

Abstract

 Microbial decomposition of plant litter is a crucial process for the land 20 carbon (C) cycle, as it directly controls the partitioning of litter-C between $CO₂$ released to the atmosphere versus the formation of new soil organic matter (SOM). Land surface models used to study the C cycle rarely considered flexibility in the 23 decomposer C use efficiency (CUE_d) defined by the fraction of decomposed litter-C that is retained as SOM (as opposed to be respired). In this study, we adapted a 25 conceptual formulation of CUE_d based on assumption that litter decomposers 26 optimally adjust their CUE_d as a function of litter substrate C to nitrogen (N) stoichiometry to maximize their growth rates. This formulation was incorporated into the widely used CENTURY soil biogeochemical model and evaluated based on data from laboratory litter incubation experiments. Results indicated that the CENTURY model with new CUE_d formulation was able to reproduce differences in respiration rate of litter with contrasting C:N ratios and under different levels of mineral N 32 availability, whereas the default model with fixed CUE_d could not. Using the model 33 with flexible CUE_{d} , we also illustrated that litter quality affected the long-term SOM formation. Litter with a small C:N ratio tended to form a larger SOM pool than litter with larger C:N ratios, as it could be more efficiently incorporated into SOM by microorganisms. This study provided a simple but effective formulation to quantify the effect of varying litter quality (N content) on SOM formation across temporal scales. Optimality theory appears to be suitable to predict complex processes of litter decomposition into soil C, and to quantify how plant residues and manure can be harnessed to improve soil C sequestration for climate mitigation.

 Keywords: microbial carbon use efficiency, litter decomposition, litter stoichiometry, soil organic matter, litter decay model, nitrogen

1 Introduction

 Plant litter decomposition plays a key role in global carbon (C) cycle, thus needs to be well represented in land surface models. The decomposition and transformation processes of plant litter control the formation of soil organic matter (SOM) (Prescott, 2010; Schmidt *et al.*, 2011; Walela *et al.*, 2014; Cotrufo *et al.*, 2015) and associate immobilization and mineralization of essential plant nutrients (Moorhead and Sinsabaugh, 2006; Parton *et al.*, 2007; Manzoni *et al.*, 2008; Manzoni and Porporato, 2009). Hence, a reliable litter decay model is necessary for estimating soil C balance and turnover of ecosystem C (Allison, 2012; Bonan *et al.*, 2013; Wieder *et al.*, 2013; Campbell and Paustian, 2015). In particular, a realistic representation of litter decomposition in land surface models is helpful to decrease the uncertainties in predicted effects of climate change and anthropogenic management on ecosystems (Gholz *et al.*, 2000; Campbell and Paustian, 2015; Luo *et al.*, 2016). As litter decomposition is a very complex process determined by climate (e.g. temperature and moisture), litter quality (e.g. nitrogen (N) concentration), soil nutrients and the physiological characteristics of microorganisms (Lekkerkerk *et al.*, 1990; Prescott, 2010; Manzoni *et al.*, 2012; Frey *et al.*, 2013; Sinsabaugh *et al.*, 2013; García-Palacios *et al.*, 2016), there remain large uncertainties in existing litter decay models (Zhang *et al.*, 2008; Bonan *et al.*, 2013; Campbell and Paustian, 2015). Many litter decay models, especially those incorporated in global land surface models, have ignored stoichiometric constraints to microbial processes (Bonan *et al.*, 2013; Cotrufo *et al.*, 2013; Wieder *et al.*, 2013; Wieder *et al.*, 2014). Microbial carbon use efficiency (CUE),defined as the ratio of microbial

 biomass production to material uptake from substrates (Lekkerkerk *et al.*, 1990; Manzoni *et al*., 2012), is an important parameter constraining litter decay, but it has rarely been represented as a flexible quantity in land surface models. During litter decomposition, only a part of the decomposed litter-C is being transferred into SOM, 72 while the remaining C is being released as $CO₂$ to the atmosphere by microbial respiration (Campbell and Paustian, 2015; Cotrufo *et al*., 2015). While CUE is a

 physiological property of each decomposer community, it also determines the ecosystem-level efficiency at which litter C is transferred into SOM a step further from simple microbial incorporation. We denote this efficiency as carbon use efficiency of 77 litter decomposition (CUE_d). With higher CUE_d , more plant-produced litter is transformed biologically into SOM, and soil C storage can reach higher values (Six *et al.*, 2006; Sinsabaugh *et al.*, 2013). In most existing soil biogeochemical models, CUE^d is assumed to be same as microbial CUE and considered as a fixed parameter. 81 The Verberne model (Verberne *et al.*, 1990) assumes for instance CUE_d ≈ 0.25 . In the 82 Yasso model (Liski *et al.*, 2005), the CUE_d is set to 0.2. The CENTURY model sets 83 the CUE_d for decomposition of surface and belowground metabolic litter to 0.55 and 0.45, respectively (Parton *et al.*, 1988). In Daisy (Hansen *et al.*, 1991), NCSOIL 85 (Molina *et al.*, 1983) and ICBM (Kätterer and Andr $\acute{\text{m}}$, 2001), CUE_d = 0.6 for the labile litter pools and takes a lower value for recalcitrant substrates. Only a few models account for variable CUE, letting it vary in response to substrate stoichiometry (Schimel and Weintraub, 2003) or temperature (Allison *et al*., 2010). The increasing evidence for a variable microbial CUE leads to a conceptual CUE model which can explain trends in CUE of microorganisms along stoichiometric 91 gradients (Manzoni *et al.*, 2017). The values of CUE_d used in existing litter decay models are mostly derived from CUE obtained in laboratory studies on microbial physiology or limited observations in certain ecosystems, thus show large variations without a dynamic link to environmental conditions (Parton *et al.*, 1988; Verberne *et al.*, 1990; Hansen *et al.*, 1991; Liski *et al.*, 2005; Manzoni *et al.*, 2012). Recent studies (Manzoni *et al.*, 2008, 2012) suggested that the microbial CUE in terrestrial ecosystems ranges from less than 0.1 for wood decomposers to about 0.5 for decomposition of N-rich and high-quality litter. To explain those differences, Manzoni *et al*. (2017) proposed a conceptual model of microbial CUE based on the assumption that decomposers adapt their metabolism (and hence CUE) to maximize their growth rate. This model based on optimality theory links CUE to substrate and decomposers stoichiometry, where the optimal CUE decreases with increasing substrate C-to-nutrient ratio, and increases with soil nutrient availability. The predictions of this

 theoretical model have been verified by empirical evidence from CUE estimates for different microorganisms in both aquatic and terrestrial ecosystems (Manzoni *et al.*, 2017).

107 Besides variable CUE_d, many previous studies have also indicated the necessity for litter decomposition models to consider soil mineral N availability as a driver of litter decomposition rates, in particular under low N availability (Wieder *et al.*, 2015; Luo *et al.*, 2016; Averill and Waring, 2018). Microbial biomass is nearly homeostatic (Cleveland and Liptzin, 2007; Franklin *et al.*, 2011; Allison, 2012). When the supply of N from substrates is lower than the demand of microbes to fulfill their specific stoichiometric C:N ratio, microbes will utilize mineral N (immobilization) (Manzoni *et al.*, 2012). Thus, low availability of mineral N can limit microbial activity, and in turn litter decay rate (Manzoni and Porporato 2009; Fujita *et al.*, 2014). Although there are fertilization experiments reporting insignificant or even negative impacts of added N on litter decay rate (Fog, 1988; Hobbie and Vitousek, 2000; Finn *et al.*, 2015), many incubation experiments showed a significant decrease of litter decomposition rate with declining mineral N availability (Recous *et al.*, 1995; Hobbie and Vitousek, 2000; Guenet *et al.*, 2010). Moreover, recent modeling studies have indicated that including the limiting effect of low mineral N on decomposition improved predictions of C and N fluxes (Bonan *et al.*, 2013; Fujita *et al.*, 2014). 123 Therefore, soil mineral N can alter litter C flux by affecting both the litter decay rate 124 and the partition of decayed litter-C (via flexible CUE_d). Some detailed microbial decomposition models actually have included variable microbial CUE and the limitation of low mineral N availability on litter decay rate (Ingwersen *et al.*, 2008; Pagel *et al.*, 2013; Campbell *et al.*, 2016; Huang *et al.*, 2018); however, the parameterization and evaluation of these models pose significant challenges due to their complexity and limited verification data (Wieder *et al.*, 2014; Campbell and Paustian, 2015). There is still scope for implementing the

effects of litter stoichiometry and soil mineral N availability on litter decomposition in

litter decay models with more generalizable structure. In particular, it is important to

test the role of these effects in models that have already been incorporated into land

 surface model for long-term and large-scale applications (e.g. CENTURY, Parton *et al.*, 1988). In this study, we incorporated flexible CUE_d based on substrate C:N ratios and mineral N limitations into a soil biogeochemical model based on the CENTURY equations to simulate the decomposition and transfer processes of litter-C. The study was organized as follows. First, the new model was calibrated and tested against data from laboratory litter incubation experiments for its ability to capture the effect of variable litter quality and soil mineral N on litter respiration rates (short-term 141 simulations). Second, the model parameterized assuming flexible CUE_d and mineral N limitations was used to explore the consequences of such stoichiometric constraints on the production of soil organic carbon (SOC) (long-term simulations). With these two modeling analyses, we aimed at linking stoichiometric constraints acting on short-term (months to years) decomposition dynamics to their consequences on SOC accumulation occurring at decadal to centennial time scales.

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- **2 Materials and methods**

2.1 The CENTURY decomposition model

 The basis of the litter decay model used in this study is the CENTURY model (Fig. 1), a first-order decay model that describes decomposition as a function of substrate availability and quality, clay content, soil moisture and soil temperature (Parton *et al.*, 1988). Most land surface models (e.g. Kucharik *et al.*, 2000; Sitch *et al.*, 2003; Krinner *et al.*, 2005) adopted a similar structure to simulate the litter and soil biogeochemical processes. Dead organic matter in CENTURY is separated into structural and metabolic litter and three SOM pools (active, slow, passive) with different turnover times. There is no explicit representation of microbial biomass in CENTURY, instead the biomass of microbes is assumed to be in equilibrium with labile SOM and thus implicitly included in the active SOM pool. When C is being decomposed, a fraction of the decomposed C is respired to the atmosphere and the 161 remaining fraction $(CUE_d$ conceptually equal to microbial CUE) enters the acceptor SOM pool. Three of such fractions are defined to characterize the transfer of C from

163 litter to SOM: CUE_{ma} for transfer of the metabolic litter to the active SOM pool, and *CUE*_{sa} and *CUE*_{ss} for transfer of structural litter to active and slow SOM pools, respectively (Fig. 1).These fractions are set to be time invariant in the original version of CENTURY, so that a fixed fraction of decomposed C is retained in the acceptor pool regardless of environmental conditions and changes in the quality of the donor pool. The N flows in CENTURY follow the C flows and are equal to the product of C flow by the N:C ratio of the acceptor SOM pool. N mineralization is defined as the difference between N obtained from the donor pools and N stoichiometric demand of the acceptor pool (Parton et al., 1988; Metherell et al., 1993). In this way, net N mineralization occurs when the donor pool has low C:N ratio, but N is immobilized (taken up by microbes) when the donor pool has high C:N ratio.

2.2 Optimal CUE

 To quantify how microbial CUE varies along gradients of nutrient availability, it can be hypothesized that microorganisms maximize their growth rate, and hence their ecological competitiveness, by adapting resource (C and nutrients) use efficiencies. This follows the growth maximization hypothesis (Mooshammer *et al*., 2014; Manzoni *et al*., 2017). Based on this hypothesis, Manzoni *et al*. (2017) formulated a theoretical model expressing microbial CUE as a function of the stoichiometric difference between decomposers and their substrate. The CUE for which growth rate is maximized is the optimal CUE (*CUEopt*) given by:

184
$$
CUE_{opt} = CUE_{max} \times \min\left[1, \frac{CN_D}{CUE_{max}} \times \left(\frac{1}{CN_S} + \frac{I_N}{U_0}\right)\right]
$$
 (1)

185 where CUE_{max} is the maximum microbial CUE (dimensionless) when growth is 186 limited by C from the organic substrate. CN_D and CN_S are the C:N ratio (in mass, dimensionless) of decomposer and their substrate, respectively. Although Manzoni *et al*. (2017) indicated that mineral phosphorus (P) could also affect optimal CUE we 189 only considered N as a limiting nutrient. I_N (g N kg⁻¹ soil) is the maximum rate at 190 which mineral N can be taken up by microbes, and U_0 (g C kg⁻¹ soil) is the C-limited uptake rate (corresponding to the decomposition rate at optimal mineral N

 concentration). When litter C:N is low or soil mineral N is in excess, the second term 193 in the minimum function (Eq.(1)) is higher than one, and $CUE_{opt} = CUE_{max}$ (C limited conditions, as in nutrient-rich litter). In contrast, when mineral N is scarce, *CUEopt* decreases with increasing substrate C:N ratio (N limited conditions, N-poor litter). Lack of N in the organic substrates can be compensated by mineral N being immobilized by microorganisms from the soil solution. Immobilization meets the 198 nutrient demands as long as it is lower than the maximum supply rate I_N , at which point microbial CUE starts being down regulated. Thus, for any given C:N ratio in the substrate, *CUEopt* increases with inorganic N concentration in the soil solution until *CUE_{max}* is reached. It should also be noted that Eq. (1) is interpreted at the microbial community scale, not for individual organisms.

2.3 Adaption of the optimal CUE model in the CENTURY model

205 CUE of decomposition (CUE_d) is also assumed to be equivalent to microbial CUE in this study. Then we followed the theory from Manzoni *et al*. (2017) (Eq. (1)) 207 to parameterize CUE_d during litter decomposition into CENTURY (Fig.1). Due to the implicit representation of microbial growth in CENTURY, we replaced the original 209 optimality CUE model (Eq. (1)) by a simpler equation that involves the C:N ratios of the donor and acceptor pools, rather than microbial C:N ratios:

$$
211\\
$$

$$
CUE_{opt} = CUE_{max} \times \min\left[1, \left(\frac{CN_{lit}}{CN_{SOM}}\right)^a\right]
$$
 (2)

212 where CN_{lit} and CN_{SOM} are the C:N ratio (dimensionless) of litter (metabolic or structural) and SOM pools (active, slow or passive), respectively. The C:N ratio of SOM (around 9:1 on a mass basis in CENTURY) is representative of the decomposer biomass, its value being between the average C:N ratio of soil microbial communities including fungi and bacteria (7.4:1 in Cleveland and Liptzin, 2007) and the C:N ratio of soil fungi (13.4:1 in Zhang and Elser, 2017), which are probably largely responsible for fresh litter decomposition. *CUEmax* (dimensionless) is the maximum 219 CUE_d achieved when nutrients are not limiting, and it is set to 0.8 based on a synthesis of observed CUE of soil microbes (Manzoni *et al*., 2012). The exponent *a* (g

221 N kg⁻¹ soil) captures the effect of mineral N uptake by microbes on CUE_d . Because 222 CUE_d is expected to increase with mineral N availability (Eq. (1)), *a* is assumed to be 223 a linear function of the mineral N concentration $(N_{min}, g N kg^{-1})$ soil): 224 $a = m_1 \times (N_{min} - n_1)$ (3) 225 where m_l (kg g⁻¹ N) and n_l (g N kg⁻¹ soil) are two coefficients that need to be 226 calibrated. Eqs. (2) and (3) modulate the decrease in CUE_d with decreasing litter 227 quality when mineral N availability changes– the exponent *a* increases with 228 increasing mineral N availability, causing an increase in CUE_d at any given litter C:N 229 ratio. Hence, increasing *a* mimics an increase in I_N in Eq. (1). Fig. 2a illustrates how 230 CUE_d from Eq. (2) varies as a function of mineral N concentration, for different 231 values of litter C:N. 232 Eqs. (2) and (3) were implemented in CENTURY to modify the originally 233 fixed CUE_d (Fig. 1). With this change, the fractions of C from litter that remain in 234 SOM are mediated by stoichiometric constraints and mineral N availability, at the 235 expense of additional parameters to fit. The CUE_d for C transfers between SOC pools 236 (active, slow and passive) are not modified. 237 238 2.4 Constraint of soil nutrient availability on litter decomposition rate 239 CENTURY is a first-order decay model in which decomposition rates of 240 metabolic and structural litter are modulated by scaling factors of soil temperature 241 (*f*(*tem*)) and moisture (*f*(*water*)) (Parton *et al.*, 1988). Here, we introduced an 242 additional mineral N scaling factor $(f(N_{min}), 0-1,$ dimensionless) to account for the 243 limitation imposed by low mineral N availability on litter decay rate $(D(C_{lit}))$. 244 $D(C_{lit}) = C_{lit} \times k \times f(tem) \times f(water) \times f(N_{min})$ (4) 245 where C_{lit} is the C (g C kg⁻¹ soil) in litter pool (metabolic or structural) and *k* is the 246 potential maximum turnover rate $\text{(day}^{-1})$ at optimal soil temperature, moisture and 247 nutrient conditions. 248 In this study, we assumed that the scaling factor of mineral N increases 249 linearly with increasing soil mineral N concentration (*Nmin*, Eq. (5)) below a threshold

250 value of $1/m_2$ g N kg⁻¹ soil, where m_2 is a positive coefficient which needs to be

251 calibrated (Fig.2b). The inhibition effect of mineral N only occurs in case of

252 immobilization $(1/CN_{lit} < CUE_{opt}/CN_{SOM})$. The specific function $f(N_{min})$ can be 253 expressed as:

254
$$
f(N_{min}) = \begin{cases} \min[\mathbb{R}1, m_2 \times N_{min}) , & \frac{CUE_{opt}}{CN_{SOM}} - \frac{1}{CN_{lit}} > 0 \\ 1 , & \frac{CUE_{opt}}{CN_{SOM}} - \frac{1}{CN_{lit}} \le 0 \end{cases}
$$
(5)

 Existing studies have adopted approaches that differ from our definition to explicitly represent the N inhibition effects on microbial processes (Manzoni and Porporato, 2009; Bonan *et al.*, 2013; Fujita *et al.*, 2014; Averill and Waring, 2018). In 258 these previous studies, $f(N_{min})$ was assumed equal to the ratio between immobilized mineral N and the N deficit for maintaining a stable C:N of decomposer biomass or 260 other receiver pools. Using the notation of Section 2, this definition of $f(N_{min})$ can be expressed as:

262
$$
f(N_{min}) = \begin{cases} \min\left(1, \frac{m_3 \times N_{min}}{U_0 \times (\frac{CUE_{opt}}{C N_{SOM}} - \frac{1}{C N_{lit}})}\right), & \frac{CUE_{opt}}{C N_{SOM}} - \frac{1}{C N_{lit}} > 0\\ 1 & , \frac{CUE_{opt}}{C N_{SOM}} - \frac{1}{C N_{lit}} \le 0 \end{cases}
$$
(6)

263 where m_3 is a coefficient that needs to be optimized. U_0 (g C kg⁻¹ soil day⁻¹) is the C 264 uptake rate (equivalent to the litter decomposition rate in absence of leaching) when 265 soil mineral N is fully adequate for litter decay (i.e. $f(N_{min}) = 1$), and can be calculated 266 from Eq. (7) as:

$$
U_0 = C_{lit} \times k \times f(tem) \times f(water)
$$
 (7)

268 In this study, we also tested this formulation in the CENTURY-based model, in

269 addition to the Eq. (5) (see model M4 in Table A3).

270

271 2.5 Model parameterization and validation

272 To determine the respective impacts of including flexible CUE_d and N 273 availability constraining decay rates, we built four conceptual litter decay models 274 (Table 1). Model M0 corresponds to the default CENTURY parameterization of a 275 fixed CUE_d and no constraints of N availability on litter decay rates $(f(N_{min}) = 1)$. 276 Model M1accounts for flexibility in CUE from Eq. (2) and N constraints on decay 277 rates by Eq. (5). Model M2 has flexible CUE_d but no N constraints on decay rates $(f(N_{min})=1)$. Model M3 has N constraints on decay rates but a fixed CUE_d (Table 1). All of these four models are run at a daily time step. Finally, model M4 also accounts for flexibility in CUE and N constraints on decays (Table A3), but it uses Eq. (6) to represent the N constraints on decays rate rather than Eq. (2). Results from model M4 are presented in the main text, but only shown in the Appendix. This range of models allows identifying which mechanisms are at play during decomposition – flexible 284 CUE_d only (M3), mineral N limitation only (M2), both mechanisms (M1, M4), or none (M0).

 For calibrating model parameters and evaluation of their results, we collected data of laboratory litter incubation experiments from Recous *et al*. (1995) (5 experiments) and Guenet *et al*. (2010) (9 experiments, Table A2). The incubation experiments of Recous *et al*. (1995) and Guenet *et al*. (2010) continued 80 and 124 days, respectively. Recous *et al*. (1995) used corn residues (C:N = 130) and Guenet *et al.* (2010) used wheat straw $(C:N = 44)$ in their incubation experiments. The C:N ratios of those corn residue and wheat straw span the range of litter C:N ratios among different ecosystems (Harmon *et al*., 2009; Brovkin *et al*., 2012; Manzoni *et al*., 2010). In the incubation experiments, plant litter was firstly cut into fine fragments before it was mixed with mineral soil. Soil temperature and moisture condition were kept constant during the experiment. Respired C from the incubated litter and SOC, as well as the soil mineral N concentrations were measured continuously across the 298 incubation period. To distinguish the litter- and SOC-derived $CO₂$ flux, Guenet *et al.* (2010) used straw from wheat grown under ¹³C labeled CO₂ and they are therefore 300 able to track the CO_2 coming from litter and the CO_2 coming from soil. In the 301 experiments by Recous et al. (1995), litter-derived $CO₂$ flux is calculated as the 302 difference in $CO₂$ flux between the incubation samples with both soil and litter, and the control samples without added litter. More detailed information about the incubation experiments of Recous *et al*. (1995) and Guenet *et al*. (2010) can be found in Table A2.

The initial C storage and C:N ratios of litter and SOM pool, as well as soil

 temperature and moisture condition for decomposition in all of the five versions of the model (M0-M4) were set based on observations (Table A2). Plant litter was firstly separated into metabolic and structural litter pools based on its lignin to C ratio (*LC*_{lit}, dimensionless). The fraction of metabolic litter-C (*fm*, 0-1, dimensionless) is calculated by:

$$
f_m = f_{max} - m_4 \times L C_{lit} \tag{8}
$$

313 where m_4 is a coefficient to be calibrated; f_{max} =0.85 is the maximum fraction of metabolic litter (i.e., the default value in CENTURY; Parton *et al*., 1988). The 315 fraction of structural litter-C is thus $1-f_m$. The C:N ratios of both metabolic and structural pools are assumed to be equal to the C:N ratio of litter input.

 In M1 and M3 models, the observed mineral N concentrations across the incubation period were used to calculate the daily N inhibition effect (Eq. (5)). The 319 observed cumulative respired litter-C (g C kg⁻¹ soil) measured in the incubation experiments was used to calibrate the model parameter values. Moreover, to quantify 321 the simulated $CO₂$ flux derived from the litter, we also performed a set of control 322 simulations with only SOM (initial litter pools were set to 0 g kg^{-1} soil) using the four 323 model versions. The simulated litter-derived $CO₂$ flux is calculated as the difference in $CO₂$ flux between the simulation with both litter and SOM inputs and the simulation with only SOM input.

 Parameter calibration was performed for each model with the shuffled complex evolution (SCE) algorithm developed by Duan *et al*., (1993). The SCE algorithm relies on a synthesis of four concepts that have proved successful for global optimization: combination of probabilistic and deterministic approaches; clustering; systematic evolution of a complex of points spanning the space in the direction of global improvement and competitive evolution(Duan *et al.*, 1993). A more detailed description of this SCE optimization method can be found in Duan *et al*. (1993, 1994). In this study, the RMSE (root mean square error, Eq.(9)) between simulated and measured cumulative respired litter-C (%) on all observation days (Table A2) of each incubation experiment was used as the objective function, and the parameters minimizing RMSE between simulated and observed cumulative respired litter-C were

regarded as optimal parameter values.

$$
RMSE = \sqrt{\left(\frac{\sum_{i=1}^{n} (Sim_i - Obs_i)^2}{n}\right)}
$$
(9)

 where *n* is the number of observation days, *Simⁱ* and *Obsⁱ* (%) are the simulated and observed percent of cumulative litter-C flux on day *i*, respectively.

 We used leave-one-out cross-validation (Kearns and Ron, 1997; Tramontana *et al.*, 2016) to evaluate each of the four models (i.e. M0-M3), a cross validation method used when data is scarce. The number of cross-validations corresponds to the number of incubation experiments (14). Each time, one of the 14 incubation experiments was left out as the validation sample, and the remaining 13 experiments were used to train model parameters. In addition to RMSE, we also adopted the Akaike Information Criterion (AIC, Bozdogan, 1987, Eq. (10)) to determine the relative quality of the four version models on estimating cumulative respired litter-C.

$$
AIC = n \times ln\left(\frac{\sum_{i=1}^{n} (Sim_i - Obs_i)^2}{n}\right) + 2n_p \tag{10}
$$

350 where n_p is the number of model parameters. The evaluation of AIC is important here because depending on the model version, different numbers of parameters have to be determined (Table 1), requiring us to weigh both model accuracy and robustness.

 Note that the turnover times of SOM pools (active, slow and passive) used in this study are obtained from the ORCHIDEE-MICT (v8.4.1, Guimberteau *et al*., 2018). The turnover times of litter pools (metabolic and structural), as well as the 356 coefficient m_4 in Eq. (8) were optimized against the observed cumulative respired litter-C from all of the 14 incubation experiments using the M0 and M1 models (Table A3). A previous study has shown that litter decomposability is negatively correlated to its physical size (for example, Tuomi *et al*., 2011). Therefore, the turnover times of the fine litter fragments used in the incubation experiments of Recous *et al*. (1995) and Guenet *et al*. (2010) are expected to be shorter than the values set in ORCHIDEE-MICT, which are representative of the turnover times of natural plant residues. In addition, the mixing of soil and litter particle in the incubation experiment likely enhances decomposition as spatial disconnection of decomposer and substrate, which can occur under natural soil conditions (Barnes *et al*., 2012; Hewins *et al*.,

 2013), is prevented. The calibrated turnover times of the metabolic and structural pools and the value of *m⁴* in Eq. (8) are 3.5 and 30 days and 0.5, respectively.

 2.6 Impacts of litter stoichiometry and mineral N availability on SOM accumulation We used the model M1, with flexible CUE_d and decomposition rate function of available N to study the impacts of litter stoichiometry (C:N ratio) and soil mineral N availability on the formation and accumulation of SOM. In total, 24 idealized simulation experiments with different values of litter C:N ratios and soil mineral N availabilities were conducted (Table A4). The assumed litter C:N ratios (*CNlit*) of 10, 15, 30, 60, 120 and 200 span the variation among most natural substrates and soil amendments from organic matter input in agriculture (Harmon et al., 2009; Brovkin *et al*., 2012; Manzoni *et al*., 2010). The assumed range of mineral N availability (*Nmin*) 378 of 0.001, 0.005, 0.01 and 0.05 g N kg⁻¹ soil span the observed concentrations of soil mineral N in major terrestrial ecosystems (Metherall *et al.*, 1993).

 In each simulation experiment, model M1 was run for 5000 years to bring the litter and SOM pools in equilibrium with the prescribed litter input flux. The daily 382 input rate of plant litter was set to 0.006 $g C kg^{-1}$ soil day⁻¹, and the initial C stock of 383 litter and SOM pools were all set to be 0 g C kg^{-1} soil. During the simulation, soil 384 temperature and soil water content were assumed to be $25 \degree C$ and 60% of water holding capacity, respectively. We emphasized that our goal with this simplified scenario was to single out the effects of stoichiometric constraints, not to simulate the 387 effects of a realistic climatic regime. Parameter values for M1 (with $m_1 = 0.54$, $n_1 =$ 388 0.50 and $m_2 = 296.8$) used here were optimized based on all of the 14 incubation experiments from Recous *et al*. (1995) and Guenet *et al*. (2010) (see above). More detailed information about the specific settings of our simulation experiments can be found in Table A4.

3 Results

3.1 Evaluation of different models

 Results of leave-one-out cross-validation suggest that model M1 provides more accurate prediction of cumulative respired litter-C than other models (Fig. 3). The differences between simulated and observed cumulative respired litter-C from M1 are mostly less than 6% for over 93% of the data (Fig. A1b). The average RMSE of predicted cumulative respired litter-C from M1 (3.0%) is lower than that of model M0 (4.1%). Models M2 and M3 have slightly lower RMSE values than M0 (3.7% and 3.8%, respectively) but perform worse than M1 (Fig.4). However, the average AIC of all the models are comparable, suggesting that models with more fitted parameters do not over-fit the observations (Fig. 4).

 Model M1 captures the differences in respiration rates due to different C:N ratios of substrate and varying levels of mineral N availability across the 14 incubation experiments (Fig.5). While model M3 can reproduce the observed effect of soil mineral N availability on litter respirations rates (Fig.5d), it underestimates the 408 cumulative respired CO_2 from low quality litter ($CN_{lit} = 130$) at high mineral N 409 concentrations (> 0.04 g N kg⁻¹ soil). Models M0 and M2 cannot represent the effects of soil mineral N on litter respiration rate (Figs.5a, c), and their predictions are more biased from the observed values compared to M1. The Model M4, which uses the alternative formulation for N constraints on litter decay (Eq. (6)), reproduces the different respiration rates of substrates with contrasting C:N ratios and at different levels of mineral N availability (Fig. A2), but with a slightly higher average RMSE of cumulative respired litter-C than model M1.

416 The predicted CUE_d and the limitation effects of soil mineral N availability on litter decay rate (*f*(*Nmin*) function Eq. (5)) are different among the four tested 418 models (Fig. A3). In models M0 and M3, which used a fixed CUE_d , the fitted values 419 of CUE_d calculated with optimized parameters during the incubation period are about 420 0.57 and 0.54, respectively (Figs. A3a, d). In models M1 and M2, the CUE_d varies with the C:N ratios of plant litter, and is only slightly affected by soil mineral N concentrations (Figs. A3b, c). For very low quality litter with a C:N ratio of 130, the 423 CUE_d in models M1 and M2 are 0.40 and 0.44, respectively, which are lower than for better quality litter with C:N ratio of 44 (approximately 0.55 and 0.56 in M1 and M2,

respectively). Models M0 and M2 do not include the N inhibition effects on litter

426 decay rate, thus the $f(N_{min})$ in these two models is always 1 (Figs. A3e, g). In M1 and

427 M3, the N inhibition effect changes with both the litter C:N ratio and the mineral N

availability (Figs. A3f, h).

429 CUE_d from Eq. (2) calibrated with the data of the two incubation 430 experiments, decreases with increasing $CN_{li}(CN_{SOM})$ (Fig. 6). The average CUE_d value is larger than the average of data compiled for microbial CUE of litter decomposition in terrestrial ecosystems by Manzoni *et al*. (2017). This is shown by the gray circles in 433 Fig. 6. Our optimized values of CUE_d for a given C:N ratio are more comparable with microbial CUE observed in incubations of soil mixed with litter (Gilmour and Gilmour, 1985; Devêvre and Horwáth, 2000; Thiet *et al*.,2006), shown as black

squares in Fig. 6.

3.2 The effect of litter quality *vs* quantity on equilibrium SOM stocks

 Model M1 predicts that the size of the SOM pool at equilibrium is mainly determined by litter stoichiometry, with a minor effect of soil mineral N (Fig. 7). The lower C:N ratio of litter is, the higher equilibrium SOC stock. For litter with a specific 442 C:N ratio, high soil mineral N concentration (e.g. above 0.05 g N kg⁻¹ soil) generally produces a slightly larger equilibrium SOC stock than a low mineral N concentration (Fig. 7). Further analysis suggests that the SOC at equilibrium increases with 445 decreasing litter C:N because the SOC pool is positively related to the CUE_d ; however the limitation of soil mineral N on litter decomposition rate almost shows no impact on SOC (Fig. A4).

4 Discussion

450 We hypothesized that stoichiometric constraints (flexible CUE_d or inhibition of decomposition under N limited conditions) played a role in shaping the trajectory of litter decomposition, with potential consequences on predicted SOC stocks. Our 453 results suggest that flexible CUE_d and inhibition effects of soil mineral N on litter

 decay rate improve prediction of litter decomposition when using a modified version of CENTURY model (denoted as M1). Evaluation of the model M1 using data from incubation experiments indicate that this modified model captures the effects of both variable litter quality (stoichiometry) and mineral N availability on respiration rates (Fig. 5), without strongly inflating the complexity of CENTURY (Table 1). As the stoichiometric constraints are implemented in the generalizable and widely used structure of CENTURY and require only three parameters to be calibrated, they can also be easily implemented into land surface models for large spatial scale applications.

 Accurately representing N control of microbial processes during litter decomposition has been suggested to be important for modeling the connection between the litter inputs, CUEd, and soil C dynamics (Gerber *et al*., 2010; Manzoni *et al.*, 2012; Cotrufo *et al.*, 2013; Sinsabaugh *et al.*, 2013). In model M1, soil mineral N affects the litter-C flux via two mutually different pathways: (1) mineral N availability 468 affects the litter decay rate and (2) flexible CUE_d determining the partition of 469 decomposed C into SOC products and respired $CO₂$ (Fig. 1). Therefore, an increase in soil mineral N concentration enhances litter decay rates, which alone will increase the 471 flux of litter-derived CO_2 (Eq. (5) and Fig. A5). However, as higher N concentration 472 also results in a higher CUE_{d} (Eq. (2)), more C is transferred to SOC and less C is respired. In this way, SOC is predicted to accumulate with increasing mineral N availability when using model M1 (Fig. 7). In addition, the model M4, which is comparable to model M1 but uses an alternative formulation for N effects on the decomposition rate (Eq. (6)), performed slightly worse than model M1 (Fig. A2). Arguably, Eq. (6) represents the underlying mechanisms of N inhibition effects (Manzoni *et al*., 2009; Bonan *et al.*, 2013; Fujita *et al.*, 2014; Averill and Waring, 2018) better than Eq. (5) and due to the minor differences in RMSE and AIC (Figure A2b) between these formulations it can serve as an alternative to M1. Our results indicate that the observed diversity of responses of litter respiration rate to mineral N additions (Hobbie and Vitousek, 2000; Guenet *et al.*, 2010; Janssens *et al.*, 2010) is likely due to the combined effects of changes in litter

484 decay rate and CUE_d (Fig. A5). Thus, N addition effects can differ among fertilization experiments if litter quality and background N availability vary. In addition to altering 486 litter decay rate and CUE_d , mineral N addition can induce abiotic formation of compounds that resist microbial attack, inhibit oxidative enzymes involved in lignin degradation, stimulate microbial biomass production early in decomposition, or lead to the accumulation of microbial residues that are resistant to decay (Fog, 1988; Hobbie, 2015). All these effects might decrease litter respiration rate by inhibiting the decomposition process, but have not been considered in our current model.

 This study provides insights on processes leading to increased SOM sequestration. Enhancing the efficiency at which plant residuals are transformed into stable SOM has been suggested as an effective strategy to sequester C in soil (Prescott, 2010; Cotrufo *et al.*, 2013). Simulation results from our model suggest a positive linear relationship between equilibrium SOC stock and CUE of decomposed litter (Fig. A4), in line with the earlier findings with a similar model (for example Frey *et al*. 2013). In fact, with linear models such as CENTURY it can be shown that the steady state SOC scales linearly with CUE, different from nonlinear models predicting that higher CUE can trigger SOC loss (Allison *et al*., 2010). Our model goes beyond earlier attempts (Bonan *et al.*, 2013; Fujita *et al.*, 2014; Averill and Waring, 2018) by adapting the optimal metabolic regulation hypothesis of Manzoni *et al*. (2017) to link CUE, litter quality and SOM formation in a process-oriented way.

 The importance of litter quality for SOM formation as found here is in line with recent experiments (Bahri *et al*., 2008; Rubino *et al*., 2010; Walela *et al*., 2014) and modeling studies (Grandy and Neff, 2008; Cotrufo *et al.*, 2013). SOM is mainly formed though the partial decomposition of plant debris by microorganisms (Paul, 2007; Knicker, 2011; Cotrufo *et al.*, 2013). The conceptual model developed by Cotrufo *et al*. (2013) suggested that although labile litter was decomposed faster than recalcitrant litter, a higher fraction of this labile litter-C would be incorporated into microbial biomass and subsequently incorporated into SOM pool (corresponding to a 512 higher CUE_d). Therefore, labile litter inputs tend to form a larger SOM pool than the poor-quality (high C:N ratio) litter that is generally used by microbes at lower

 efficiency. Our simulations of decomposition process of plant litter with different C:N ratios also suggest that litter of good quality (with low C:N ratio) can induce a larger 516 SOM pool than the poor-quality litter (Fig. 7). CUE_d plays a more important role than the inhibition effect of low mineral N concentration in determining the size of the stable SOM pool (Fig. A4).

 The predictions from Cotrufo *et al*. (2013) and this study contrasts with the conventional hypothesis whereby the poor-quality litter with low decay rate and small 521 CUE_d are preferential to be accumulated in SOM (Berg and Mcclaugherty, 2008; Walela *et al.*, 2014).This view of SOM stabilization, however, seems to apply to N-limited systems with high C:N litter and where microbial remains are recalcitrant to decomposition (e.g., boreal forests) – in these systems SOC does accumulate despite 525 its low quality (Kyaschenko et al. 2017). Moreover, one could argue that higher CUE_d implies larger microbial biomass, allowing faster decomposition (Allison et al., 2010). These feedbacks between microbial biomass and decomposition rate were not implemented in the current model, but could offer additional flexibility – again at the expense of more difficult model parameterization.

530 The CUE_d formulation from Eq. (2) with parameters calibrated from the two sets of incubation experiments might underestimate the impacts of litter quality on microbial CUE under natural conditions, in particular in case of SOM decomposition. In both incubation experiments, litter is firstly cut into fine fragments and then fully mixed with mineral soil (Recous *et al*.,1995; Guenet *et al*., 2010). Thus, the nutrient accessibility, air permeability and some other environmental factors (e.g. pH) of incubated litter are different from those of decaying litter in more natural, heterogeneous soil conditions. Those different decomposition conditions might be responsible for the differences observed in Fig. 6 between our CUE estimates and previously reported values. We speculate that more heterogeneous conditions reduce nutrient availability and thus might cause lower CUE. Similarly, CUE of surface litter decomposers may be lower than we estimated (Fig. 6), because litter not mixed with soil is probably subject to stronger nutrient limitation.

Further validation and development of our model are still necessary to

 decrease the model uncertainties. Soil mineral N which affects both litter decay rate and CUE of decayed litter is seldom monitored in litter incubation experiments (e.g. Walela *et al*., 2014; Stewart *et al*., 2015) and field litter decay experiments (e.g. Gholz *et al*., 2000; Harmon *et al*., 2009), with few exceptions(Recous *et al*., 1995; Guenet *et al*., 2010). An increasing number of land surface models (Wang *et al*., 2010; Zaehle *et al*., 2014; Goll *et al*., 2017) have representations of the terrestrial N cycle. By incorporating our newly developed formulations of *CUE^d* and *f*(*Nmin*) in these land surface models that simulate the dynamics of soil mineral N concentration, it will be possible to test and validate our developments with more extensive data from laboratory and field experiments. Moreover, similar to N, P has also been suggested as another important factor for litter decomposition and SOM formation (Güsewell and Verhoeven, 2006; Talkner *et al*., 2009; Manzoni *et al*., 2010; Prescott, 2010), especially in regions with highly weathered soil (Goll *et al*., 2012, 2017; Yang *et al*., 2014). So it might be necessary to include the effects of P on litter decay rate and CUE_d into our model for further decrease the simulation uncertainties.

5 Conclusions

 By adapting the hypothesis of optimal microbial CUE proposed by Manzoni *et al*. (2017) for use in a CENTURY-based model and also introducing a N scaling function to represent the limits of mineral N availability on litter decay rate, we developed a simple but effective litter decomposition model that accounts for key stoichiometric constraints during decomposition. Validation using observation data obtained from laboratory incubation experiments indicated that our model could well predict the respiration rates of litter in different qualities at various levels of mineral N availability. Idealized simulations using our model revealed that the quality of litter inputs plays an important role in determining the soil C stock at equilibrium. High-quality litter (i.e. with low C:N ratio) tends to form a larger SOM pool as it can be more efficiently utilized by microorganisms than recalcitrant litter (e.g. high C:N ratio). Overall, the developed model captures the microbial mechanisms mediating

- litter stoichiometry and soil mineral N effects on litter decomposition and SOM
- formation. Due to the simple and generalizable structure of our model, it can be
- incorporated into existing land surface models for further long-term and large spatial
- scale applications.
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Code and data availability

- The CENTURY-based model used here is programmed in MATLAB
- language. The source code is available online
- [\(https://github.com/hchzhang/CENYUTY_CUE/tree/v1.0,](https://github.com/hchzhang/CENYUTY_CUE/tree/v1.0) DOI:
- 10.5281/zenodo.1307384). All the data used in this study can be obtained from
- published literatures. Specific references of these data can be found in section 2.5.

Competing interests

The authors declare that they have no conflict of interest.

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References

- Allison, S. D., Wallenstein, M. D., and Bradford, M. A.: Soil-carbon response to warming
- dependent on microbial physiology, Nature Geoscience, 3, 336, 10.1038/ngeo846, 2010.
- Allison, S. D.: A trait-based approach for modelling microbial litter decomposition, Ecology

letters, 15, 1058-1070, 10.1111/j.1461-0248.2012.01807.x, 2012.

- Averill, C., and Waring, B.: Nitrogen limitation of decomposition and decay: How can it occur?, Glob Chang Biol, 24, 1417-1427, 10.1111/gcb.13980, 2018.
- Bahri, H., Rasse, D.P., Rumpel, C., Dignac, M.F., Bardoux, G., Mariotti, A. (2008) Lignin
- 602 degradation during a laboratory incubation followed by ¹³C isotope analysis. Soil Biology and Biochemistry, 40, 1916-1922.
- Barnes, P. W., Throop, H. L., Hewins, D. B., Abbene, M. L. Archer, S. R.: Soil coverage reduces
- photodegradation and promotes the development of soil microbial films on dryland leaf litter, Ecosystems, 15, 311-321, 2012.
- Berg, B., and Mcclaugherty, C.: Plant Litter. Decomposition, Humus Formation, Carbon Sequestration, Springer Verlag, Heidelberg, 2008.
- Bonan, G. B., Hartman, M. D., Parton, W. J., and Wieder, W. R.: Evaluating litter decomposition
- in earth system models with long-term litterbag experiments: an example using the
- Community Land Model version 4 (CLM4), Glob Chang Biol, 19, 957-974.
- Bozdogan, H.: Model selection and Akaike's Information Criterion (AIC): The general theory and its analytical extensions, Psychometrika, 52, 345-370, 1987.
- Brovkin, V., Bodegom, P. M. V., Kleinen, T. and Wirth, C.: Plant-driven variation in
- decomposition rates improves projections of global litter stock distribution. Biogeosciences Discussions*,*8, 565-576, 2011.
- Campbell, E. E., and Paustian, K.: Current developments in soil organic matter modeling and the
- expansion of model applications: a review, Environmental Research Letters, 10, 123004,
- 10.1088/1748-9326/10/12/123004, 2015.
- Campbell, E. E., Parton, W. J., Soong, J. L., Paustian, K., Hobbs, N. T., and Cotrufo, M. F.: Using

litter chemistry controls on microbial processes to partition litter carbon fluxes with the Litter

Decomposition and Leaching (LIDEL) model, Soil Biology and Biochemistry, 100, 160-174,

- 10.1016/j.soilbio.2016.06.007, 2016.
- 624 Cleveland, C. C., and Liptzin, D.: C:N:P stoichiometry in soil: is there a "Redfield ratio" for the microbial biomass?, Biogeochemistry, 85, 235-252, 10.1007/s10533-007-9132-0, 2007.
- Cotrufo, M. F., Wallenstein, M. D., Boot, C. M., Denef, K., and Paul, E.: The Microbial
- Efficiency-Matrix Stabilization (MEMS) framework integrates plant litter decomposition
- with soil organic matter stabilization: do labile plant inputs form stable soil organic matter?,
- Glob Chang Biol, 19, 988-995, 10.1111/gcb.12113, 2013.
- Cotrufo, M. F., Soong, J. L., Horton, A. J., Campbell, E. E., Haddix, Michelle L., Wall, D. H., and Parton, W. J.: Formation of soil organic matter via biochemical and physical pathways of
- litter mass loss, Nature Geoscience, 8, 776-779, 10.1038/ngeo2520, 2015.
- Devêvre, O. C., and Horwáth, W. R.: Decomposition of rice straw and microbial carbon use
- efficiency under different soil temperatures and moistures, Soil Biology & Biochemistry, 32, 1773-1785, 2000.
- Duan, Q., Sorooshian, S., and Gupta, V. K.: Optimal use of the SCE-UA global optimization method for calibrating watershed models, Journal of Hydrology, 158, 265-284, 1994.
- Duan, Q. Y., Gupta, V. K., and Sorooshian, S.: Shuffled complex evolution approach for effective and efficient global minimization, Plenum Press, 501-521 pp., 1993.
- Finn, D., Page, K., Catton, K., Strounina, E., Kienzle, M., Robertson, F., Armstrong, R., and Dalal,
- R.: Effect of added nitrogen on plant litter decomposition depends on initial soil carbon and
- nitrogen stoichiometry, Soil Biology and Biochemistry, 91, 160-168,
- 10.1016/j.soilbio.2015.09.001, 2015.
- Fog, K.: The effect of added nitrogen on the rate of decomposition of organic matter, Biological Reviews, 63, 433-462, 1988.
- Franklin, O., Hall, E. K., Kaiser, C., Battin, T. J., and Richter, A.: Optimization of Biomass
- Composition Explains Microbial Growth-Stoichiometry Relationships, American Naturalist, 177, E29, 2011.
- Frey, S. D., Lee, J., Melillo, J. M., and Six, J.: The temperature response of soil microbial
- efficiency and its feedback to climate, Nature Climate Change, 3, 395-398,
- 10.1038/nclimate1796, 2013.
- Fujita, Y., Witte, J.-P. M., and van Bodegom, P. M.: Incorporating microbial ecology concepts into
- global soil mineralization models to improve predictions of carbon and nitrogen fluxes,
- Global Biogeochemical Cycles, 28, 223-238, 2014.
- García-Palacios, P., McKie, B. G., Handa, I. T., Frainer, A., Hättenschwiler, S., and Jones, H.: The
- importance of litter traits and decomposers for litter decomposition: a comparison of aquatic
- and terrestrial ecosystems within and across biomes, Functional Ecology, 30, 819-829,
- 10.1111/1365-2435.12589, 2016.
- Gerber, S., Hedin, L. O., Oppenheimer, M., Pacala, S. W., and Shevliakova, E.: Nitrogen cycling
- and feedbacks in a global dynamic land model, GB1001, Global Biogeochemical Cycles, 2010.
- Gholz, H. L., Wedin, D. A., Smitherman, S. M., Harmon, M. E., and Parton, W. J.: Long-term
- dynamics of pine and hardwood litter in contrasting environments: toward a global model of decomposition, Global Change Biology, 6, 751-765, 2000.
- Gilmour, C. M., and Gilmour, J. T.: Assimilation of carbon by the soil biomass, Plant & Soil, 86, 101-112, 1985.
- Goll, D. S., Brovkin, V., Parida, B. R., and Reick, C. H.: Nutrient limitation reduces land carbon uptake in simulations with a model of combined carbon, nitrogen and phosphorus cycling, Biogeosciences Discussions, 9, 3547-3569, 2012.
- Goll, D. S., Vuichard, N., Maignan, F., Jornet-Puig, A., Sardans, J., Violette, A., Peng, S., Sun, Y.,
- Kvakic, M., and Guimberteau, M.: A representation of the phosphorus cycle for ORCHIDEE
- (revision 4520), Geoscientific Model Development, 10, 3745-3770, 2017.
- Grandy, A. S., and Neff, J. C.: Molecular C dynamics downstream: the biochemical decomposition sequence and its impact on soil organic matter structure and function, Science of the Total
- Environment, 404, 297-307, 2008.
- Guenet, B., Neill, C., Bardoux, G., and Abbadie, L.: Is there a linear relationship between priming
- effect intensity and the amount of organic matter input?, Applied Soil Ecology, 46, 436-442, 10.1016/j.apsoil.2010.09.006, 2010.
- 679 Guimberteau, M., Zhu, D., Maignan, F., Huang, Y., Yue, C., Dantec-Nédélec, S., Ottlé, C.,
- Jornet-Puig, A., Bastos, A., Laurent, P., Goll, D., Bowring, S., Chang, J., Guenet, B., Tifafi,
- M., Peng, S., Krinner, G., Ducharne, A., Wang, F., Wang, T., Wang, X., Wang, Y., Yin, Z.,
- Lauerwald, R., Joetzjer, E., Qiu, C., Kim, H. & Ciais, P. (2018) ORCHIDEE-MICT (v8.4.1),
- a land surface model for the high-latitudes: model description and validation. Geoscientific Model Development, 11, 121-163.
- Güsewell, S., and Verhoeven, J. T. A.: Litter N:P ratios indicate whether N or P limits the decomposability of graminoid leaf litter, Plant & Soil, 287, 131-143, 2006.
- Hansen, S., Jensen, H. E., Nielsen, N. E., and Svendsen, H.: Simulation of nitrogen dynamics and
- biomass production in winter wheat using the Danish simulation model DAISY, Fertilizer Research, 27, 245-259, 1991.
- Harmon, M. E., Silver, W. L., Fasth, B., Chen, H. U. A., Burke, I. C., Parton, W. J., Hart, S. C.,
- and Currie, W. S.: Long-term patterns of mass loss during the decomposition of leaf and fine
- root litter: an intersite comparison, Global Change Biology, 15, 1320-1338,
- 10.1111/j.1365-2486.2008.01837.x, 2009.
- Hewins, D. B., Archer, S. R., Okin, G. S., McCulley R. L., Throop, H. L.: Soil-litter mixing
- accelerates decomposition in a Chihuahuan Desert grassland, Ecosystems, 16, 183-195, 2013.
- Hobbie, S.E.: Plant species effects on nutrient cycling: revisiting litter feedbacks. Trends Ecol Evol, 30, 357-363, 2015.
- Hobbie, S. E., and Vitousek, P. M.: Nutrient limitation of decomposition in Hawaiian forests,
- Ecology, 81, 1867-1877, 2000.
- Huang, Y., Guenet, B., Ciais, P., Janssens, I. A., Soong, J. L., Wang, Y., Goll, D., Blagodatskaya,
- E., and Huang, Y.: ORCHIMIC (v1.0), a microbe-driven model for soil organic matter
- decomposition designed for large-scale applications, Geoscientific Model Development
- Discussions, 1-48, 10.5194/gmd-2017-325, 2018.
- Ingwersen, J., Poll, C., Streck, T., and Kandeler, E.: Micro-scale modelling of carbon turnover
- driven by microbial succession at a biogeochemical interface, Soil Biology and Biochemistry, 40, 864-878, 10.1016/j.soilbio.2007.10.018, 2008.
- Janssens, I. A., Dieleman, W., Luyssaert, S., Subke, J. A., Reichstein, M., Ceulemans, R., Ciais, P.,
- Dolman, A. J., Grace, J., Matteucci, G., Papale, D., Piao, S. L., Schulze, E. D., Tang, J., and
- Law, B. E.: Reduction of forest soil respiration in response to nitrogen deposition, Nature
- Geoscience, 3, 315-322, 10.1038/ngeo844, 2010.
- Kätterer, T., and Andrén, O.: The ICBM family of analytically solved models of soil carbon,
- nitrogen and microbial biomass dynamics descriptions and application examples,
- Ecological Modelling, 136, 191-207, 2001.
- Kearns, M. & Ron, D. (1997) Algorithmic stability and sanity-check bounds for leave-one-out cross-validation. Neural Computation, 11, 1427-1453.
- Knicker, H.: Soil organic N An under-rated player for C sequestration in soils?, Soil Biology & Biochemistry, 43, 1118-1129, 2011.
- Krinner, G., Viovy, N., de Noblet-Ducoudré, N., Ogée, J., Polcher, J., Friedlingstein, P., Ciais, P.,
- Sitch, S., and Prentice, I. C.: A dynamic global vegetation model for studies of the coupled
- atmosphere-biosphere system, Global Biogeochemical Cycles, 19,

doi:10.1029/2003GB002199, 10.1029/2003gb002199, 2005.

- Kucharik, C. J., Foley, J. A., Delire, C., Fisher, V. A., Coe, M. T., Lenters, J. D., Young-Molling,
- C., Ramankutty, N., Norman, J. M., and Gower, S. T.: Testing the performance of a dynamic
- global ecosystem model: Water balance, carbon balance, and vegetation structure, Global
- Biogeochemical Cycles, 14, 795-825, 10.1029/1999gb001138, 2000.
- Kyaschenko, J., Clemmensen, K. E., Karltun, E., and Lindahl, B. D.: Below-ground organic
- matter accumulation along a boreal forest fertility gradient relates to guild interaction within fungal communities, Ecology letters, 20, 1546-1555, 10.1111/ele.12862, 2017.
- Lekkerkerk, L., Lundkvist, H., Ågren, G. I., Ekbohm, G., and Bosatta, E.: Decomposition of
- heterogeneous substrates; An experimental investigation of a hypothesis on substrate and microbial properties, Soil Biology & Biochemistry, 22, 161-167, 1990.
- Liski, J., Palosuo, T., Peltoniemi, M., and Sievänen, R.: Carbon and decomposition model Yasso
- for forest soils, Ecological Modelling, 189, 168-182, 10.1016/j.ecolmodel.2005.03.005, 2005.
- Luo, Y., Ahlström, A., Allison, S. D., Batjes, N. H., Brovkin, V., Carvalhais, N., Chappell, A.,
- Ciais, P., Davidson, E. A., and Finzi, A.: Toward more realistic projections of soil carbon
- dynamics by Earth system models, Global Biogeochemical Cycles, 30, n/a-n/a, 2016.
- Manzoni, S., Jackson, R. B., Trofymow, J. A., and Porporato, A.: The global stoichiometry of litter nitrogen mineralization, Science, 321, 684-686, 2008.
- Manzoni, S., and Porporato, A.: Soil carbon and nitrogen mineralization: Theory and models
- across scales, Soil Biology and Biochemistry, 41, 1355-1379, 10.1016/j.soilbio.2009.02.031,
- 2009.
- Manzoni, S., Trofymow, J. A., Jackson, R. B., and Porporato, A.: Stoichiometric controls on

carbon, nitrogen, and phosphorus dynamics in decomposing litter, Ecological Monographs,

80, 89-106, 2010.

- Manzoni, S., Taylor, P., Richter, A., Porporato, A., and Agren, G. I.: Environmental and
- stoichiometric controls on microbial carbon-use efficiency in soils, The New phytologist, 196,
- 79-91, 10.1111/j.1469-8137.2012.04225.x, 2012.
- Manzoni, S., Capek, P., Mooshammer, M., Lindahl, B. D., Richter, A., and Santruckova, H.:
- Optimal metabolic regulation along resource stoichiometry gradients, Ecology letters, 20, 1182-1191, 10.1111/ele.12815, 2017.
- Metherall, A. K., Harding, L. A., Cole, C. V., and Parton, W. J.: CENTURY Soil Organic Matter
- Model Environment Technical Documentation, Agroecosystem Version 4.0, Great Plains
- System Research Unit, Technical Report No. 4. USDA-ARS, Ft. Collins., 1993.
- Molina, J. A. E., Clapp, C. E., Shaffer, M. J., Chichester, F. W., and Larson, W. E.: NCSOIL, A
- Model of Nitrogen and Carbon Transformations in Soil: Description, Calibration, and Behavior1, Soil Science Society of America Journal, 47, 85-91, 1983.
- Moorhead, D. L., and Sinsabaugh, R. L.: A Theoretical Model of Litter Decay and Microbial Interaction, Ecological Monographs, 76, 151-174, 2006.
- Mooshammer, M., Wanek, W., Hammerle, I., Fuchslueger, L., Hofhansl, F., Knoltsch, A.,
- Schnecker, J., Takriti, M., Watzka, M., Wild, B., Keiblinger, K. M., Zechmeister-Boltenstern,
- S., and Richter, A.: Adjustment of microbial nitrogen use efficiency to carbon:nitrogen
- imbalances regulates soil nitrogen cycling, Nature communications, 5, 3694,
- 10.1038/ncomms4694, 2014.
- Pagel, H., Ingwersen, J., Poll, C., Kandeler, E., and Streck, T.: Micro-scale modeling of pesticide
- degradation coupled to carbon turnover in the detritusphere: model description and sensitivity
- analysis, Biogeochemistry, 117, 185-204, 10.1007/s10533-013-9851-3, 2013.
- Parton, W., Silver, W. L., Burke, I. C., Grassens, L., Harmon, M. E., Currie, W. S., King, J. Y.,
- Adair, E. C., Brandt, L. A., Hart, S. C., and Fasth, B.: Global-scale similarities in nitrogen
- release patterns during long-term decomposition, Science, 315, 361-364,
- 10.1126/science.1134853, 2007.
- Parton, W. J., Stewart, J. W. B., and Cole, C. V.: Dynamics of C, N, P and S in grassland soils: a
- model, Biogeochemistry, 5, 109-131, 1988.
- Paul, E. A.: Soil Microbiology, Ecology and Biogeochemistry, Academic Press, San Diego, CA, USA., 2007.
- Prescott, C. E.: Litter decomposition: what controls it and how can we alter it to sequester more
- carbon in forest soils?, Biogeochemistry, 101, 133-149, 10.1007/s10533-010-9439-0, 2010.
- Recous, S., Robin, D., Darwis, D., and Mary, B.: Soil inorganic N availability: Effect on maize
- residue decomposition, Soil Biology & Biochemistry, 27, 1529-1538, 1995.
- Rubino, M., Dungait, J. A. J., Evershed, R. P., Bertolini, T., Angelis, P. D., D'Onofrio, A.,
- Lagomarsino, A., Lubritto, C., Merola, A., and Terrasi, F.: Carbon input belowground is the major C flux contributing to leaf litter mass loss: Evidences from a 13 C labelled-leaf litter experiment, Soil Biology & Biochemistry, 42, 1009-1016, 2010.
- Schimel, J. P., and Weintraub, M. N.: The implications of exoenzyme activity on microbial carbon and nitrogen limitation in soil: a theoretical model, Soil Biology & Biochemistry, 35,
- 549-563, 2003.
- Schmidt, M. W., Torn, M. S., Abiven, S., Dittmar, T., Guggenberger, G., Janssens, I. A., Kleber, M., Kogel-Knabner, I., Lehmann, J., Manning, D. A., Nannipieri, P., Rasse, D. P., Weiner, S., and
- Trumbore, S. E.: Persistence of soil organic matter as an ecosystem property, Nature, 478,
- 789 49-56, 10.1038/nature10386, 2011.
- Sinsabaugh, R. L., Manzoni, S., Moorhead, D. L., and Richter, A.: Carbon use efficiency of microbial communities: stoichiometry, methodology and modelling, Ecology letters, 16,
- 930-939, 10.1111/ele.12113, 2013.
- Sitch, S., Smith, B., Prentice, I. C., Arneth, A., Bondeau, A., Cramer, W., Kaplan, J. O., Levis, S.,
- Lucht, W., Sykes, M. T., Thonicke, K., and Venevsky, S.: Evaluation of ecosystem dynamics,
- plant geography and terrestrial carbon cycling in the LPJ dynamic global vegetation model,
- Global Change Biology, 9, 161-185, 10.1046/j.1365-2486.2003.00569.x, 2003.
- Six, J., Frey, S. D., Thiet, R. K., and Batten, K. M.: Bacterial and Fungal Contributions to Carbon Sequestration in Agroecosystems, Soil Science Society of America Journal, 70, 555--569, 2006.
- Stewart, C. E., Moturi, P., Follett, R. F., and Halvorson, A. D.: Lignin biochemistry and soil N
- determine crop residue decomposition and soil priming, Biogeochemistry, 124, 335-351,
- 10.1007/s10533-015-0101-8, 2015.
- Talkner, U., Jansen, M., and Beese, F. O.: Soil phosphorus status and turnover in central-European
- beech forest ecosystems with differing tree species diversity, European Journal of Soil 805 Science, 60, 338-346, 2010.
- Thiet, R. K., Frey, S. D., and Six, J.: Do growth yield efficiencies differ between soil microbial
- communities differing in fungal:bacterial ratios? Reality check and methodological issues,
- Soil Biology and Biochemistry, 38, 837-844, 10.1016/j.soilbio.2005.07.010, 2006.
- Tramontana, G., Jung, M., Schwalm, C. R., Ichii, K., Campsvalls, G., Ráduly, B., Reichstein, M.,
- Altaf Arain, M., Cescatti, A., and Kiely, G.: Predicting carbon dioxide and energy fluxes
- across global FLUXNET sites with regression algorithms, Biogeosciences Discussions, 13,
- 1-33, 2016.
- Tuomi, M., Laiho, R., Repo, A., Liski, J.: Wood decomposition model for boreal forests,
- Ecological Modelling, 222, 709-718, 2011.
- Verberne, E. L. J., Hassink, J., Willigen, P. D., Groot, J. J. R., and Veen, J. A. V.: Modelling
- organic matter dynamics in different soils, Netherlands Journal of Agricultural Science Issued by the Royal Netherlands Society for Agricultural Science, 38, 221-238, 1990.
- Walela, C., Daniel, H., Wilson, B., Lockwood, P., Cowie, A., and Harden, S.: The initial
- lignin:nitrogen ratio of litter from above and below ground sources strongly and negatively
- influenced decay rates of slowly decomposing litter carbon pools, Soil Biology and
- Biochemistry, 77, 268-275, 10.1016/j.soilbio.2014.06.013, 2014.
- 822 Wang, Y.P., Law, R.M. and Pak, B.: A global model of carbon, nitrogen and phosphorus cycles for the terrestrial biosphere, Biogeosciences,7, 7, 9891-9944, 2010.
- 824 Wieder, W. R., Bonan, G. B., and Allison, S. D.: Global soil carbon projections are improved by
- modelling microbial processes, Nature Climate Change, 3, 909-912, 10.1038/nclimate1951, 2013.
- Wieder, W. R., Cleveland, C. C., Smith, W. K., and Todd-Brown, K.: Future productivity and
- carbon storage limited by terrestrial nutrient availability, Nature Geoscience, 8, 441-444, 2015.
- Yang, X., Post, W. M., Thornton, P. E., and Ricciuto, D. M.: The role of phosphorus dynamics in
- tropical forests a modeling study using CLM-CNP, Biogeosciences, 11, 14439-14473, 2014.
- Zaehle, S., Medlyn, B.E., De Kauwe, M.G., Walker, A.P., Dietze, M.C., Hickler, T., Luo, Y., Wang,

844 **Table 1** Optimized parameter values for the five versions of the litter decomposition 845 model used in this study. *cue*^{*fit}* is the optimized value of CUE, $m₁$ and $n₁$ are the</sup> 846 coefficients in Eq. (3), m_2 is the coefficient in Eq. (5), and m_3 is the coefficient in Eq. 847 (6). Values in brackets following each parameter are the means $(\pm$ standard deviations) 848 of the fitted parameter values based on "leave-one-out" cross-validation (see Section 849 2.5 for more details).

Version	CUE	$f(N_{min})$	Parameters
M ₀	Fixed	1	<i>cue</i> _{fit} (0.57 ± 0.004)
M1	Eqs. (2) , (3)	Eq. (5)	m_1 (0.61 ±0.34), n_1 (0.53 ±0.21), m_2 (297.4 ±38.0)
M ₂	Eqs. (2) , (3)	-1	m_l (0.11±0.01), n_l (1.96±0.13)
M ₃	Fixed	Eq. (5)	<i>cue</i> _{fit} (0.54±0.01), m_2 (396.9±23.6)
M4	Eqs. (2) , (3)	Eq. (6)	m_1 (0.13±0.07), n_1 (1.91±0.37), m_3 (0.58±0.12)

 Figure 2. Schematic plot of (a) the optimal carbon use efficiency (*CUEopt*) as a function of soil mineral nitrogen for different litter C:N ratios (from Eq. (2) in the 873 main text with $m_l = 0.3$, $n_l = 1.0$) and (b) the N limitation function $f(N_{min})$ applied to 874 litter decomposition rates (from Eq. (5) in the main text). CN_{lit} and CN_{SOM} are the C:N ratios of the litter and SOM pools, respectively. *CUEmax*= 0.8 is the maximum *CUE* 876 under optimal nutrient condition (C limitation only). m_l and n_l are the parameters of 877 Eq. (3) and m_2 is the parameter of Eq. (5).

 Figure 3 Comparison of the predicted cumulative respired litter-C to observed values at different times during litter decomposition process. Each dot denotes an observation of cumulative respired litter-C at a certain day. In total, there are 149 points. M0-M3 are the four versions of litter decay model tested in this study (Table 1).

 Figure 4 The RMSE and AIC of the simulated cumulative respired litter-C from the four versions of litter decay model used in this study. Error bars denote the standard deviation of RMSE or AIC for different incubation experiments. M0-M3 denote the four models tested in this study (Table 1).

 Figure 5 Time series of the simulated (lines) and observed (dots) cumulative respired litter-C (% of initial litter-C) at four different levels of soil mineral N availability (*Nmin*, 898 g N kg⁻¹ soil). *CN*_{lit} is the C:N ratio of plant litter. M0-M3 denote the four models tested in this study (Table 1). Here the simulation results of each model were calculated with parameters optimized based on all of the 14 samples of incubation experiments (Table A2).

 Figure 6 Comparison of *CUE^d* (lines) predicted by Eq. (2) with parameter values (*m²* 905 $= 0.54$, $n_1 = 0.50$) calibrated based on the incubation experiments (Table A2) of Recous et al. (1995) and Guenet et al. (2010)to observed *CUE* of terrestrial 907 microorganisms along a gradient of CN_S/CN_D . For observed CUE (dots), CN_D and *CN^S* are the C:N ratio of decomposers and their substrates, respectively. For simulated 909 CUE (lines), CN_S and CN_D correspond to the C:N ratio of donor (litter pool) and acceptor (the active SOM pool of the CENTURY), respectively. Gray dots are the estimated microbial CUE of litter decomposition in natural terrestrial ecosystems from Manzoni *et al*. (2017). Black squares are the microbial CUE measured via laboratory incubation experiments of Gilmour and Gilmour, (1985), Devêvre and Horwáth (2000) and Thiet *et al.*(2006). Error bars represent the standard deviations. 915 N_{min} (g N kg⁻¹ soil) is the concentration of soil mineral N.

 Figure7 (a) Accumulation of soil organic carbon (SOC) for constant substrates input (plant litter) with different C:N ratios (*CNlit*) at different levels of soil mineral N 920 concentrations (N_{min} , g N kg⁻¹ soil), (b) Change trends of equilibrium SOC stock and 921 carbon use efficiency of decomposed litter (*CUE_d*) with increasing litter C:N ratio.

923 **Appendix:**

924 **Table A1** List of symbols used in this study; stoichiometric ratios are all expressed on

925 a mass basis.

Table A2 Information about the 14 samples of laboratory incubation experiment used in this study. *CNlit* and *LClit* are the C to N ratio and lignin

Table A2 Information about the 14 samples of laboratory incubation experiment used in this study. CN_{lit} and LC_{lit} are the C to N ratio and lignin

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933 **Table A3** List of parameters calibrated for two versions of the litter decomposition

934 model (M0, M1): k_{lim} and k_{lim} are respectively the turnover rates of metabolic and

935								structural litter pools, m_4 is the coefficient in Eq. (8), <i>cuefit</i> is the optimized value of	
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936 CUE, m_l and n_l are the coefficients in Eq. (3), and m_2 is the coefficients in Eq. (5).

 Table A4 Specific setting of litter and SOM properties, and soil conditions in the 16 idealized simulations for exploring the impacts of litter stoichiometry (i.e. C:N ratio) and soil mineral N on SOC accumulation. *CNlit* and *LClit* are the C to N ratio and 941 lignin to C ratio of plant litter, respectively. Lit_{inp} (g C kg⁻¹ soil day⁻¹) is the daily input 942 rate of plant litter. CN_{SOM} is the C to N ratio of SOM pool. N_{min} (g N kg⁻¹ soil) is the 943 concentration of soil mineral N (NO_3^- -N + NH_4^+ -N). *Tem* (°C) and *SWC* (%) are the temperature and soil water content, respectively.

Experiment	CN_{lit}	LC_{lit}	Lit_{inp}	CN_{SOM}	N_{min}	Tem	SWC
$\mathbf{1}$	15	0.2	0.006	12	0.001	25	60
2	30	0.2	0.006	12	0.005	25	60
3	60	0.2	0.006	12	0.01	25	60
$\overline{4}$	120	0.2	0.006	12	0.05	25	60
5	15	0.2	0.006	12	0.001	25	60
6	30	0.2	0.006	12	0.005	25	60
7	60	0.2	0.006	12	0.01	25	60
8	120	0.2	0.006	12	0.05	25	60
9	15	0.2	0.006	12	0.001	25	60
10	30	0.2	0.006	12	0.005	25	60
11	60	0.2	0.006	12	0.01	25	60
12	120	0.2	0.006	12	0.05	25	60
13	15	0.2	0.006	12	0.001	25	60
14	30	0.2	0.006	12	0.005	25	60
15	60	0.2	0.006	12	0.01	25	60
16	120	0.2	0.006	12	0.05	25	60

 Figure A1 Distribution of the difference between the predicted cumulative respired litter-C (*Rssim*, %) and the observed values (*Rsobs*, %) for all experiments and points in time. SD is standard deviation of the biases. M0-M3 denote the four models tested in 951 this study (Table 1).

 Figure A2 Comparison between simulated cumulative respired litter-C with *f*(*Nmin*) (inhibition effect of soil mineral N on litter decomposition) calculated by Eq. (6) and the observed results from incubation experiments. In figure (b), M0-M4 denote the five versions of litter decay model in Table 1. M4 denote the model which used Eq. (2) 958 to calculate the dynamic CUE and Eq. (6) to calculate $f(N_{min})$.

 Figure A3 Dynamic of the simulated carbon use efficiency (*CUE*) and *f*(*Nmin*) during 962 the incubation experiments (Table A4). *CN*_{*lit*} is the C:N ratio of incubated litter, and 963 *N_{min}* is the initial soil mineral N concentration (g N kg⁻¹ soil). M0-M3 are the four models in Table 1. Here the simulation results of each model were calculated with parameters optimized based on all of the 14 samples of incubation experiments (Table A2).

 Figure A4 Relationship between C stock of the potentially equilibrated SOM pool and the carbon use efficiency of decomposed metabolic litter (*CUEd*) at the dynamic 971 equilibrium stage. $f(N_{min})$ denote the inhibition factor $(0-1)$ of soil mineral N on litter decomposition.

 Figure A5 Schematic plot for change trends of *f*(*Nmin*) (inhibition effect of mineral N, Eq. (6)), *CUE^d* (carbon use efficiency of decomposed litter, Eqs. (2), (3)) and *Rs_litt* (litter respiration rate) with increasing concentration of soil mineral N. *CUEmax* (= 0.8) is the maximum CUE set in this study. *opt_Nmin* denotes the concentration of soil 979 mineral N at which litter respiration is maximized. U_0 is the potential decomposition rate when mineral N is fully adequate for litter decay.