Revision of Global soil nitrous oxide emissions in a dynamic carbon-nitrogen model

Dear Soenke,

Thank you very much for your competent handling of our manuscript and the opportunity to publish our work in Biogeosciences.

We have taken care of the suggested corrections, as follows

p6 l28 Te-> The
Done

p7 line 29: Please change Zaehle & Friend 2010 to Smith et al. 2014 BGS. I actually do not use a transpiration relationship in a transient sense.
Done

p10 l 6: generate -> generates
Corrected

p23 l17ff. If is unclear here whether you talk about the LM3V modelled soil moisture or the “data-derived” models. please be very explicit here.
We changed the sentence as follows:

“However, despite using data-derived soil moisture, it appears that the prediction of soil moisture is an impediment towards validating N2O emissions at field scale for both LM3V-N simulated and reanalysed soil moisture.”

p25 l8 include references here
We added the following references: Xu-Ri et al., 2012; Zaehle et al., 2011

p25 l13: provide reference or explain briefly why the BNF in models requires improvement (please no exhaustivistic discussion!)

We changed this part of the paragraph as follows

The representation of BNF in models requires improvement: currently, simple empirical relationships are used, yet BNF is the largest source of new N in terrestrial systems, and therefore is critical in the determination of N availability to nitrifiers and denitrifiers. Here we showed that different implementations of BNF (prescribed vs. responsive to plant N demand) are globally important for N2O emissions.

p27 l23 opposite -> of opposite sign?
Corrected

p27 l 23ff “while for climate it is the combination of temperature and interaction”. Sorry I don’t understand this.

We rephrased to:

“On the other hand, simulations with O-CN (Zaehle et al., 2011) showed the effects of CO2 and climate to be approximately equal and of opposite sign for historic simulations covering the past 300 years that also include land-cover changes. They evaluated the effect of climate change as the difference between a simulation that consider both CO2 and climate and a simulation that does not consider climate change. Thus their climate effect contains both the single effect of climate and the interaction of climate with CO2. The temperature (i.e. climate) response on top of CO2 can in our simulation be calculated as the temperature effect plus the interaction effect (Fig 8). Analyzed in this fashion, LM3V-N’s results are congruent with those of Zaehle et al. (2011), albeit we found a slightly weaker temperature effect compared to CO2.”

p29 l 4 “distinct and new results ... is treated differently”. Please be more precise!

We are now more explicit:

“Our sensitivity analysis shows that processes of the larger plant-soil N cycle affect fast N cycle processes as evidenced by the response to the fire and BNF modification. This sensitivity can lead to differences in N2O across models (e.g. in the response to CO2 and climate) even if existing nitrification-denitrification schemes are identical.”

p29 l 5. a response of N2O, or the N cycle in general. And - what does this imply...?

We added the central implication of the rarity of tropical manipulative experiments.

“Further, our work suggests a much stronger response to warming and CO2 in tropical forests compared to extratropical forest, thus extrapolation of mostly extra-tropical field studies to the globe warrants caution.”

check spelling and formatting (N2O etc).

We did further editorial changes to eliminate spelling, grammatical, and formatting error. Please see the attached marked-up version.

Best regards,
Stefan Gerber and Yuanyuan Huang
Global soil nitrous oxide emissions in a dynamic carbon-nitrogen model

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Abstract

Nitrous oxide (N\textsubscript{2}O) is an important greenhouse gas that also contributes to the depletion of stratospheric ozone. Due to its high temporal and spatial heterogeneity, a quantitative understanding of terrestrial N\textsubscript{2}O emission, its variabilities and responses to climate change is challenging. We added a soil N\textsubscript{2}O emission module to the dynamic global land model LM3V-N\textsubscript{2}O, and tested its sensitivity to mechanisms that affect the level of mineral nitrogen (N\textsubscript{2}) in soil such as plant N uptake, biological N fixation, amount of volatilized N redeposited after fire, and nitrification-denitrification. We further tested the relationship between N\textsubscript{2}O emission and soil moisture, and assessed responses to elevated CO\textsubscript{2} and temperature. Results extracted from the corresponding gridcell (without site-specific forcing data) were comparable with the average of cross-site observed annual mean emissions, although differences remained across individual sites if stand-level measurements were representative of gridcell emissions. Processes, such as plant N uptake and N loss through fire volatilization that regulate N availability for nitrification-denitrification have strong controls on N\textsubscript{2}O fluxes in addition to the parameterization of N\textsubscript{2}O loss through nitrification and denitrification. Modelled N\textsubscript{2}O fluxes were highly sensitive to water filled pore space (WFPS), with a global sensitivity of approximately 0.25 TgN per year per 0.01 change in WFPS. We found that the global response of N\textsubscript{2}O emission to CO\textsubscript{2} fertilization was largely determined by the response of tropical emissions with reduced N\textsubscript{2}O fluxes in the first few decades and increases afterwards. The initial reduction was linked to N limitation under higher CO\textsubscript{2} level, and was alleviated through feedbacks such as biological N fixation. The extratropical response was weaker and generally positive, highlighting the need to expand field studies in tropical ecosystems. We did not find synergistic effects between warming and CO\textsubscript{2} increase as reported in analyses with different models.
Warming generally enhanced N₂O efflux and the enhancement was greatly dampened when combined with elevated CO₂, although CO₂ alone had a small effect. The differential response in the tropics compared to extratropics with respect to magnitude and sign suggests caution when extrapolation from current field CO₂ enrichment and warming studies to the global scale.

1 Introduction

Nitrous oxide (N₂O) is a major reactant in depleting stratospheric ozone as well as an important greenhouse gas (Ravishankara et al., 2009; Butterbach-Bahl et al., 2013; Ciais et al., 2013). With a global warming potential of 298 times more (per unit mass) than that of carbon dioxide (CO₂) over a 100-year period (Forster et al., 2007), the contributions of N₂O emissions to global radiative forcing and climate change are of critical concern (Zaehle and Dalmonech, 2011). The concentration of atmospheric N₂O has been increasing considerably since the industrial revolution with a linear rate of 0.73±0.03 ppb yr⁻¹ over the last three decades (Ciais et al., 2013). Although applications of synthetic fertilizer and manure during agriculture intensification have been identified as the major causes of this increase (Davidson, 2009; Zaehle and Dalmonech, 2011; Zaehle et al., 2011), nonagricultural (natural) soil is still an important N₂O source (Ciais et al., 2013; Syakila and Kroeze, 2011). N₂O fluxes from nonagricultural soils are highly heterogeneous, which limits our ability to estimate and predict global scale budgets, and quantify the response of natural N₂O fluxes to global environmental changes (Butterbach-Bahl et al., 2013; Ciais et al., 2013).

Most of the N₂O fluxes from soil are produced by microbial nitrification and denitrification (Braker and Conrad, 2011; Syakila and Kroeze, 2011). Nitrification is an aerobic process that oxidizes ammonium (NH₄⁺) to nitrate (NO₃⁻), during which some N is lost as N₂O. Denitrification reduces nitrate or nitrite to gaseous N (i.e. NOₓ, N₂O and N₂), a process that is fostered under anaerobic conditions. During denitrification N₂O is generated in intermediary steps where a small portion can escape from the soil before further reduction to N₂ takes place. Soil texture, soil NH₄⁺, soil water filled pore space (WFPS), mineralization rate, soil pH, and soil temperature are well-known regulators of nitrification N₂O fluxes (Parton et al., 1996; Li et al., 2000; Parton et al., 2001). Denitrification and associated N₂O emissions depend primarily on carbon supply, the redox potential and soil NO₃⁻ (Firestone and Davidson, 1989; Parton et
Soil moisture has a particular strong impact (Galloway et al., 2003; Schlesinger, 2009) as it influences nitrification and denitrification rates through its regulations on substrate availability and soil redox potential (as oxygen diffusion proceeds at much slower rate in water filled than in air filled pore space), thereby also controlling the partitioning among various denitrification products (i.e., NO$_3$, N$_2$O and N$_2$) (Firestone and Davidson, 1989; Parton et al., 2001). Although emissions are known to be sensitive to soil moisture, quantitative understanding of its role in terrestrial N$_2$O fluxes and variability is limited (Ciais et al., 2013). At regional to global scale, the application of the “hole-in-pipe” concept (Firestone and Davidson, 1989) in the CASA biosphere model pioneered one of the earliest process-based estimation of natural soil N$_2$O fluxes. The model calculated the sum of NO, N$_2$O and N$_2$ fluxes as a constant portion of gross mineralized N, and the relative ratios of N trace gases (NO$_x$:N$_2$O:N$_2$) as a function of soil moisture (Potter et al., 1996). While the early models of nitrification and denitrification are primarily conceptual driven, recent global N$_2$O models combine advancements in global dynamic land models with more detailed processes, including microbial dynamics. Xu-Ri and Prentice (2008) simplified nitrification and denitrification modules from DNDC (i.e., DeNitrification-DeComposition) (Li et al., 1992; Li et al., 2000) in their global scale dynamic N scheme (DyN) and incorporated DyN into the LPJ dynamic global vegetation model. In the DNDC approach, nitrification and denitrification were allowed to occur simultaneously in aerobic and anaerobic microsites. Zaehele et al. (2011) incorporated a nitrification-denitrification scheme into the O-CN land model following largely the LPJ-DyN with minor modifications and additions of the effects of soil pH and chemo-denitrification that originated from DNDC (Li et al., 2000). Stocker et al. (2013) embedded the LPJ-DyN approach into an Earth System Model and investigated the feedbacks of N$_2$O emissions, together with CO$_2$ and CH$_4$, to climate. Compared to LPJ-DyN approach, Saikawa et al. (2013) retained the explicit simulation of nitrifying and denitrifying bacteria from DNDC in their CLMCN-N2O module based on CLM V3.5 land model. Simulations with O-CN demonstrated a positive response of N$_2$O emissions to historical warming and a negative response to historical CO$_2$ increase, globally. While CO$_2$ and interaction with climate change resulted in an increase in historical and future N$_2$O emissions in LPJ-DyN (Xu-Ri et al., 2012) and its application in LPX-Bern (Stocker et al., 2013), respectively, historical CO$_2$ change alone, i.e. single factor of Xu-Ri et al. (2012), caused a slight decrease in historical N$_2$O emissions. The negative CO$_2$ response seems to be in disagreement with one meta-analysis of manipulative field experiments.
showing an increase in N\textsubscript{2}O emissions at elevated levels of CO\textsubscript{2} (Zaehle et al., 2011; Xu-Ri et al., 2012; van Groenigen et al., 2011). The discrepancy in response to global change factors needs to be addressed both in models and in the interpretation of manipulative field experiments.

Here we add a N\textsubscript{2}O gas emission module to LM3V-N, a land model developed at the Gephysical Fluid Dynamics Laboratory (GFDL). In this paper, we will first briefly introduce LM3V-N and describe the added N\textsubscript{2}O emission module. We then subject the model to historic changes in CO\textsubscript{2}, N deposition, and recent climate change to infer natural N\textsubscript{2}O emissions in the past few decades. We test the model’s sensitivity to soil water regime, by addressing the parameterization of soil WFPS, and by replacing the model soil moisture with two different soil moisture reanalysis products. We also conduct sensitivity tests with regard to the general N cycling and parameterization of N\textsubscript{2}O emissions. We then subject the model to step changes in atmospheric CO\textsubscript{2} and temperature to understand modelled responses to CO\textsubscript{2} fertilization and climate change. Since we build largely on existing parameterization of nitrification-denitrification processes, we will briefly discuss implications from transferring process formulations to LM3V-N where other aspects of the N cycle are treated differently.

2 Methods

2.1 Model description

LM3V is capable of simulating ecosystem dynamics and exchange of CO\textsubscript{2}, water and energy between land and atmosphere with the fastest time step of 30 minutes (Shevliakova et al., 2009). LM3V-N expands the LM3V land model with a prognostic N cycle (Gerber et al., 2010), and includes five plant functional types (PFTs): C3 and C4 grasses, tropical, temperate deciduous and cold evergreen trees. Each PFT has five vegetation C pools (leaf, fine root, sapwood, labile, and wood), two litter and two soil organic C pools and their corresponding N pools based on the specific C:N ratios. Photosynthesis is coupled with stomatal conductance on the basis of the Collatz et al.‘s (1991,1992) simplification of the Farquhar scheme (Farquhar et al., 1980). Soil hydrology in LM3V-N follows partly on Land Dynamics (LaD) with further improvements (Shevliakova et al., 2009; Milly and Shmakin, 2002; Milly et al., 2014). N enters the ecosystem through atmospheric N deposition and biological N fixation (BNF), losses via fire and leaching of dissolved organic N (DON) as well as mineral N. We briefly describe the major
characteristics of LM3V-N in the next subsection (2.1.1), and details are available in Gerber et al. (2010).

2.1.1 Main characteristic of LM3V-N

2.1.1.1 C-N coupling in vegetation

We briefly describe the larger plant-soil N cycle and how it links to mineral N (ammonium and nitrate). Plants adjust their uptake of C and N to maintain their tissue specific C:N ratios, which are PFT-dependent constants. Instead of varying C:N ratios in tissues, short-term asynchronies in C and N assimilations or temporary imbalances in stoichiometry are buffered by an additional N storage pool (S) in which N is allowed to accumulate once plant N demand is satisfied. The optimum storage size $S_{\text{target}}$ is based on tissue turnover $Q_{\text{liv}}$:

$$S_{\text{target}} = t_h Q_{\text{N,liv}}$$

where $t_h$ is the time span that buffer plant N losses (currently set as 1 year). Plant N status ($x$) is defined as the fraction of the actual N storage compared to the target storage: $x = S / S_{\text{target}}$. Consequently, N constraints on photosynthesis and soil N assimilation are based on plant N status:

$$A_{g,N} = A_{g,pot}(1 - e^{-\varphi x})$$

$$U_{N,P} = U_{N,P,pot} \cdot \begin{cases} 1 & \text{if } S < S_{\text{target}} \\ 0 & \text{else} \end{cases}$$

where $A_{g,N}$ indicates N constrained rate of gross photosynthesis ($\mu$mol C m$^{-2}$ s$^{-1}$) and $A_{g,pot}$ corresponds to the potential photosynthetic rate without N limitation. The parameter $\varphi$ mimics the metabolic deficiency as plant N decreases. $U_{N,P,pot}$ is the potential inorganic N uptake rate from soil available ammonium and nitrate pools. The actual inorganic N uptake rate ($U_{N,P}$) operates at its potential if plants are N limited and drops to zero when N storage ($S$) reaches its target size. Overall this set-up intends to overcome short-term asynchronies between C and N supply.

2.1.1.2 Soil C-N interactions in organic matter decomposition

Organic matter decomposition is based on a modified CENTURY approach (Bolker et al., 1998), and amended with formulations of N dependent C and N mineralization rates. Here, we use a 3 pool model where the pools broadly represent labile and structural litter, and processed
soil organic matter. Decomposition is the main source of available N for nitrification and denitrification. In turn, NO$_3^-$ and NH$_4^+$ can both trigger the decomposition of “light” organic matter and stabilize C in “heavy” organic matter in LM3V-N. Formation of a slow decomposable organic matter pool leads to immobilization of ammonium and nitrate to satisfy the fixed carbon to nitrogen ratio of this pool.

2.1.1.3 Competing sinks of available N

The fate of soil mineral N (i.e. ammonium and nitrate) depends on the relative strength of the competing sinks, with the broad hierarchy of sorption > soil immobilization > plant uptake > leaching/denitrification. This creates a tight N cycle, since internal (plant and soil) sinks dominate over N losses. Denitrification thus far has been lumped with leaching losses and summed into a generic N loss term. Sorption/desorption buffers available N and is assumed to have the highest priority and be at steady state in each model time step. N immobilization into organic matter occurs during transfers among litter and soil organic matter pools. Leaching losses of available N are simulated on the basis of drainage rate. Plant uptake of mineral N is a combination of both active and passive processes. The active uptake is modeled as a Monod function, and the passive transport is a function of available N and plant transpiration:

$$ U_{N,plant,i} = \frac{\gamma_{\text{max}} C_r [N_{\text{tot}}]}{h_i \left(k_{p,1/2} [N_{\text{tot}}]\right)} + \left[N_{\text{cor}}\right] Q_{w,t} $$  \hspace{1cm} (4)

where $\gamma_{\text{max}}$ (yr$^{-1}$ kg C$^{-1}$) stands for the maximum uptake rate per unit root mass $C_i$, $h_i$ is soil depth, $k_{p,1/2}$ is the half saturation constant, and $Q_{w,t}$ represents the transpiration flux of water. The subscript $i$ refers to either ammonium or nitrate, while $[N_{\text{tot}}]$ is the concentration of the combined dissolved ammonium nitrate pool. Potential uptake and thus effective removal of available N occurs if plants are N limited (see Equation 3).

2.1.1.4 N losses from organic pools

With the implementation of high ecosystem N retention under limiting condition where internal N sinks outcompeting losses from the ammonium/nitrate pools, losses via organic pathways become important (Gerber et al., 2010; Thomas et al., 2015). Over the long term, N losses via fire and DON are thus critical factors limiting ecosystem N accumulation and maintaining N limitation in LM3V-N. N volatilized via fire is approximated as a function of CO$_2$ produced in a fire, stoichiometric ratio of burned tissues but reduced by a global retention factor representing the fraction of N that is retained as ash (ash$\_fraction$, currently set as 0.45). DON leaching is
linked to hydrologic losses of dissolved organic matter ($L_{DOM}$) and its C:N ratio. In turn $L_{DOM}$ is based on drainage rate ($Q_W$) and a buffer or sorption parameter $b_{DOM}$ (currently set as 20).

$$L_{DOM} = \frac{Q_W}{n_{b_{DOM}}} \cdot DOM$$  \hspace{1cm} (5)

where $DOM$ is the amount of dissolve organic matter in the soil column. Soil depth ($h_s$) is used to convert DOM unit to concentration (in unit of kgC m$^{-3}$). Production of DOM (in unit of kgC m$^{-2}$) is assumed to be proportional to the decomposition flux of the structural litter and soil water content. Both, losses via fire and via DOM are losses from a plant-unavailable pool (Thomas et al., 2015), and have the potential to increase or maintain N limitation over longer timescales, and consequently reduce N availability for N$_2$O production through sustained and strong plant N uptake.

2.1.1.5 Biological nitrogen fixation (BNF)

BNF in LM3V-N is dynamically simulated on the basis of plant N availability, N demand, and light condition. BNF increases if plant N requirements are not met by uptake. The rate of up-regulation is swift for tropical trees but constrained by light penetrating the canopy for other PFTs, mimicking the higher light requirements for new recruits that possibly can convert atmospheric N$_2$ into plant available forms. In turn, sufficient N uptake reduces BNF. The BNF parameterization thus creates a negative feedback, where high plant available N and thus the potential for denitrification is counteracted with reduction of N input into the plant-soil system.

This explicit negative feedback is different to other models where BNF is parameterized based on NPP (Thornton et al., 2007), or transpiration (Zaehle and Friend, 2010; Smith et al., 2014). The inclusion of BNF as a negative feedback contributes to a rather tight cycling within LM3V-N, with low overall rates of BNF under unperturbed conditions (Gerber et al., 2013).

2.1.2 Soil N$_2$O emission

LM3V-N assumes that nitrification is linearly scaled to ammonium content, and modified by soil temperature and soil moisture. Gaseous losses so far were not differentiated from hydrological leaching. We add a soil nitrification-denitrification module which accounts for N gaseous losses from NH$_3$ volatilization, nitrification and denitrification. The nitrification-denitrification scheme implemented here combines features from both the DNDC model (Li et al., 1992; Li et al., 2000) and the CENTURY/DAYCENT model (Parton et al., 1996; Parton et al., 2001; Del Grosso et al., 2000). In this subsection, we provide details on the nitrification-
denitrification module which explicitly simulates N\textsubscript{2}O gas losses from\textsubscript{via} nitrification and denitrification, as well as other process modifications compared to the original LM3V-N.

### 2.1.2.1 Nitrification-Denitrification

Transformation among mineral N species (ammonium and nitrate) occurs mainly through two microbial pathways: nitrification and denitrification. Although ongoing debate exists in whether nitrification rates may be well described by bulk soil ammonium concentration or soil N turnover rate (Parton et al., 1996; Zaehle and Dalmonech, 2011), we adopt the donor controlled scheme (ammonium concentration). In addition to substrate, soil texture, soil water filled pore space (WFPS, the fraction of soil pore space filled with water), and soil temperature are all well known regulators of nitrification. As a first order approximation, nitrification rate ($N$, in unit, kgN m\textsuperscript{-2} year\textsuperscript{-1}) is simulated as a function of soil temperature, NH\textsubscript{4}\textsuperscript{+} availability and WFPS,

$$N = k_n f_n(T) f_n(WFPS) \frac{N_{NH_4^+}}{b_{N,NH_4^+}}$$

where $k_n$ is the base nitrification rate (11000 year\textsuperscript{-1}, the same as in LM3V-N) (Gerber et al., 2010); $N_{NH_4^+}$ is ammonium content (in unit, kgN m\textsuperscript{-2}); $b_{N,NH_4^+}$ is the buffer or sorption parameter for NH\textsubscript{4}\textsuperscript{+} (unitless, 10 in LM3V-N) (Gerber et al., 2010); $f_n(T)$ is the temperature response function following Li et al. (2000), with an optimum temperature for nitrification at 35°C; and $f_n(WFPS)$ is the soil water response function. The effect of WFPS on nitrification is texture dependent, with most of the reported optimum value around 0.6 (Parton et al., 1996; Linn and Doran, 1984). We adopt the empirical WFPS response function from Parton et al. (1996) with medium soil texture.

$$f_n(T) = \left(\frac{60-T_{soil}}{25.78}\right)^{3.503} \times e^{\frac{3.503(T_{soil}-34.22)}{25.78}}$$

$$f_n(WFPS) = \left(\frac{WFPS-0.67}{0.59988}\right)^{-1.27} \times \left(\frac{WFPS-0.0012}{0.59988}\right)^{2.84}$$

where $T_{soil}$ is the soil temperature in degree Celsius.

Denitrification is controlled by substrate NO\textsubscript{3} (electron acceptor), labile C availability (electron donor), soil moisture and temperature. Labile C availability is estimated by soil heterotrophic respiration (HR). Following LPJ-DyN (Xu-Ri and Prentice, 2008), denitrification is assumed to have a $Q_{10}$ value of 2 when the soil temperature is between 15 and 25 °C. The soil moisture response function is adopted from Parton et al. (1996). Soil pH is reported to be an important
indicator of chemodenitrification which occurs predominantly in acidic soils (pH<5) under conditions of high nitrite concentration (Li et al., 2000). However, its role for N\textsubscript{2}O production is not well studied (Li et al., 2000) and we do not model chemodenitrification explicitly.

\[ D = k_df(T)df(WFPS) f_T^\text{NO}_2 \]  
\[ \text{And } f_T = \frac{HR}{HR + Kc \text{NO}_2 + K_n} \]  
\[ \text{NO}_2 = \frac{N_{\text{NO}_3}^3}{b_{\text{NO}_3}} \]

where \(D\) is the denitrification rate (in unit, kgN m\textsuperscript{-2} year\textsuperscript{-1}); \(k_d\) is the base denitrification rate (8750 year\textsuperscript{-1}); \(f_T\) mimics the impact of labile C availability and substrate (nitrate) on the growth of denitrifiers, adapted from Li et al. (2000); \(Kc\) and \(K_n\) are half-saturation constants taken from Li et al. (2000) \((0.0017 and 0.0083 \text{ kgN m}\textsuperscript{-2}\) respectively, assuming an effective soil depth of 0.1m); \(b_{\text{NO}_3}\) is the buffer or sorption parameter for NO\textsubscript{3}\^- (unitless, 1 in LM3V-N) (Gerber et al., 2010); \(N_{\text{NO}_3}^3\) and \(\text{NO}_2\) are nitrate content before and after being buffered (in unit, kgN m\textsuperscript{-2}); and \(f(df\text{T})\) and \(f(df\text{WFPS})\) are empirical soil temperature and water response function for denitrification, adopted from Xu-Ri and Prentice (2008) and Parton et al. (1996), respectively.

\[ f_T(T) = e^{308.56 \times \left( \frac{1}{b_{\text{NO}_3}} - \frac{1}{(b_{\text{NO}_3} + D)} \right)} \]  
\[ f_T(WFPS) = \frac{1.56}{12.0 \times \left( 12^{2.68 \times WFPS} \right)} \]

### 2.1.2.2 Gaseous partitions from nitrification-denitrification

\(\text{N}_2\text{O}\) is released as a byproduct during both nitrification and denitrification. The fraction of \(\text{N}_2\text{O}\) lost during net nitrification is uncertain (Li et al., 2000; Xu-Ri and Prentice, 2008). Here we set this fraction to be 0.4\%, which is higher than Goodroad and Keeney (1984), but at the low end provided by Khalil et al. (2004). Nitrification also generates NO\textsubscript{3} gas, in addition to \(\text{N}_2\text{O}\). N losses as NO\textsubscript{3} emissions during nitrification are scaled to the \(\text{N}_2\text{O}\) release using a variable NO\textsubscript{3}:N\textsubscript{2}O ratio \(R_{\text{NO}_3:\text{N}_2\text{O}}\). \(R_{\text{NO}_3:\text{N}_2\text{O}}\) varies with relative gas diffusivity \(D_r\), the relative gas diffusivity in soil compared to air (Parton et al., 2001), which is calculated from air filled porosity (AFPS, i.e., the portion of soil pore space that is filled by air) (Davidson and Trumbore, 1995).
\[ R_{\text{NO}_x:\text{N}_2\text{O}} = 15.2 + \frac{35.5 \times \text{ATAN}(0.68 \times \pi \times (10 \times D_r - 1.68))}{\pi} \]  
\[ D_r = 0.209 \times \text{AFPS}^{\frac{4}{3}} \]  
where ATAN stands for the trigonometric arctangent function; \( \text{AFPS} \) is the air filled porosity (1-WFPS), and \( \pi \) is the mathematical constant, approximately 3.14159.

During denitrification, the gaseous ratio between \( \text{N}_2 \) and \( \text{N}_2\text{O} \) \( (R_{\text{N}_2:\text{N}_2\text{O}}) \) is calculated following the empirical function derived by Del Grosso et al. (2000), which combines the effects of substrate (\( \text{NO}_3^- \)) to electron donor \( (HR, \text{the proxy for labile C}) \) ratio and WFPS. \( R_{\text{N}_2:\text{N}_2\text{O}} \) is updated at every time step and for each grid cell.

\[ R_{\text{N}_2:\text{N}_2\text{O}} = Fr\left(\frac{\text{NO}_3^-}{HR}\right) \cdot Fr(WFPS) \]  
With

\[ Fr\left(\frac{\text{NO}_3^-}{HR}\right) = \max(0.16 \times k, k \times e^{-(0.8 \times \frac{\text{NO}_3^-}{HR})}) \]  
\[ Fr(WFPS) = \max(0.1, 0.015 \times WFPS - 0.32) \]

where \( k \) is a texture dependent parameter (Table 1) estimated from Del Grosso et al. (2000). \( k \) controls the maximum value of the function \( Fr\left(\frac{\text{NO}_3^-}{HR}\right) \).

15 Other modified processes

To complete the N loss scheme in LM3V-N, we also added \( \text{NH}_3 \) volatilization into LM3V-N. \( \text{NH}_3 \) volatilization in soil results from the difference between the equilibrium \( \text{NH}_3 \) partial pressure in soil solution and that in the air. Dissolved \( \text{NH}_3 \) is regulated by ammonium concentration and pH. The net flux of \( \text{NH}_3 \) from soil to the atmosphere varies with soil \( \text{NH}_3 \), moisture, temperature, therefore

\[ \text{NH}_3 = k_{\text{nh}}f(pH)f_{\text{NH}_3}(T)(1 - WFPS) \frac{N_{\text{NH}_3}}{\sigma_{\text{NH}_3}} \]  
where \( \text{NH}_3 \) is the net ammonia volatilization flux (in unit, kgN m\(^{-2}\) year\(^{-1}\)); \( k_{\text{nh}} \) is the base ammonia volatilization rate (365 year\(^{-1}\)); \( f(pH) \) is the pH factor and \( f(T) \) is the temperature factor which are given by the following two equations:

\[ f(pH) = e^{2 \times (pH - 10)} \]  
\[ f_{\text{NH}_3}(T) = \min(1, e^{308.566 \times \left(\frac{1}{T+273.15} - \frac{1}{871 + 0.843 T}\right)}) \]
where $pH_{soil}$ is the soil pH which is prescribed instead of simulated dynamically, $f(pH)$ and $f(T)$ follow largely on the NH$_3$ volatilization scheme implemented in the dynamic global vegetation model LPJ-DyN (Xu-Ri and Prentice, 2008).

2.2 Model experiments

2.2.1 Global hindcast with potential vegetation

To understand the model performance and compare with other models and observations, we conducted a hindcast simulation with potential vegetation. The model resolution was set to 3.75 degrees longitude by 2.5 degrees latitude. We forced the model with 3 hourly reanalysis weather data based on Sheffield et al. (2006). We used a 17 year recycled climate of 1948-1964 for the spin-up and simulation years prior to 1948. Atmospheric CO$_2$ concentration was prescribed with 284 ppm for model spin-up and based on ice core and atmospheric measurements for transient simulations (Keeling et al., 2009). N deposition was set as natural background for simulations before 1850 (Dentener and Crutzen, 1994), and interpolated linearly between the natural background and a snapshot of contemporary (1995) deposition (Dentener et al., 2006) for simulations after 1850. Soil pH was prescribed and derived from the Harmonized World Soil Database (HWSD) version 1.1, the same as NACP model driver data (Wei et al., 2014).

The model was spun up from bare ground without C-N interactions for the first 68 years and with C-N interactions for the following 1200 years to develop and equilibrate C and N stocks.

To accelerate the spin-up process, slow litter and soil C and N pools were set to the equilibrium values based on litterfall inputs and decomposition/leaching rates every 17 years. We determined the model to reach a quasi-equilibrium state by confirming the drift to be less than 0.03 PgC yr$^{-1}$ for global C storage and 0.2 TgN yr$^{-1}$ for global N storage. From this quasi-equilibrium state, we initialized the global hindcast experiment starting from 1850 using the corresponding climatic forcings, CO$_2$ and N deposition data. In the following analysis, we will focus mostly on the last three decades (1970-2005).

2.2.2 Sensitivity to soil water filled pore space (WFPS)

While LM3V-N carries a simplified hydrology, we bracketed effects of soil moisture by exploring the parameterization of WFPS and by substituting the predicted soil moisture with 3-hourly re-analysis data. Levels of soil water (in unit kg m$^{-2}$) therefore stem from: (1) the
simulated water content based on LM3V-N soil water module, hereafter LM3V-SM (2) the
Global Land Data Assimilation System Version 2 with the land surface model NOAH 3.3
(Rodell et al., 2004), hereafter NOAH-SM, and (3) the ERA Interim reanalysis dataset from
European Center for Medium range Weather Forecasting (ECMWF) (Dee et al., 2011),
hereafter ERA-SM. The latter two datasets integrate satellite and ground based observations with
land surface models. When overriding soil moisture, we linearly interpolated the 3 hourly data
onto the 30 minutes model time step. In these simulations, we allowed soil C and N dynamics
to vary according to different soil moisture datasets, but kept the model prediction of soil water
to use for plant productivity and evapotranspiration.

Parameterization of the soil moisture effect on nitrification and denitrification are based on
WFPS. LM3V-N uses the concept of plant available water, where water that is available to
plants varies between the wilting point and field capacity. Water content above the available
water capacity (i.e., the difference between field capacity and wilting point) leaves the soil
immediately (Milly and Shmakin, 2002), and thus WFPS does not attain high values typically
observed during denitrification. To explore the effect of WFPS – soil moisture relationship on
N\textsubscript{2}O emissions, we calculated WFPS using three methods. Method 1 assumes WFPS is the ratio
of available water and the available water capacity in the rooting zone. In Method 2 we assumed,
WFPS is the ratio of the water filled porosity and total porosity which is derived from bulk
density (BD, in unit g cm\textsuperscript{-3}). BD was obtained from the Harmonized World Soil Database
(HWSD) version 1.1 (Wei et al., 2014). The calculation is given by

\[ WFPS = \frac{\theta}{\theta_{w}} \quad (22) \]

where \( \theta \) (kg m\textsuperscript{-2}) is the root zone soil water; \( h \) (m) is the effective rooting depth of vegetation;
\( \rho \) is the density of water (1000 kg m\textsuperscript{-3}); and \( PD \) is the particle density of soil (2650 kg m\textsuperscript{-3}).
Method 1 generally leads to an overestimation of WFPS because the available water capacity
is smaller than total pore space. In contrast, the use of Method 2 with LM3V-SM creates an
underestimation since water is not allowed to accumulate beyond field capacity and misses high
WFPS to which nitrification and denitrification are sensitive. Meanwhile, for NOAH-SM and
ERA-SM data, Methods 2 is more close to the “real” WFPS and is the default method when
using these data sets. The third approach, which is also the default method with LM3V-SM that
is applied in the global hindcast experiment, the subsequent elevated CO\textsubscript{2} and temperature
responses experiment, and sensitivity tests with regard to N cycling, calculates WFPS as the average of the previous two methods.

For each soil moisture dataset (3 in total, 2 replacements and 1 simulated by LM3V-N), we calculated WFPS using three methods mentioned above. We conducted transient simulations with the nine different WFPSs (3 datasets × 3 methods) starting from the near equilibrium state obtained in the global hindcast experiment in 2.2.1. The use of less realistic Method for WFPS for each soil moisture driver (LM3V-SM, NOAH-SM and ERA-SM) offers insights of the sensitivity of N₂O emissions to soil moisture. The simulation procedure was the same as that in global hindcast experiment except for the WFPS. ERA-SM is only available starting from 1979, prior to which simulations were conducted with model default soil moisture (LM3V-SM). Results from ERA-SM were analyzed starting from 1982, leaving a short period for adjustment.

2.2.3 Sensitivity to N cycling processes and parameterization

N₂O emission is constrained by ecosystem availability of mineral N, which is linked to different N cycling processes in addition to nitrification and denitrification processes. To test the sensitivity of modelled N₂O emission to the larger plant-soil N cycle, we conducted the following sensitivity analyses, in form of a one at a time perturbation. We replaced the dynamic BNF scheme with empirically reconstructed preindustrial fixation rates (Cleveland et al., 1999; Green et al., 2004), removing the negative feedback between BNF and plant N availability. We further shut off N loss pathways through DON leaching and fire volatilization (with ash_fraction = 1). We expect that these three modifications alleviate N limitation: Prescribed BNF may continuously add N beyond plant N demand. Further eliminating fire and DOM N losses leave loss pathways that have to pass the available N pool thereby opening the possibility of increasing gaseous losses. Further, removing these plant-unavailable pathways (Thomas et al., 2015) increases N retention and opens the possibility of alleviating N limitation. In addition, we modified key parameters related to general N cycling and N₂O emissions one-at-a-time. We multiplied several parameters that directly affect ammonium and nitrate concentration or N₂O fluxes by 10 (×10) or 0.1 (×0.1), while kept other parameters as defaults. Those parameters control the active root N uptake rates (v_{max}), nitrification rate (k_n), denitrification rate (k_d, Kc, Kn) and the fraction of net nitrification lost as N₂O (frac).
2.2.4 Responses to elevated CO$_2$ and temperature

Respones of N$_2$O emissions to atmospheric CO$_2$ and global warming have been reported at field scale (Dijkstra et al., 2012; van Groenigen et al., 2011). Here, we evaluate the model’s response to step changes in form of a doubling of preindustrial CO$_2$ level (284 ppm to 568 ppm) and a 2K increase in atmospheric temperature. Starting from the same quasi-equilibrium state with potential vegetation as in the global hindcast experiment in 2.2.1, we conducted four transient model runs: (1) the CONTROL run with the same drivers as spin-up; (2) the CO2_FERT run with the same drivers as the CONTROL except a doubling of atmospheric CO$_2$ level; (3) the TEMP run with the same drivers as the CONTROL except a 2K rise in atmospheric temperature; and (4) the CO2_FERT×TEMP run with both the doubling of CO$_2$ and 2K rise in temperature. For each experiment, we ran the model for 100 years and evaluated the corresponding results.

2.3 Comparisons with observations

We compared our model results for annual N$_2$O gas loss with field data: We compiled annual N$_2$O emissions from peer-reviewed literature (see Appendix A for more information). To increase the representativeness of the measurements, we included only sites with more than 3 months or 100 days experimental span. We limited our datasets where there was no reference to a disturbance of any kind. Only locations with at least 50 years non-disturbance history for forests and 10 years for vegetation other than forests were included. The compiled 61 measurements cover a variety of spatial ranges with vegetation types including tropical rainforest, temperate forest, boreal forest, tundra, savanna, perennial grass, steppe, alpine grass and desert vegetation. Multiple measurements falling into the same model grid cell were averaged. If the authors had indicated the dominant vegetation or soil type, we used the values reported for the dominant type instead of the averaged. For multiyear measurements, even if the authors gave the individual year’s data, we averaged the data to avoid overweighting of long term studies. If the location was between borders of different model grid cells, we averaged across the neighboring grid cells.

We also compared monthly N$_2$O fluxes at a group of sites: (a) the Tapajós National Forest in Amazonia (3°S, 55°W), taken from Davidson et al. (2008); (b) the Hubbard Brook Experimental Forest in New Hampshire, USA (44°N, 72°W), as described in Groffman et al. (2006); (c) the cedar forest from Oita, Japan (33°N, 131°E), as described in Morishita et al.
1 (2007); (d) the *Leymus chinensis* (LC) and *Stipa grandis* (SG) steppe in Inner Mongolia, China
2 (44°N, 117°E), taken from Xu-Ri et al. (2003); (e) the cedar forest in Fukushima, Japan (37°N,
3 140°E), taken from Morishita et al. (2007); and (f) the primary (P1 and P2) and secondary (L1
4 and L2) forests located at the Pasir Mayang Research Site (1°S, 102°E), Indonesia, taken from
5 Ishizuka et al. (2002). In addition, daily measurements of soil temperature, soil moisture and
6 N₂O emissions were compared at four German forest sites located in the same grid cell (50°N,
7 8°E), as described in Schmidt et al. (1988).

3 Results

3.1 Global budget, seasonal and inter-annual variability

Our modelled global soil N₂O flux is 6.69±0.32 TgN yr⁻¹ (1970-2005 mean and standard
1 deviation among different years) (Fig.1) with LM3V-SM (Method 3, default method for
2 LM3V-N calculated soil moisture), 5.61±0.32 TgN yr⁻¹ with NOAH-SM (Method 2) and
3 7.47±0.30 TgN yr⁻¹ with ERA-SM (1982-2005, Method 2) which is within the range of reported
4 values: The central estimate of N₂O emission from soils under natural vegetation is 6.6 TgN yr⁻¹
5 based on the Intergovernmental Panel on Climate Change (IPCC) AR5 (Ciais et al., 2013)
6 (range, 3.3–9.0 TgN yr⁻¹) for the mid-1990s. Mean estimation for the period of 1975-2000
7 ranged from 7.4 to 10.6 TgN yr⁻¹ with different precipitation forcing data (Saikawa et al.,
8 2013). Xu-Ri et al. (2012) reported the decadal-average to be 8.3-10.3 TgN yr⁻¹ for the 20th century.
9 Potter and Klooster (1998) reported a global mean emission rate of 9.7 TgN yr⁻¹ over 1983-
10 1988, which is higher than the earlier version of their model (6.1 TgN yr⁻¹) (Potter et al., 1996).
11 Other estimates include 6-7 TgN yr⁻¹ (Syakila and Kroeze, 2011), 6.8 TgN yr⁻¹ based on the O-
12 CN model (Zaehle et al., 2011), 3.9-6.5 TgN yr⁻¹ for preindustrial periods from a top-down
13 inversion study (Hirsch et al., 2006), 1.96-4.56 TgN yr⁻¹ in 2000 extrapolated from field
14 measurements by an artificial neural network approach (Zhuang et al., 2012), 6.6-7.0 TgN yr⁻¹
15 for 1990 (Bouwman et al., 1995), and 7-16 TgN yr⁻¹ (Bowden, 1986) as well as 3-25 TgN yr⁻¹
16 (Banin, 1986) from two earlier studies.

Following Thompson et al. (2014), El Niño years are set to the years with the annual
18 1993, 1994, 1997 and 1998 were chosen as El Niño years. We detected reduced emissions
during El Niño years (Fig. 1), in line with the global atmospheric inversion study of Thompson et al. (2014) and the process based modelling study from Saikawa et al. (2013).

Figure 2 shows the simulated global natural soil N\textsubscript{2}O emissions in 4 seasons averaged over the period of 1970-2005 based on LM3V-SM (Method 3). The northern hemisphere displays a large seasonal variability, with the highest emissions in the northern summer (JJA, June to August) and lowest in winter (DJF, December to February). Globally, northern spring (MAM, March to May) has the highest emission rate (2.07 TgN) followed by summer (1.89 TgN). The smaller emissions in summer compared to spring stems from a reduced contribution of the southern hemisphere during northern summer.

As expected, a large portion (more than 60%) of the soil N\textsubscript{2}O fluxes have tropical origin (23.5 S to 23.5N), while emissions from cooler regions are limited by temperature and arid/semi-arid regions by soil water. Our modelling results suggest year-round high emission rates from humid zones of Amazonia, east central Africa, and throughout the islands of Southeast Asia, with small seasonal variations (Fig. 2). Emissions from tropical savannah are highly variable, with locations of both high fluxes (seasonal mean > 30 mgN m\textsuperscript{-2} month\textsuperscript{-1} or 3.6 kg ha\textsuperscript{-1} yr\textsuperscript{-1}) and low fluxes (seasonal mean < 1.3 mgN m\textsuperscript{-2} month\textsuperscript{-1} or 0.16 kg ha\textsuperscript{-1} yr\textsuperscript{-1}). The simulated average tropical emission rate is 0.78 kgN ha\textsuperscript{-1} yr\textsuperscript{-1} (1970-2005), within the range of estimates (0.2-1.4 kgN ha\textsuperscript{-1} yr\textsuperscript{-1}) based on site-level observations from the database of Stehfest and Bouwman (2006), but smaller than a more detailed simulation study (1.2 kgN ha\textsuperscript{-1} yr\textsuperscript{-1}) carried out by Werner et al. (2007). Our analysis here excluded land cover, land use changes and human management impacts, while most of the observation-based or regional modelling studies did not factor out those impacts. Our modelling result in natural tropics is comparable with another global modelling study (average emission rate, 0.7 kgN ha\textsuperscript{-1} yr\textsuperscript{-1}) (Zaehle et al., 2010), in which the authors claimed they may underestimate the tropical N\textsubscript{2}O sources compared to the inversion estimates from the atmospheric transport model TM3 (Hirsch et al., 2006).

### 3.2 Sensitivity to WFPS

The different parameterization of WFPS and the use of different soil moisture modeling and data allows to test the sensitivity of soil N\textsubscript{2}O emissions to variable WFPS. Globally, emissions generally increase with WFPS (Fig. 3). WFPS derived from Method 1 is higher than that based on Method 2. Data-derived soil moisture datasets combined with different calculation methods together produced a range of 0.15-0.72 for the global mean WFPS (1982-2005). While mean
values greater than 0.6 (approximately field capacity) are less realistic, these high WFPS values provide the opportunity to test the model’s response to the soil moisture-based parameterization of redox conditions in soils. Global soil \( \text{N}_2 \)O emissions are highly sensitive to WFPS, with approximately 0.25 TgN per year per 0.01 change in global mean WFPS in the range 0 to 0.6. The spatial and temporal characteristic of WFPS also matters. Emission rate from LM3V-SM (Fig. 3 green cycle) is 1.13 TgN yr\(^{-1}\) higher than that from NOAH-SM (Fig. 3 blue triangle), while both model configuration have the same mean WFPS (ca. 0.21), highlighting effects of regional and temporal differences between the soil moisture products.

3.3 Model-observation comparisons

Modelled \( \text{N}_2 \)O emissions capture the average of cross-site observed annual mean emissions (0.54 vs. 0.53 kgN ha\(^{-1}\) yr\(^{-1}\) based on LM3V-SM) reasonably (Appendix A and Fig. 4a), but spread considerably along the 1:1 line. The points deviating the most are from tropical forests, with overestimations from montane tropical forest and underestimations from lowland tropical forests if those measurements are representative of gridcell emissions. These patterns are similar as results from NOAH-SM (Appendix A and Fig. 4b) and ERA-SM (Appendix A and Fig. 4c), except that the application of WFPS from NOAH-SM slightly underestimates the observed global mean (0.54 vs. 0.47 kgN ha\(^{-1}\) yr\(^{-1}\) from NOAH-SM with WFPS based on Method 2).

At the Tapajós National Forest, results from LM3V-SM capture some of the variations in \( \text{N}_2 \)O fluxes, but the model is not able to reproduce the high emissions observed during spring (Panel (a), Fig. 5), which might be caused by the underestimation of WFPS from models. We used a total porosity of 0.62 (Davidson et al., 2004) to estimate root zone WFPS based on the reported soil water content (Davidson et al., 2008). The average WFPS from observation is estimated to be 0.49, which is higher than the modelled average of root zone WFPS for all 3 model configurations (LM3V-SM 0.27, NOAH-SM 0.30, and ERA-SM 0.37). WFPS varies between < 0.05 and 0.45 in LM3V-SM (range from 0.20 to 0.36 in NOAH-SM and 0.30 to 0.41 in ERA-SM), and contrasts with observation that show seasonal variations with WFPS in the range of 0.37 to 0.58. At the Hubbard Brook Experimental Forest, the correlations between model results and observations are 0.51 (LM3V-SM), 0.56 (NOAH-SM) and 0.62 (ERA-SM) for yellow birch, 0.66 (LM3V-SM), 0.68 (NOAH-SM) and 0.70 (ERA-SM) for sugar maple. However, the model is less robust in reproducing the magnitude of emission peaks. Groffman et al. (2006)
suggested high emissions of N$_2$O in winter were associated with soil freezing. However, the model assumes little emissions when soil temperature is under 0 °C. In addition, observations suggested N$_2$O uptake (negative values in Panel (b), Fig. 5) while the model does not incorporate mechanisms to represent N$_2$O uptake. At the Oita cedar forest, LM3V-N reproduces the seasonality of N$_2$O emissions accurately (Panel (c), Fig. 5). ERA-SM overestimates the magnitude of N$_2$O fluxes from Inner Mongolia grassland, while the magnitudes produced from LM3V-SM and NOAH-SM are comparable with observations. However, the timing of the emission peaks are one or two month in advance in the model compared to observations (Panel (d), Fig. 5). WFPS at a nearby meteorological station fluctuated between 0 and 0.5 for 0-20cm depth (Xu-Ri et al., 2003) which agrees with our values based on LM3V-SM and ERA-SM, but the range is lower for NOAH-SM (0.05 to 0.35). At the specific plots, Xu-Ri et al. (2003) reported a mean WFPS of 0.32 in one plot (LC) and 0.20 from in the other plot (SG) for the 0 to 0.1 m depth interval which are close to ERA-SM and NOAH-SM (LM3V-SM 0.14, NOAH-SM 0.19, ERA-SM 0.30), however, no temporal information was provided for the specific sites. At the Fukushima cedar forest, similar as at the Oita cedar forest, models are less robust at capturing the magnitude of high peaks of N$_2$O emissions although the seasonality produced by the model are good (Panel (e), Fig. 5). Emissions in the primary and secondary tropical rainforest at the Pasir Mayang Research Site are highly variable, which makes the comparison difficult (Panel (f), Fig. 5). LM3V-SM (but not ERA-SM and NOAH-SM) reproduces the low emissions in September-November 1997 and the increase of emissions from secondary forests in December, 1997. Overall, modeled variability is smaller compared to observation across these sites.

The strong variability of measured N$_2$O emissions is further illustrated in Fig. 6. Differences in measured N$_2$O fluxes between different forest sites are large, reflecting heterogeneity that is not captured within one model grid cell. In addition, the error bars, which represent the standard deviation of measured N$_2$O fluxes at three different plots of the same forest, are large. The standard deviation is as high as 49.27 μgN m$^{-2}$h$^{-1}$, indicating the strong variability of measured N$_2$O fluxes at the plot scale. Modeled N$_2$O fluxes are generally within the range of measured N$_2$O emissions. Model outputs slightly underestimate N$_2$O emissions largely due to the underestimation of soil water content (Panel (b) Fig. 6).
3.4 Sensitivity to N cycling processes and parameterization

Disallowing N losses through DON and fire volatilization enhance ecosystem N accumulation and availability to plants and microbes, and therefore increases N$_2$O emissions (Panel (a), Fig.7). The gain in N$_2$O emissions from disallowing DON loss is small (0.12 TgN yr$^{-1}$). However, N$_2$O emission is on average (1950-2005) increased by 3.63 TgN yr$^{-1}$ in the absence of fire volatilization N loss (we note, that fires do occur, but N is retained as ash in the litter). The gain is most evident in tropical regions (not shown), indicating the importance of fire in regulating ecosystem N status. Simulated preindustrial BNF is smaller than the empirical reconstructed BNF (72 in LM3V-N vs. 108 TgN yr$^{-1}$ from empirical based data Green et al., 2004). However, BNF in LM3V-N increases with time under historical varying climate, increasing atmospheric CO$_2$ level and N deposition. The global average BNF during simulated for the period 1950-2005 is 100 TgN yr$^{-1}$, close to the empirical value. Nevertheless, substitution of BNF in LM3V-N by empirical preindustrial value increased N$_2$O flux by 1.2 TgN yr$^{-1}$ (Panel (a), Fig.7).

Among the specific parameters tested, N$_2$O emission is most sensitive to the 10 times change (x10) of the fraction of net nitrification lost as N$_2$O gas. The relative magnitude of N$_2$O flux on average (1950-2005) reaches 6.5 times of the default (Panel (b), Fig.7). Reduction (x0.1) of maximum active plant N uptake strength (v$_{max}$) strongly increases N$_2$O emissions (ca. by 3 times of the default). Meanwhile, enhancement of v$_{max}$ also increases N$_2$O fluxes, reflecting the non-linear response of N$_2$O emissions to v$_{max}$. x10 increase in the maximum nitrification rate $k_n$ and denitrification rate $k_d$ increase N$_2$O emissions, while x0.1 decrease N$_2$O flux. N$_2$O increases more with increasing $k_d$ than with increasing $k_n$, whereas reduction of $k_n$ (x0.1) produces a stronger response than reduction of $k_d$. The half-saturation constant that represents the regulation of labile carbon availability on denitrification rate, $K_c$, is the least sensitive parameter. Meanwhile, reduction (x0.1) of the half-saturation constant $K_n$ that represents the regulation of substrate availability on denitrification rate on average increased N$_2$O fluxes by 4.5 TgN yr$^{-1}$ (Panel (b), Fig.7).

3.5 CO$_2$ and temperature responses

Globally, N$_2$O emissions respond to a step CO$_2$ increase first with a decline to ultimately increased levels after approximately 40 years (Fig. 8a, black line). The simulated global response follows largely the behaviour as simulated for tropical forests (Fig. 8a, yellow line). The shift from a negative to a positive response indicates possible competing mechanisms.
operating on different time scales. Field level experiments revealed the highly variable effects of CO₂ fertilization on N₂O emissions. Based on a meta-analysis, van Groenigen et al. (2011) suggested that elevated CO₂ significantly increased N₂O emission by 18.8%, while Dijkstra et al. (2012) argued for a non-significant response in non-N-fertilized studies. In contrast to observation studies, the global C-N cycle model analyses from O-CN suggested negative CO₂ fertilization effects on N₂O emissions (Zaehle et al., 2011). The negative impacts (reduced N₂O flux), which are also reported in manipulative experiments, are likely from increased plant N and immobilization demand under CO₂ fertilization, reducing N availability for nitrifiers and denitrifiers (Dijkstra et al., 2012). CO₂ fertilization on average (over 100 years) increased the global mean plant nitrogen uptake rate by 10.02 kgN ha⁻¹ yr⁻¹, as shown in Fig. 9 (Panel (b)).

Modelled soil inorganic N content (ammonium and nitrate) is reduced at first, but the reduction is not sustained. One mechanism to alleviate CO₂ fertilization induced N limitation is through BNF, which is on average (over 100 years) more than doubled (Fig. 9 Panel (e)). Similar to manipulative field experiments (Dijkstra et al., 2012), positive effects (increase N₂O fluxes) can result from the impacts of elevated CO₂ level to increase litter production (Fig. 9 Panel (a)) and consequently C sources for denitrifiers, and to increase soil moisture (Fig. 9 Panel (d)) from reduced stomatal conductance and leaf transpiration (Fig. 9 Panel (c)). With both positive and negative mechanisms embedded in our model, the net effects depend on the relative strength of the opposing forces.

Temperate deciduous forests, where most of the forest CO₂ fertilization experiments are conducted, respond positively to elevated CO₂ levels (Fig. 8a, green line). The slight increase in modelled N₂O emission are comparable with the mean response of field data compiled for temperate forests (ca. 0.01-0.03 kgN yr⁻¹ ha⁻¹) (Dijkstra et al., 2012). A similar positive response was detected for cold evergreen forests (Fig. 8a, pink line) with stronger magnitude compared to temperate deciduous forests. For grasslands, Dijkstra et al. (2012) reported small negative mean response from northern mixed prairie (ΔN₂O, ca. -0.01 to -0.03 kgN yr⁻¹ ha⁻¹), zero mean response from shortgrass steppe and positive mean response from annual grassland (ca. 0.03-0.06 kgN yr⁻¹ ha⁻¹). Our model shows a small negative mean response from C4 grassland (Fig. 8a, cyan line) with the similar magnitude of that reported for the Northern mixed prairie, where the composition of C4 grass varies (Dijkstra et al., 2012). A CO₂ increase in C3 grassland initially reduces N₂O emission (Fig. 8a, blue line). However, this slight negative response turns into a small positive within one decade.
Elevated temperature generally increases N$_2$O emissions except for the slight negative effect in C4 grass (Fig. 8b). Overall the response to a 2 degree warming is bigger than that of doubling of CO$_2$. The simulated temperature effects are more pronounced in the first decade and decrease over time in tropical forests (Fig. 8b, yellow line), while for the temperate deciduous forests (Fig. 8b, green line) and boreal forests (Fig. 8b pink line), the temperature effects become more pronounced over time. Simulated temperate forest response (in the first decade) is close to that of observed mean (ca. 0.2-0.5 kgN yr$^{-1}$ ha$^{-1}$) (Dijkstra et al., 2012). Our modelled slight negative response in C4 grass and positive in C3 grass are in alignment with data compiled by Dijkstra et al. (2012) who reported both positive and negative responses in grasslands.

The results of combining CO$_2$ and temperature are similar to the CO$_2$ effect alone (Fig. 8c), despite the fact, that the individual effect of temperature is much stronger than that of CO$_2$. This antagonistic interaction (i.e. the combined enhancement in N$_2$O flux from elevated CO$_2$ and temperature are smaller than the summary of their individual effects) is also evident for C3 grass (first 50 years), temperate deciduous tree and cold evergreen forests (Fig. 8d).

4 Discussion

Our model combines two of the most widely applied biogeochemical models (DNDC and CENTURY) with current advancements in field level studies. The model was capable of reproducing the global mean natural N$_2$O emissions in other modeling and inverse methods, and the average of observed cross-site annual mean behavior. By focusing on the role of soil moisture in N$_2$O emissions, we found on a global scale a high dependence of simulated N$_2$O emissions on soil moisture (WFPS), mainly driven by emissions from tropical regions. The model broadly reproduced the magnitude and direction of responses to elevated CO$_2$ and temperature from manipulative field experiments where data is available. Both the global total emission as well as the global response to temperature and CO$_2$ increase followed largely the response of tropical forests, where field experiments are rare and no evaluation of CO$_2$ responses exist.

Soil moisture is a key variable in climate system but difficult to derive or measure at the global scale (Seneviratne et al., 2010). Our modelled fluxes were highly sensitive to WFPS, which is in agreement with observation and model synthesis studies (Heinen, 2006; Butterbach-Bahl et al., 2013). The large range when calculating WFPS from different methods resulted in a difference of more than 5 TgN yr$^{-1}$ in global soil N$_2$O fluxes. Saikawa et al. (2013) found
an up to 3.5 TgN yr⁻¹ gap induced by different precipitation forcing data from CLMCN-N2O. It is difficult to single out the difference caused by soil moisture alone from their results. Nevertheless, those two studies did suggest our and previous results highlight the importance of improving the dynamics of soil water and representation of WFPS for the purpose of predicting soil N₂O emission and climate feedbacks.

The root zone soil water in LM3V-N is based on a single layer bucket model. This simplified treatment of soil water dynamics may increase the difficulty in reproducing the temporal and spatial dynamics of WFPS. As a first step, we used the average between the original analog in LM3V-N and a formulation that was derived from soil total porosity to account for actual soil moisture and the possibility of soil water above field capacity. Meanwhile, overriding soil moisture with data-derived products (NOAH-SM and ERA-SM) suggests that the most realistic average (1970-2005) soil N₂O emission is in the range of 5.61-7.47 TgN yr⁻¹. However, despite using data-derived soil moisture, it appears that the prediction of soil moisture is an impediment towards validating N₂O emissions at field scale, for both LM3V-N simulated and reanalysed soil moisture. If evaluated against field data, the model was capable of representing the mean across sites and to a certain degree also compared adequately with site-specific time series. However, across the models there are differences between sites, model and data (Fig. 4) and also particularly peak emissions were poorly represented in the model (Fig. 5), and they can at least partly be attributed to mismatches in WFPS. Overall, comparison against field data revealed that the model’s variability is smaller compared to observation for both across field sites (Fig. 4) and at different sites (Figs. 5 and 6). One of the reason for this shortcoming may be that fast transitions, such as freeze-thaw cycle (Groffman et al., 2006) and pulsing (Yienger and Levy, 1995) are not sufficiently captured.

Perhaps equally important to address in future analysis, is the tremendous variability of N₂O emissions from site to site within the same region (see Fig. 6) This field-scale variability highlights the complexity of microscale interactions for N₂O production, which creates notorious large spatial and temporal variabilities and are undoubtedly difficult to constrain even at the stand level (Butterbach-Bahl et al., 2013). The homogeneous representation of environmental drivers within model grid cells casts doubt on site-specific model-observation comparison in global simulations. For example, N₂O emissions vary with topography which are not treated explicitly in most of the global C-N models. 3.8 times difference was detected in a montane forest (Central Sulawesi, Indonesia) moving from 1190 m to 1800m (Purbospito
et al., 2006), and 4.3 times difference was found from a tropical moist forest (Brazilian Atlantic Forest) with the altitude changing from 100m to 1000m (Sousa Neto et al., 2011).

Globally, N\textsubscript{2}O emissions from nitrification-denitrification were similar to O-CN and LPJ-DyN as they are all derived from DNDC’s formulation. (Xu-Ri et al., 2012; Zaehle et al., 2011).

Embedding an established N\textsubscript{2}O emission module into LM3V-N enables evaluation of the response of N\textsubscript{2}O emissions under different assumptions across models with respect to the dynamics of the larger plant-soil N cycle. Generally higher inputs from BNF or restriction of losses through organic N (fire, DON) enhance N\textsubscript{2}O emissions. The representation of BNF in models requires improvement, but currently, simple empirical relationships are used, yet BNF is the largest source of new N in terrestrial systems, and therefore is critical in the determination of N availability to nitrifiers and denitrifiers. Here we showed that different implementations of BNF (prescribed vs. responsive to plant N demand) are globally important for N\textsubscript{2}O emissions. Similarly, the magnitude of N lost through fire impacted N\textsubscript{2}O emissions in fire prone regions, while N emission factors are poorly constrained globally (Andreae and Merlet, 2001). The strength of plant uptake of N posed a strong constraint on the availability of N for nitrification-denitrification losses as it can draw down N substantially (Gerber and Brookshire, 2014). A reduction of plant uptake strength allows for relatively more N allocated for denitrification. More surprising was the positive effect of a stronger plant uptake capacity on N\textsubscript{2}O emissions: Enhanced plant uptake allow increased vegetation production and N throughput through litterfall and mineralization in the long run, which ultimately may allow higher N\textsubscript{2}O losses. In addition to those N cycling processes, N\textsubscript{2}O emissions were highly sensitive to the fraction of N lost as N\textsubscript{2}O during net nitrification. The fraction of N\textsubscript{2}O lost during net nitrification is uncertain. Goodroad and Keeney (1984) suggested a value of 0.1-0.2%, while Khalil et al. (2004) reported a range of 0.16%-1.48% depending on the O\textsubscript{2} concentration. We applied a global constant of 0.4% in our default simulation, bearing in mind the large uncertainties associated with this parameter.

Our results showed that tropical forests play a major role in both rates of emission and responses to perturbations. Tropical forests contributed with more than 60% to the global soil N\textsubscript{2}O fluxes. El Niño events triggered reduced soil N\textsubscript{2}O emissions that are in our simulations similar to earlier estimates (Saikawa et al., 2013; Thompson et al., 2014). El Niño events are known to have induced several of the most well known large scale droughts and altered soil moisture dynamics (Schwalm et al., 2011) in the tropics. Tropical forest N\textsubscript{2}O emissions were highly correlated with...
root zone soil water content and contributed strongly to the global-scale fluxes of N$_2$O in our model. Similarly, global responses to elevated CO$_2$ and temperature were dominated by the tropical response. In contrast to temperate and boreal forests, tropical forests responded negatively to elevated CO$_2$ in the first few decades. The initial negative response of N$_2$O emissions to CO$_2$ fertilization in tropical forests produced by LM3V-N stemmed largely from increased demand and uptake of mineral N due to enhanced vegetation growth under elevated atmospheric CO$_2$ level. Consequently, less N is available for gaseous losses as the N cycle tightens. If gross mineralization is used as an indicator of the rate of N flow in the “hole-in-the-pipe” concept and gaseous losses are proportional to mineralization, the initial negative response is unlikely to be detected. We found increased mineralization rate with increased litterfall under elevated CO$_2$, while N availability is reduced from LM3V-N. The mineralization based approach is likely to predict an increase of losses regardless of N limitation.

The marked decrease in our simulation for the tropical forests also contrasts somewhat findings from manipulative field experiments where CO$_2$ enrichment caused decrease, unchanged, or increase across extratropical ecosystems (Dijkstra et al., 2012; van Groenigen et al., 2011), whereas no empirical evidence is available in tropical forests. Overall, the marked differences between tropics and extratropics in the response to environmental forcing, and the large contribution of tropical forests to global N$_2$O emissions suggests caution when extrapolating field studies mostly carried out in extrapolar regions to the globe.

Based on single factor analysis with LM3V-N, the initial response of N$_2$O emission to a temperature increase was much larger than the response to increased atmospheric CO$_2$ (Fig. 8). However, we found large interactions between warming and CO$_2$ fertilization, in that the combined effect much more resembled the CO$_2$ effect alone. This interaction is the result of two antagonistic responses where a warming lead to increased N mineralization and potential N surplus, whereas a CO$_2$ increase fostered plant N demand that competed with microbial N$_2$O production. While these mechanisms are part of most models, both comparison against different models show notable differences when analyzing these two opposing effects. For example, Stocker et al. (2013) found that under future climate change scenarios, CO$_2$ and climate effects are amplifying each other, in accordance with earlier model experiments (Xu-Ri et al., 2012).

In LM3V-N we found that these interactions are negative. On the other hand, simulations with O-CN (Zaehle et al., 2011) showed the marginal effects of CO$_2$ and climate to be approximately equal and of opposite sign for historic simulations covering the
modeling setup of Zaehle et al. (2011) compare best with our past 300 years that also include land-cover changes. They evaluated the effect of climate change as the difference between a simulation that consider both CO$_2$ and climate and a simulation that does not consider climate change. Thus their climate effect contains both the single effect for CO$_2$, while for of climate, it is and the combination of interaction of climate with CO$_2$. The temperature and interaction (i.e. climate) response on top of CO$_2$ can in our simulation be calculated as the temperature effect plus the interaction effect (Fig. 8). Analyzed in this fashion, LM3V-N’s results are congruent with those of Zaehle et al. (2011), albeit we found a slightly weaker temperature effect compared to CO$_2$. This over time, the initial response then transitions into a much larger CO$_2$ effect, while the response to temperature diminishes. This long-term response of a positive CO$_2$ effect can be expected in a model that strongly retains N under limiting conditions such as LM3V-N. Retention ultimately allows build-up of N stocks, thereby alleviating limitation and increasing the substrate for nitrifiers and denitrifiers. This transition into a positive CO$_2$ response was likely facilitated by up-regualtion of BNF (Figure 9), which acts to reduce ecosystem N deficits and plant N demand in medium- to long-term. Up-regulation is expected to be much weaker or absent in models where BNF is parameterized based on evapotranspiration (Thomas et al., 2015). We realize that strong interactions as found here and elsewhere (Xu-Ri et al., 2012; Stocker et al., 2013) may also play out when other factors are considered (Brown et al., 2012), including N deposition, precipitation and land use change (disturbance). Therefore some of the discrepancy with other models may be caused by differences in the modeling setup. In addition, step changes in atmospheric CO$_2$ and temperature compared to gradual and sustained increases may also lead to differences. Yet applying step changes is useful to test our conceptual understanding and may help explain the discrepancy between the previous modeling study and meta-analysis of manipulative field experiments with regard to CO$_2$ fertilization responses (Zaehle et al., 2011; van Groenigen et al., 2011).

5 Conclusions

We present estimates of terrestrial soil N$_2$O fluxes under natural vegetation (1970 to 2005) based on existing N$_2$O emission formulations embedded into the global C-N cycle model LM3V-N. To determine the sensitivity of the modelling result to soil water (WFPS), we replaced the root zone soil water with two other derived datasets and altered the way in which WFPS is calculated. Our best estimate of modelled global soil N$_2$O flux is 5.61-7.47 TgN yr$^{-1}$
(1970-2005), within the range of current understanding of soil $\text{N}_2\text{O}$ emissions, but highly sensitive to WFPS, general N cycling and parameterization of $\text{N}_2\text{O}$ losses through nitrification and denitrification. Comparison against field experiments suggests that LM3V-N was able to capture mean values, although site-to-site and temporal mismatches remained. Given the sensitivity of $\text{N}_2\text{O}$ emissions to WFPS, improvements in soil hydrology are likely to improve soil $\text{N}_2\text{O}$ emission estimates. As expected, we found that processes in the model that alleviate ecosystem N limitation, such as reduced N losses through fire volatilization and increased N inputs through higher biological nitrogen fixation (BNF) rate, enhance $\text{N}_2\text{O}$ emissions. Responses to CO$_2$ and temperature perturbations showed differences compared to other models. In particular elevated CO$_2$ curbs $\text{N}_2\text{O}$ emissions sharply initially, but this negative response is alleviated after a few decades, likely in conjunction with fast N replenishment from up-regulated BNF. Our sensitivity analysis and the comparison with other models show that existing parameterizations of processes of the larger plant-soil N cycle affect fast N cycle processes such as evidenced by the response to the fire and BNF modification. This sensitivity can lead to differences in $\text{N}_2\text{O}$ across models (e.g. in the response to CO$_2$ and climate) even if existing nitrification-denitrification lead to distinct and new results if the larger plant-soil N cycle is treated differently. More importantly, schemes are identical. Further, our work suggests a much stronger response to warming and CO$_2$ in tropical forests, where few manipulative compared to extratropical forest, thus extrapolation of mostly extra-tropical field studies has been carried out to the globe warrants caution.
Appendix A: Observed annual N$_2$O fluxes data

Annual N$_2$O fluxes data were compiled from peer-reviewed literature. We applied simple selection criteria (see the main text) to reduce the mismatches between model outputs and field measurements, bearing in mind the gaps between complex field conditions and idealized model forcings. Latitudes (Lat) and longitudes (Lon) in Table A1 are based on model grids.

Table A1 Observed annual N$_2$O emission data for model comparison

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Acknowledgements

The soil moisture data used in this study were acquired as part of the mission of NASA’s Earth Science Division and archived and distributed by the Goddard Earth Sciences (GES) Data and Information Services Center (DISC). We thank the European Centre for Medium-Range Weather Forecasts for providing the reanalysed soil moisture dataset and the Oak Ridge National Laboratory (ORNL) Distributed Active Archive Center (DAAC) for sharing N₂O observation and soil property dataset. We would like to thank Matthew J. Cohen, Patrick Inglett and Jeremy W. Lichstein for their constructive comments throughout the study. We would also like to thank Lex Bouwman, Benjamin Stocker and an anonymous reviewer for constructive comments and suggestions.

References


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Figure 1. Simulated annual global soil N$_2$O emissions based on potential vegetation (1970-2005). Shaded grey area indicates El Niño years with the annual multivariate ENSO index (MEI) greater than 0.6. Colours refer to different soil moisture dataset used in the estimation: red for LM3V-SM (with WFPS calculated by Method 3); blue for NOAH-SM (Method 2) and green for ERA-SM (Method 2). Details for these soil moisture dataset and WFPS calculating methods is available in the main text.
Figure 2. Global seasonal mean soil N$_2$O emissions (with potential vegetation) averaged over the years 1970-2005. DJF (December, January and February), stands for Northern Hemisphere Winter; MAM (March, April and May) for Spring; JJA (June, July and August) for Summer; and SON (September, October and November) for Autumn.
Figure 3. Sensitivity of simulated global soil N\textsubscript{2}O emissions (with potential vegetation) to water filled pore space (WFPS). The x-axis is the WFPS averaged globally over 1982-2005; the y-axis represents the corresponding global total N\textsubscript{2}O fluxes. A total of nine sets of WFPS are obtained through either different soil water datasets (colours) or varied calculation methods (symbols). Maximum water, porosity and average correspond to method 1, method 2 and method 3 in the main text, respectively. Coloured symbols represent means and error bars indicate interannual standard deviations.
Figure 4. Observed vs. simulated annual N$_2$O emissions from natural soils. Dashed green lines are the 1:1 lines. The solid circles represent the overall means. Different panels represent simulations with different soil moisture data: (a) LM3V-SM (simulated by LM3V-N); (b) NOAH-SM (based on land surface model NOAH 3.3 in Global Land Data Assimilation System Version 2); and (c) ERA-SM (reanalysis data from ECMWF). Water filled pore space (WFPS) is calculated using the average of the one based on available water capacity and the one based on the total porosity (Method 3, see the main text for detailed description) for panel (a); and using the total porosity (Method 2) for panel (b) and (c).
Figure 5. Observed vs. simulated monthly N\textsubscript{2}O emissions at (a), the Tapajós National Forest in east-central Amazonia (3°S, 55°W), taken from Davidson et al. (2008); (b), the Hubbard Brook Experimental Forest in New Hampshire, USA (44°N, 72°W), taken from Groffman et al. (2006); (c), a cedar forest at Oita, Japan (33°N, 131°E), taken from Morishita et al. (2007); (d), the *Leymus chinensis* (LC) and *Stipa grandis* (SG) steppe in Inner Mongolia, China (44°N, 117°E), taken from Xu-Ri et al. (2003); (e), a cedar forest in Fukushima, Japan (37°N, 140°E), taken from Morishita et al. (2007); and (f), the primary (P1 and P2) and secondary (L1 and L2) forests located at the Pasir Mayang Research Site, Indonesia, taken from Ishizuka et al. (2002) (1°S, 102°E). Shown are modeled results from three WFPS schemes (LM3V-SM, NOAH-SM and ERA-SM) the same as in Figure 4.
Figure 6. Comparison of (a) soil temperature (2cm from observation and 1 cm from model) in °C; (b) soil moisture (2cm from observation and root zone from model) in % and (c) soil N$_2$O emissions in µgN m$^{-2}$ h$^{-1}$ from observations and model outputs at four forest sites from Germany (50°N, 8°E), taken from Schmidt et al. (1988). Shown are modeled results from two WFPS schemes (LM3V-SM and NOAH-SM) similar as in Figure 4.
Figure 7. Changes in simulated global average N₂O (1950-2005) emissions from modifying general N cycling processes (a) and model parameters one-at-a-time (b). Altered processes include disallowing N losses through dissolved organic matter (DON in (a)) and fire volatilization (Ash in (a)), and replacing simulated biological N fixation with preindustrial N fixation rate (BNF in (a)). Parameters include: \( V_{\text{max}} \), the maximum active N uptake rate per unit root biomass; \( k_n \), the optimum nitrification rate; \( k_d \), the optimum denitrification rate; \( K_c \) and \( K_n \), the half saturation constants for labile C availability and nitrate respectively; and \( \text{frac} \) is the fraction of net nitrification lost as N₂O. Parameters are either increased by multiplying 10 (lightblue) or reduced by multiplying 0.1 (lightgreen) relative to the defaults.
Figure 8. Soil N₂O emissions in response to step increases in atmospheric CO₂ and temperature.
Panel (a) is the response to CO₂ fertilization alone, expressed as the difference between CO₂ increased run and the control run (CO₂_FERT - CONTROL), the inset zooms into the y axis (flux difference) around zero; Panel (b) is the response to temperature increase alone (TEMP_CONTROL); Panel (c) is the combined response to both CO₂ enrichment and temperature rise (CO₂_FERT×TEMP_CONTROL); and Panel (d) is the interactive effect of CO₂ and temperature responses, which is the difference between the combined (results from Panel (c)) and minus the individual responses (results from Panel (a) and (b)). Results are shown as annual values (thin dashed lines) and as running average with a moving window of 17 years (period of recycled climate forcing, thick solid lines). The black lines represent the global average response. Coloured lines indicate responses for biome as represented by each plant functional type (PFT) considered in LM3V-N: C4 grass (cyan), C3 grass (blue), tropical forest (yellow), temperate deciduous forest (green) and cold evergreen forest (pink). Dashed red line represents the zero line.
Figure 9. CO₂ fertilization effects (no temperature change) on litter pool size (Panel (a)), plant nitrogen uptake rate (Panel (b)), canopy transpiration rate (Panel (c)), soil water content in the root zone (Panel (d)) and biological nitrogen fixation (BNF) rate (Panel (e)). Shown are the 100-year average of global means (spatial) for control (284 ppm, red) and with elevated CO₂ (568 ppm, blue).

Table 1 Texture dependent parameter $k$, which partitions N₂O/N₂ gas fractions during denitrification, estimated from Del Grosso et al. (2000)

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