photosynthesis and the resulting long-term changes in GPP and LAI (Fig. 7d–e) play a smaller role than reduced stomatal conductance in shaping simulated ozone (Fig. 3b–c), the impacts are not negligible (up to 3 ppb), espe-

cially as these changes are also nonlinearly coupled to stomatal changes. Photosynthetic rate decreases by up to 20 %, directly due to the ozone effect (Fig. 7f), which is quite similar both in magnitude and spatial pattern to the results of Lombardozzi et al. (2015), but the corresponding GPP and LAI changes are relatively small ( $\sim 5\%$  over regions concerned, except for Southeast Asia, where the highest ozone-induced LAI reduction is simulated and leads to isoprene emission decrease despite higher surface temperature). Grid-level GPP and LAI in certain areas increase despite reduced leaf-level photosynthetic rate, likely reflecting more carbon allocation to leaves to compensate for the reduced photosynthetic rate and relaxation of resource limitation, as nutrients and water become less limiting upon lower photosynthetic and evaporative demands, as well as favorable hydrometeorological changes following ozone exposure (enhanced soil moisture and precipitation as shown in Fig. S5). These LAI increases induced by ozone are not represented in Fig. 1 because they more likely reflect the fully coupled effect of changing hydrometeorology, instead of the direct effect of ozone on LAI as is typically observed in experimental studies (Ainsworth et al., 2012).

## 5 Conclusions and discussion

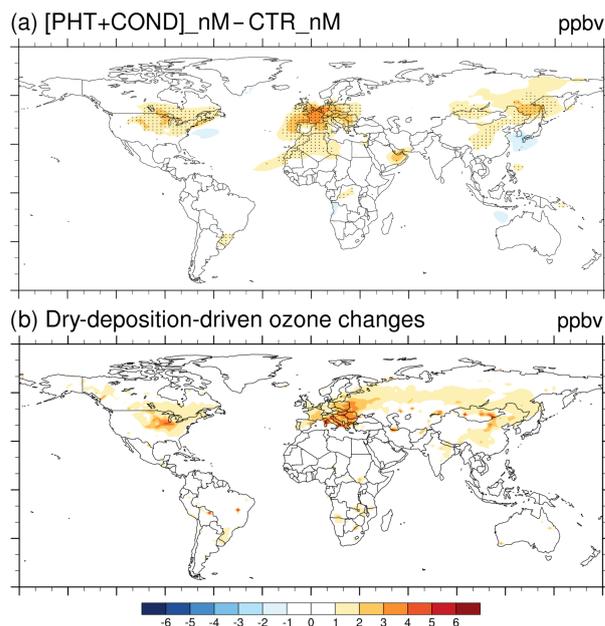
Tropospheric ozone is one of the most hazardous air pollutants due to its harmful effects on human health and damage to forest and agricultural productivity. Stomatal uptake of ozone by leaves reduces both photosynthetic rate and stomatal conductance. These vegetation changes can induce a cascade of biogeochemical and biogeophysical (or meteorological) effects (Fig. 1) that ultimately modulate climate, carbon cycle and also feedback onto ozone air quality itself. The direct, biogeochemical feedback pathways include reduced



**Figure 5.** Changes in (a) dry deposition velocity, (b) transpiration rate and (c) isoprene emission in the [PHT + COND] case, where both photosynthetic rate and stomatal conductance are modified by ozone uptake, relative to the control case (CTR).

ozone dry deposition and biogenic VOC emissions. The indirect, meteorological feedback pathways are facilitated by transpiration-driven changes in the meteorological environment that influence ozone formation and removal. A few land surface modeling studies have estimated the direct effects of ozone on ecosystem production and land–atmosphere water exchange (Yue and Unger, 2014; Lombardozzi et al., 2015), and predicted a possible positive radiative forcing from the ozone-induced decline in the land carbon sink (Sitch et al., 2007).

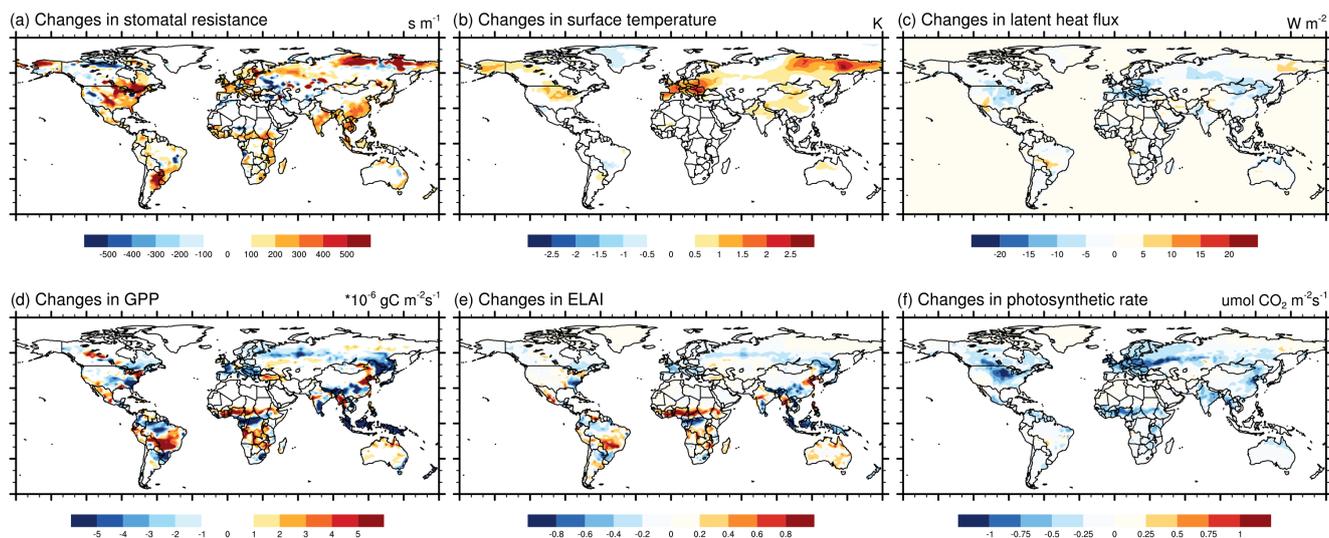
In this study, we implement a semi-empirical parameterization of ozone damage on vegetation (Lombardozzi et al., 2015) into the CESM (CAM4-Chem-CLM4CN) modeling



**Figure 6.** Changes in surface ozone concentration in (a) the case where both photosynthesis and stomatal conductance are modified by ozone uptake, but with prescribed isoprene emission from the original control case (CTR) by turning off MEGAN (stippling with dots indicates significant changes at 90 % confidence from Student's *t* test); and (b) theoretical changes calculated by multiplying our simulated dry deposition changes with the change in ozone concentration per unit change in dry deposition from Val Martin et al. (2014), which did not include ozone damage on vegetation.

framework to enable online ozone–vegetation coupling so that vegetation variables can evolve in response to ozone exposure, and at the same time simulated ozone concentration can respond to ecosystem changes. Our scheme modifies leaf-level photosynthesis and stomatal conductance separately via the ozone impact factors, which are assumed to have empirical linear relationships with CUO and account for different plant groups. Sensitivity simulations are conducted to determine the relative importance of different feedback pathways.

With ozone–vegetation coupling, surface ozone is simulated to be higher by up to 4–6 ppbv over Europe, North America and China. This coupling effect is significant in view of the 2000–2050 effects of climate and land cover changes on surface ozone (+1–10 ppbv) as found in previous work (Jacob and Winner, 2009; Ganzeveld et al., 2010; Tai et al., 2013), and should be considered in future air quality projection studies. Reduced dry deposition velocity following the modification contributes to  $\sim 40$ – $100$  % and enhanced biogenic isoprene emission contributes to  $\sim 0$ – $60$  % of the higher ozone concentrations. The dry deposition-driven ozone increases (by up to 4 ppbv) arise mostly from reduced stomatal conductance, and are consistent with the sensitivity of ozone to perturbations in dry deposition veloc-



**Figure 7.** Changes in (a) stomatal resistance, (b) surface temperature, (c) latent heat flux, (d) gross primary production (GPP), (e) effective leaf area index (ELAI) and (f) photosynthetic rate in the [PHT + COND] case, where both photosynthetic rate and stomatal conductance are modified by ozone uptake, relative to the control case (CTR).

ity found by Val Martin et al. (2014). This pathway constitutes a significant positive biogeochemical feedback on surface ozone. The other major feedback associated with enhanced isoprene emission is mostly driven by higher vegetation temperature that results from lower transpiration rate. This pathway constitutes an indirect, positive meteorological feedback on surface ozone. Depending on the region, transpiration-driven meteorological changes such as lower humidity and deeper mixing depth may also influence surface ozone. Transpiration rate is simulated to decrease by 6.4 % globally, which is a larger change compared with the decrease estimated by Lombardozzi et al. (2015), who used prescribed instead of synchronously simulated atmospheric forcings. This also suggests an augmented effect on transpiration due to changes in carbon allocation and foliage density arising from ozone–vegetation coupling.

Modification of photosynthesis and further long-term changes in ecosystem productivity and structure, including LAI changes, are found to play a smaller role in contributing to the ozone–vegetation feedbacks than direct stomatal changes, but are not insignificant (up to +3 ppbv). The simulated changes in LAI (less than 5 %) in this study are similar in magnitude to that by Yue and Unger (2015), who included an active carbon cycle though using Yale Interactive terrestrial Biosphere (YIBs) model with a different ozone–vegetation parameterization. However, prognostic treatment of the carbon cycle and LAI calculation in CLM4CN are still known to be problematic, with large uncertainties and biases in the estimation of global carbon fluxes (Sun et al., 2012), arising from incomplete model parameterization and from uncertainty in photosynthetic parameters (Bonan et al., 2011). It is not surprising that changes in GPP as simulated

here do not replicate the results of Lombardozzi et al. (2015), in which vegetation phenology is prescribed and the carbon and nitrogen cycles are not active (CLM4.5SP). Implementing ozone damage on vegetation in a model with more sophisticated and realistic representation of prognostic carbon–nitrogen cycle is highly warranted, so that the possible effects of ozone-induced long-term ecosystem changes can be examined more fully.

Large variability in the responses of different plants to ozone leads to considerable uncertainties in any global-scale studies (Lombardozzi et al., 2013). In some cases, such large variability in plant responses across different studies weakens the correlation between phytotoxic responses and CUO. Such correlation is usually more evident in individual studies, and in the parameterization schemes based on them (Sitch et al., 2007; Yue and Unger, 2015). The parameterization developed by Lombardozzi et al. (2013), based on the most comprehensive database available for photosynthetic and stomatal responses to CUO to date, is deemed more appropriate for the global scale of this study and the plant functional types represented in the model, despite the weaker correlation between plant responses and CUO as shown by the compilation of data across studies. The damage is applied after CUO reaches a certain threshold, so the calculation of CUO is still crucial to the application of the damage functions. The model results could possibly be improved with more detailed plant-type-specific ozone damage parameterization, including better estimates of plant vulnerability to ozone that will help refine the ozone uptake thresholds (Lombardozzi et al., 2015). An important caveat of this study is the consideration of only three plant groups to generalize the re-

sponses of global vegetation to ozone exposure, because data are largely unavailable for other plant groups.

Another potential caveat is the uncertainty and lack of cross-validation in hydrometeorological simulations with respect to the ozone phytotoxicity scheme we newly implement, as we only focus on vegetation and atmospheric chemical changes in this study. Although most simulated vegetation variables are consistent with previous work, the changes in simulated vegetation temperature from ozone–vegetation coupling are not small (by up to +2 °C) (Fig. 7b) and they result in quite substantial changes in isoprene emission, suggesting the need for further tuning of hydrometeorological processes in the model. Also, MEGAN does not consider the direct, immediate biochemical connection between photosynthesis and biogenic emissions, by which ozone damage to photosynthesis may directly reduce isoprene emission and partially offset the significant temperature-induced increase in isoprene emission as shown in Fig. 5c (Tiwari et al., 2016). Whereas the various environmental activity factors used in MEGAN to adjust baseline emissions may have implicitly encapsulated the biochemical connection with photosynthesis, further incorporating such a connection into ozone–vegetation modeling warrants more in-depth investigation. In general, we have the highest confidence in the quantification of the biogeochemical pathway via stomata-driven deposition changes, which is straightforward and accounts for the majority of the ozone–vegetation feedbacks. On the other hand, the hydrometeorological feedbacks introduce strong nonlinearity in the interactions between atmospheric chemistry, soil moisture and vegetation, which is more difficult to isolate. Parameterizing the ozone–vegetation coupling in a standalone chemical transport model with prescribed meteorology could be particularly helpful, to more confidently separate the effects of biogeochemical vs. meteorological feedbacks. This knowledge will be important in projecting the impacts of future climate and land cover changes on ozone air quality and climate feedbacks in the coming decades.

## 6 Data availability

Model output data used for analysis and plotting, as well as processed observational data from CASTNET, AQS and EMEP, can be made available in RData format by contacting the corresponding author (Amos P. K. Tai: amostai@cuhk.edu.hk).

**The Supplement related to this article is available online at doi:10.5194/acp-17-3055-2017-supplement.**

*Competing interests.* The authors declare that they have no conflict of interest.

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