On the potential vegetation feedbacks that enhance phosphorus availability – insights from a process-based model linking geological and ecological time scales

C. Buendía¹,²,³, S. Arens¹, T. Hickler⁴,⁵,⁶, S. I. Higgins⁶,⁷, P. Porada¹, and A. Kleidon¹

¹Biospheric Theory and Modelling group, Max Planck Institute for Biogeochemistry, Hans-Knöll Str. 10, 07745 Jena, Germany
²Bayreuth Academy of Advanced African Studies, University of Bayreuth, Hugo-Rüdel-Str. 10, 95445 Bayreuth, Germany
³Ecological Modelling, University of Bayreuth, Dr.-Hans-Frisch-Str. 1–3, 95448 Bayreuth, Germany
⁴Biodiversity and Climate Research Centre (LOEWE BiK-F), post address Senckenberganlage 25, visiting address Georg-Voigt-Str. 14, 60325 Frankfurt am Main, Germany
Senckenberg Gesellschaft für Naturforschung, Senckenberganlage 25, 60325 Frankfurt am Main, Germany
Institute of Physical Geography, Goethe-University, Altenhöferallee 1, 60438 Frankfurt am Main, Germany
Department of Botany, University of Otago, 479 Great King Street, 9016 Dunedin, New Zealand

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Correspondence to: C. Buendía (coribuendia@gmail.com)

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Abstract

In old and heavily weathered soils, the availability of P might be so small that the primary production of plants is limited. However, plants have evolved several mechanisms to actively take up P from the soil or mine it to overcome this limitation. These mechanisms involve the active uptake of P mediated by mycorrhiza, biotic de-occlusion through root clusters, and the biotic enhancement of weathering through root exudation. The objective of this paper is to investigate how and where these processes contribute to alleviate P limitation on primary productivity. To do so, we propose a process-based model accounting for the major processes of the carbon, water, and P cycle including chemical weathering at the global scale. We use simulation experiments to assess the relative importance of the different uptake mechanisms to alleviate P limitation on biomass production. Implementing P limitation on biomass synthesis allows the assessment of the efficiencies of biomass production across different ecosystems.

We find that active P-uptake is an essential mechanism for sustaining P availability on long time scales, whereas biotic de-occlusion might serve as a buffer on time scales shorter than 10 000 yr. Although active P uptake is essential for reducing P losses by leaching, humid lowland soils reach P limitation after around 100 000 yr of soil evolution. Given the generalized modeling framework, our model results compare reasonably with observed or independently estimated patterns and ranges of P concentrations in soils and vegetation. Furthermore, our simulations suggest that P limitation might be an important driver of biomass production efficiency (the fraction of the gross primary productivity used for biomass growth), and that vegetation on older soils becomes P-limited leading to a smaller biomass production efficiency.

With this study, we provide a theoretical basis for investigating the responses of terrestrial ecosystems to P availability linking geological and ecological time scales under different environmental settings.
1 Introduction

Phosphorus (P) is an essential nutrient for life and its availability ultimately constrains the production of biomass by primary producers, especially on old and strongly weathered soils. On ecological time scales, P availability depends on the efficient mineralization and recycling of organic matter (Cuevas and Medina, 1988), while over geologic time scales it depends mainly on rock weathering which is the only P source (Gardner, 1990; Schlesinger, 1997; Buendía et al., 2010). Because weathering rates decrease with soil age, primary productivity in old ecosystems may be P limited (Walker and Syers, 1976; Wardle et al., 2004). Low P input from weathering, however, does not mean that there is no P in old and strongly weathered soils. It just means that most of the P is not available for plants, because it is in places that roots cannot access and/or occluded in secondary minerals originating from soil formation processes (Walker and Syers, 1976; Crews et al., 1995; Vitousek and Sanford, 1986). To overcome P limitation, plants have evolved strategies to make P available from these occluded pools and take it up more efficiently. These strategies in turn induce feedback processes between plants and soils, which ultimately lead to the enhancement of P availability (for a review on those strategies refer to Lambers et al., 2008). These feedback processes might lead to a sustained biomass production and resource acquisition (e.g. more leaves), which in turn allows for increased carbon fixation by plants and more belowground investments, thus promoting P enhancement processes in the soil (DeLucia et al., 1997; Allen et al., 2003). While at the global scale these feedback mechanisms between plants and P availability in the soil determine to some extent the production and storage of biomass in ecosystems, they are also relevant for atmospheric C concentrations, and thus for the climate system (DeLucia et al., 1997; Zhang et al., 2011; Goll et al., 2012). If the biomass production of global vegetation is constrained by P availability, less carbon will be fixed to structural biomass, and therefore it will be easily respired and released to the atmosphere (Körner, 2006; Canadell et al., 2007). Hence P dynamics in soils and vegetation might be crucial for understanding the terrestrial
vegetation feedback on the global C cycle (Sardans and Peñuelas, 2012; Cernusak et al., 2013).

At the individual plant scale, C is assimilated during photosynthesis and stored as carbohydrates in the plant (hereafter referred to as gross primary productivity GPP). A proportion of the assimilated C is respired to run the metabolic machinery of plants (i.e. autotrophic respiration); another proportion is allocated to biomass production (BP) if other essential nutrients are available; and the remaining proportion is stored as non-structural carbon (i.e. reserves such as sugars and starch) that is used to overcome unfavorable conditions in the life cycle of plants. The carbon allocated to growth and non-structural carbon is the net primary productivity (NPP).

Under P limitation, biomass production is constrained, which results in an accumulation of non-structural carbon in the plant. Some of this non-structural carbon can be transferred to the soil via root exudation (Jakobsen and Rosendahl, 1990; Körner, 2006; Bais et al., 2006) (refer to Fig. 1 blue box). In such cases, NPP can be substantially greater than biomass production. There are different pathways by which plants can enhance P availability through root exudation. For example, (A) root exudates feed mycorrhizae that help the plant to actively take up P in dissolved forms; (B) root exudates, like citrate and oxalate (directly by the plant or mediated by soil microorganisms) might release P occluded in secondary minerals, thus making it available for plant uptake (Banfield et al., 1999; Bais et al., 2006); (C) root exudation stimulates microbial activity and respiration in the soil. The increased microbial activity in turn increases the CO₂ concentration in the soil leading to an enhancement of rock weathering (Berner, 1992; Taylor et al., 2009) and thus increased P availability (Lenton, 2001), even though this process works more indirectly than the other (A and B) mechanisms.

Over 80 % of terrestrial plant species have evolved P acquisition strategies (Lambers et al., 2008) and it is estimated that 10 to 20 % of carbon gained by photosynthesis is used to subsidize mycorrhizal associations (Chapin et al., 2002; Allen et al., 2003). Indeed, recent observations have shown that “fertile” ecosystems can produce biomass more efficiently than infertile ones (Vicca et al., 2012). For example, nutrient rich forests
fix about 58% of their total C uptake into biomass, whereas “nutrient poor” forests fix only about 42% of their total C uptake into biomass (Vicca et al., 2012). Ecosystems that are not limited by nutrients (solid black line in Fig. 2) are likely to have greater relative biomass production, which would imply a strong negative feedback on atmospheric CO$_2$ changes (i.e. case 1 in the blue frame of Fig. 1). Ecosystems that are limited by P availability (or other nutrients), have a lower ability to produce biomass, thus resulting in a weaker vegetation feedback on atmospheric CO$_2$ concentrations (gray solid line in Fig. 2) unless they actively ameliorate their P-limitation (grey dashed line in Fig. 2) through different processes, which also imply a C cost (red box in Fig. 1). Hence, P availability potentially constrains the biomass production efficiency of ecosystems and thereby regulates the strength of the atmospheric carbon cycle feedback.

Understanding P availability of the world’s terrestrial ecosystems not only involves understanding P-uptake but also involves understanding the pedogenetic processes (i.e. soil formation and evolution). Pedogenesis is a function of time, climate, composition of the parent material (lithology), topography, and involves the activity of the biota (e.g. soil organisms) (Jenny, 1958). Because climate, lithology, and topography vary geographically across the globe, there is a great diversity of soil types and developmental stages. Additionally, catastrophic phenomena, such as glacial retreats, soil removal and exposure of rock surfaces can reinitiate soil formation. Vegetation further affects pedogenetic processes by mediating the fluxes of carbon and water into the soil. The interaction of the different factors involved in pedogenesis ensures that soils and their phosphorus dynamics vary. It follows that taking into account pedogenetic processes and the processes by which vegetation can enhance P availability is necessary to understand global biomass productivity and the response of terrestrial ecosystems to the anthropogenic enhancement of atmospheric CO$_2$ concentrations.

This understanding has motivated the inclusion of P limitation in dynamic vegetation models (Wang et al., 2007, 2010; Mercado et al., 2011; Goll et al., 2012; Yang et al., 2013b). Given the complexity of these models and the small time step at which they are run they rely on P soil data and assumptions of C : N : P ratios in vegetation and
soil organic matter. This approach might work for site scale (Runyan and D’Odorico, 2012) or in areas where P measurements exists (Mercado et al., 2011). However, at the global scale this approach is problematic given the difficulty of accurately measuring available P (DeLonge et al., 2013) and the high uncertainty of P maps resulting from extrapolating P site specific data to the global scale (Yang et al., 2013a). Therefore, we take a rather simple and different approach in which only the concentration of P in different lithologies is used to initialize the model. We let the model mechanistically calculate soil formation, weathering and erosion processes, at the same time as vegetation carbon, water and phosphorus dynamics. By this we link the geological with the ecological time scales and avoid using data for initialization.

The aim of this study is to estimate the effect of P availability on terrestrial biomass production and the extent to which vegetation can alleviate P limitation by P enhancement processes, using a process-based model that links processes working on ecological and geological time scales.

To do so, this paper is structured as follows: First, we provide an introduction to our model including a detailed description of the P cycle and how we implemented its different components into our C-P biogeochemical model. In doing so, we also define simulation scenarios used to determine the importance of P enhancement processes on long-term NPP and BP. Second, the criteria we used to select a model scenario from the three simulation experiments and how the simulations are evaluated against observations are explained. Third, the model performance with respect to the geographic variation of P in the soil and C in vegetation as well as their temporal dynamics is assessed. In the discussion section, the limitation and general applicability of our approach are discussed. We conclude with a summary of our findings and their potential implications.
2 Methods

To understand the spatial and temporal variations of P limitation on terrestrial productivity, we coupled a P model to a regolith model. The regolith model of Arens (2013) describes how vegetation, soil, water, and carbon feedbacks influence soil formation. The model is based on the concentrations of elements in different lithological classes, topography, and climate and calculates chemical weathering of primary minerals, erosion of secondary minerals, and the associated tectonic (isostatic) uplift. This model is coupled to a simple vegetation model that is forced by daily climate (i.e. precipitation, temperature, humidity) to calculate soil-vegetation water and carbon balances (Porada et al., 2010). Phosphorus weathering is described as a process depending on P concentration in the different lithological classes and calcium weathering, which is calculated in the regolith model. The abundance of secondary minerals (clays) calculated by the regolith model determines P occlusion and erosion. The (passive) P uptake by plants depends on their water uptake. The availability of inorganic phosphate in vegetation biomass constrains photosynthesis, as without it, the transformation of energy into ATP is not possible. Synthesis of vegetation biomass requires inorganic P to form phospholipids and genetic material; without P the synthesis of metabolic active tissue is not possible. The carbohydrates that are neither respired nor synthesized into biomass accumulate in the plant. A fraction of these carbohydrates can be utilized as root exudates. Root exudates are rapidly respired in the soil thus altering the soil CO$_2$ balance as represented in Fig. 1 path A. Biotic P uptake and P de-occlusion as represented in Fig. 1 paths B and C also depend on root exudation but have different effects in the soil as the first only uses the P that is in the soil in available forms, and the second uses the P that is bound to secondary minerals. These processes are also included in the model and can be switched on and off allowing simulation experiments that test the relative importance of the three processes on the biomass production by vegetation.
2.1 Model formulation

The biochemical model used here explores the short (daily) and long-term (thousand to hundred thousand years) interactions between the biogeochemical C and P cycles, given (fixed) lithological, topographical and climatological constraints. The model runs on a global scale in which the fluxes of water, carbon, phosphorus, and heat between the soil and vegetation are calculated on a daily time step. The model also accounts for the pools of calcium, silicates, and other primary mineral fluxes in the bedrock and their export to the soil. The diagram shown in Fig. 3 summarizes the model inputs, compartments, and outputs.

The basic version of the model, which is called JESSY-SIMBA, was conceived as a simple global model of terrestrial biogeochemistry accounting for the most important fluxes of carbon and water. This version is well documented and tested in Porada et al. (2010). An extension of JESSY-SIMBA, including soil formation, erosion and weathering, called the regolith model was documented and tested by Arens (2013). The regolith model provides the carbon cycling module of the C-P model. This model was used here because it captures well the global patterns of Ca$^{2+}$ and HCO$_3^-$ fluxes across climatic and tectonic gradients, which are derived from weathering. The P model described in Buendía et al. (2010) was the basis of the P dynamics module, and it captures the general dynamics of P in pedogenesis and soil moisture effects on P dynamics. To couple C and P dynamics, a number of changes in the soil and vegetation C dynamics were necessary: (1) the carbon pool in vegetation was split into structural carbon C$^{v}_o$, and storage carbon C$^{v}_c$; (2) the storage carbon from vegetation flows to the soil generating a new flux and pool of organic carbon in the soil C$^{s}_c$; (3) the inclusion of biotic de-occlusion of P resulted in a pool of carbon associated to secondary minerals (e.g., aluminium citrate) C$^{s}_y$.

A summary of the balance equations of water, carbon, calcium, and phosphorus can be found in Appendix A. For the description of symbols and parameters used explicitly in the P cycle we refer to Tables 4 and 5. We followed the JESSY-SIMBA naming
convention. State variables are conventionally named $A_{b}^{c}$, where $A$ is a chemical compound, e.g. $P$, or a state variable such as temperature $T$, $b$ is the state of matter, e.g. $s$ for solid (see Table 1). Fluxes have names of the form $fA_{bt}^{sa}$, where $b$ is start and $t$ the end state (to consider phase change) and $s$ is the start and $a$ the end location. $P$ root uptake from soil to vegetation is thus named $fP_{d}^{sv}$.

2.2 Bedrock related processes

2.2.1 Weathering

The main input of $P$ to terrestrial ecosystems is chemical weathering; therefore, an appropriate implementation of the weathering fluxes in the model is essential. To estimate weathering rates globally, tectonic uplift rates and $P$ concentrations in primary minerals are needed. Arens (2013) solves this problem by assuming uplift equals erosion. The topographic relief is constant through simulation time, which means that there must be uplift, or else topographic relief would decline. In principle, in steady state, minerals of the bedrock constantly replace eroded material, so the weight of the soil column remains constant (on a yearly basis). Given that only few minerals contain $P$ in large amounts, in igneous rocks it is present in tiny crystals in the form of fluorapatite, while in sedimentary rocks it is found associated to authigenic carbonate-fluorapatite (Filippielli, 2008). $P$ concentration is higher in igneous rocks than in sedimentary rocks. To account for this differences the model uses the lithological map of Amiotte Suchet and Probst (1995) with apatite concentration of primary mineral estimated from Newman (1995) (see Fig. 5).

Tectonic uplift of apatite from the bedrock to soil ($fP_{m}^{bs}$) is defined as uplift rate (uplift) multiplied by the concentration of apatite ($pP$), given by the lithology (see Fig. 5 for global patterns):

$$fP_{m}^{bs} = pP(\text{uplift}).$$

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The weathering module has been already tested and documented for calcium fluxes in Arens (2013). Since phosphorus in rock material is a trace mineral, we calculate P weathering as a sub-product of calcium weathering ($f_{Ca_{pd}}$):

$$f_{P_{md}} = f_{Ca_{pd}} \frac{P_m}{Ca_p}$$  \hspace{1cm} (2)

where $P_m$ and $Ca_p$ are the concentrations of P and Ca in primary minerals in the soil (see Appendix A5 for the balance equations). Note that calcium fluxes follow the notation proposed in Arens (2013) instead of the notation proposed in Porada et al. (2010) and followed here.

### 2.2.2 Secondary mineral formation and P occlusion

Primary rocks react with CO$_2$ and water, releasing its components in dissolved forms, some of these components leave the soils with the base flow (e.g., Si, Ca) or are taken up by vegetation (e.g., P and Ca), while other minerals (e.g., Al, Fe and Mn) stay in the soil, forming secondary minerals. However, when soil is not well drained, Ca accumulates super saturating the soil solution. Super saturation stops the weathering reaction, resulting in weathering being limited by the calcium export capacity of the soils, which is referred to as the ecohydrological limitation to weathering (Arens and Kleidon, 2011). Whereas on landscapes with low topographic relief, where erosion is low and therefore tectonic uplift as prescribed in the model is low, the limitation to weathering is the supply of new fresh primary minerals.

The increase of secondary minerals with soil formation increases their capacity to co-precipitate phosphate by binding P to Al, Fe, and Mn (Filippelli, 2002). This process is referred to as P occlusion ($f_{P_{dy}}$) in this model and is formulated as a function of secondary mineral density ($Clay^s$) and P availability in the soil ($P_{d}^s$):

$$f_{P_{dy}} = \rho \tau P_{yd} Clay^s P_{d}^s$$  \hspace{1cm} (3)
where $\rho \tau P^s_y$ is the occlusion rate of P, bound by secondary minerals (y), in dissolved state (d) in the soil (s).

### 2.3 Vegetation dynamics

In regions with sufficient P availability, passive uptake of P is sufficient to satisfy vegetation demands of P. In regions with low P availability, vegetation relies on active P uptake mechanisms. Therefore, we formulate P uptake by vegetation ($fP^sv_d$) as a combination of passive and active P uptake. Root water uptake, $fH_2O^sv_l$ (i.e. flow from soils (s) to vegetation (v) in liquid form (l)), drives passive P uptake:

$$fP^sv_d = afP^sv_d fH_2O^sv_l \frac{P^s_d}{H_2O^s_l}. \quad (4)$$

In Buendía et al. (2010), mycorrhizal hyphae were found to be very important for maintaining P in vegetation, particularly in old ecosystems. Using an enhancement factor for mycorrhizal mediated P uptake, $afP^sv_d$, as defined in Buendía et al. (2010) and calibrated at the global scale, led to unrealistic amounts of $P^v_d$, particularly in areas with sufficient P supply. While accounting for carbon dynamics, it makes sense to associate a carbon cost with this symbiotic P uptake mechanism between plant and mycorrhiza. Therefore, the enhancement factor $afP^sv_d$ is defined as a function of root exudation of carbohydrates $fC^sv_c$ of the following form:

$$afP^sv_d = 1.0 + cSU \tanh(\rho \tau P^sv_d fC^sv_c), \quad (5)$$

which reaches an asymptotic limit of $cSU$ at high exudations rates. By setting the value of $cSU$ to 0, the effect of active uptake can be removed. For the simulations presented here including active uptake a biotic active P uptake of $cSU = 10$ is used (Buendía et al., 2010).
2.3.1 Gross primary productivity (GPP)

In the original JESSY-SIMBA model, gross primary productivity (GPP, $f_{C_{gc}}^{av}$) was a simple parameterization of carbon and water fluxes, constrained by light, temperature, soil water, and relative humidity (for details, we refer to Porada et al., 2010). Here, motivated by the role that ATP plays in short term energy storage for photosynthesis gross primary productivity ($f_{C_{gc}}^{av}$) is further constrained by P-availability, $P_d^v$ (Mercado et al., 2011):

$$f_{C_{gc}}^{av} = \begin{cases} \min(f_{C_{gc}}^{av}[\text{light}], f_{C_{gc}}^{av}[\text{water}]), & P_d^v > 0, \\ 0, & P_d^v = 0. \end{cases}$$

The carbohydrates taken up during photosynthesis are partially respired. Those that remain are referred to as net primary productivity (NPP). Synthesis of structural biomass occurs only when other nutrients are also available; therefore, biomass productivity equals NPP only in cases when ecosystems are not limited by other nutrients.

2.3.2 Biomass production and net primary productivity

Net primary productivity for the purpose of this study is taken as 60% of GPP, as data suggest that a maximum of 60% of GPP can be allocated to biomass (Vicca et al., 2012). To calculate the synthesis of actively metabolizing biomass (BP), first the potential fluxes are calculated. The potential carbon flux is defined as net primary productivity (NPP) plus carbohydrate storage in vegetation ($C_v^v$), $af C_{co}^v = 0.6 f_{C_{gc}}^{av} + C_c^v$ while the potential P synthesis is defined as a fraction, $c_{max} f_{P_{do}}^v$, of inorganic phosphate in vegetation $P_{d}^v$, $af P_{do}^v = c_{max} f_{P_{do}}^v P_{d}^v$. Finally the minimum between the potential fluxes is calculated:

$$f_{C_{co}}^v = \min(af C_{co}^v, c C_P^v af P_{do}^v) \quad \text{and} \quad f_{P_{do}}^v = \frac{f_{C_{co}}^v}{c C_P^v} \quad \quad (6)$$
The minimum between the potential fluxes $fC^v_{co}$ is the actual biomass synthesis (or BP). In cases where C or P limits BP, the non-limiting nutrient is then stored. Carbon is stored in C-rich structural carbohydrates $C^v_c$ and the inorganic phosphate in vegetation ($P^v_d$). Given that those two pools have no size constraints this results in a flexible C to P ratio in vegetation, which indicates the degree of P limitation. Although, in real ecosystems, a high carbon to nutrient ratio also results from structural biomass that allows the vegetation to growth taller, this is not explicitly accounted for in the model. The description of vegetation coverage is changed to account for the new formulation of carbon in vegetation. That is because the leaf area index ($m^2$ leaf surface per soil surface) depends on the biomass that contains P and not on the carbohydrate storage. Total carbon in vegetation is then the sum of structural $C^v_o$ and storage $C^v_c$ carbon.

### 2.3.3 Autotrophic respiration

In the original version of the model, the difference between GPP and NPP was denoted by autotrophic respiration $fC^g_{cg}$: half was respired into the soil ($fC^v_{cs}$) and half into the atmosphere ($fC^v_{ca}$). Here, autotrophic respiration is divided into growth and maintenance. For each carbon unit fixed in structural biomass ($C^v_o$), 0.25 carbon units are respired. Therefore, growth respiration ($pBP_{cg}$) is set to 0.25 (i.e. $pBP_{cg} = 0.25$) (Lavigne and Ryan, 1997). Maintenance respiration is defined with respect to actively metabolizing biomass ($C^v_o$), as follows:

$$fC^v_{cg} = pBP_{cg}fC^v_{co} + Q10C^v_o,$$  \hspace{1cm} (7)

Maintenance respiration depends on temperature and on the size of the structural carbon pool $C^v_o$, therefore, the Q10 relation was used:

$$Q10 = \frac{cq^{10^{\frac{T_s - cT_{Melt}-10.0}{10.0}}}}{\rho \tau C^s_o}$$  \hspace{1cm} (8)
The Q10 relation depends merely on soil temperature $T_s$. The decay rate is given by a global constant parameter $\rho \tau C_0^S$ that is obtained by matching observations (IGBP-DIS, 1998).

### 2.3.4 Carbon and phosphorus losses from vegetation

The input of C and P into the soil via litterfall ($f_{C_0}^{vs}$ and $f_{P_0}^{vs}$, respectively), root C exudation ($f_{C_c}^{vs}$) and P leaching ($f_{P_d}^{vs}$) are proportional to their respective plant pools $C_0^v$, $P_0^v$, $P_d^v$, and $C_c^v$. For litter we used the formulation of Porada et al. (2010)

$$f_{C_0}^{vs} = \frac{C_0^v}{\rho \tau C_0^v}$$

$$f_{P_0}^{vs} = f_{C_0}^{vs} \frac{P_0^v}{C_0^v}$$

(9)

$\rho \tau C_0^v$ represents the residence time of biomass in vegetation. Despite the changes in carbon dynamics of the vegetation, this parameter was not re-calibrated.

Root exudates residence time $\rho \tau C_c^v$ is not known. However, in steady state root exudation equals carbohydrate and starch formation. Therefore, the parameter $\rho \tau C_c^v$ was setted by personal assessment to match total carbon in vegetation:

$$f_{C_c}^{vs} = \frac{C_c^v}{\rho \tau C_c^v}$$

(10)

The losses of inorganic phosphate in vegetation ($P_d^v$) are decoupled from the losses of P in biomass ($P_0^v$) as it has been observed that plants reabsorb P before dropping the leaves. Probably they can reabsorb inorganic P used for photosynthesis, but not the organic P that is found in DNA or phospholipids. A minimal fractional loss was defined to avoid super-saturation of P of plants in regions with high P availability:

$$f_{P_d}^{vs} = \frac{P_d^v}{\rho \tau P_d^v}$$

(11)
2.4 Soil processes

2.4.1 Decomposition

Organic matter decomposition \((fC_{og}^s\) and \(fP_{od}^s\)) is a process mediated by soil fungi and microbes, whose rate of activity depends on soil moisture and temperature, substrate supply and nutrient availability (Davidson and Janssens, 2006). In JESSY-SIMBA decomposition depends merely on temperature since that better predicts the soil organic biomass patterns (Gholz et al., 2000).

\[
fC_{og}^s = C_{og}^s Q_{10} \quad \text{and} \quad fP_{od}^s = fC_{og}^s \frac{P_{od}^s}{C_{og}^s}
\]  

(12)

The model assumes that carbon root exudates are respired almost entirely by mycorrhizae and soil microorganisms (Kunc and Macura, 1966), which is expressed by the flux \(pMyC_{cg}fC_{vs}^s\), where \(pMyC_{cg}\) is a parameter that accounts for respiration. Only small amounts of carbon root exudates reach the mineral soil, those follow also the Q10 relation (Eq. 8). Heterotrophic respiration of root exudates is thus described as:

\[
fC_{cg}^s = pMyC_{cg}fC_{vs}^s + C_{c}^s Q_{10}.
\]  

(13)

2.4.2 Biotic de-occlusion of P

In very old weathered soils, like those of western Australia, some of the plants have cluster roots instead of mycorrhiza associations (Lambers et al., 2008). Those roots produce carboxylates (e.g., oxalate and citrate), which can replace the P that is bound to aluminum and iron, and thus release this P (Bais et al., 2006; Lambers et al., 2008). This relation requires one sugar (six moles C) per one exchanged P \((cCP_y^s)\). Although we do not account for the process in molecular detail, organic acids are respired at a very high rate, with less than 5% reaching the mineral soil (Kunc and Macura, 1966). The remaining carbon \(C_{c}^s\) might be used for P to C exchange. Therefore, P de-occlusion
is calculated as the minimum between occluded P $P_y$ and plant root exudates in the soil $C_c$

$$fP_{yd}^{sv} = \min \frac{P_y}{pdt}, cCP_y^{s}C_c^{s}$$
and $fC_{cy}^{ss} = fP_{yd}^{sv}cCP_y^{s}$. \hspace{1cm} (14)

With the introduction of this process a new pool of carbon in the soil is created, which accounts for the carbon that was exchanged with P in secondary minerals (e.g., aluminum citrate). We refer to this pool as the recalcitrant carbon in the soil, $C_y^v$.

### 2.4.3 P losses

Phosphorus is lost from the system by baseflow $fH_2O_l^{sc}$, erosion (driven by topography and atmosphere). Our model accounts for: baseflow losses of dissolved P and erosion caused losses of occluded P. The reasons for excluding losses of litter and organic matter due to erosion and runoff are given in the discussion. For consistency the model does not consider atmospheric losses of P, because atmospheric inputs of P via e.g. deposition of dust, are also not accounted for. Later, this simplification will be justified in the discussion.

#### Baseflow losses

Losses of $P_d^v$ are driven by baseflow (described in Eq. 9, Porada et al., 2010), which is proportional to the total concentration of P in the soil solution: $\left(\frac{P_d^{sc}}{H_2O_l^{s}}\right)$,

$$fP_{d}^{sc} = fH_2O_l^{sc}\frac{P_d^{sc}}{H_2O_l^{s}}. \hspace{1cm} (15)$$
Erosion of secondary minerals and P occluded

The empirical formulation of physical erosion accounting for the denudation of secondary minerals driven by topographical gradients was originally proposed in (Hilley and Porder, 2008) (Eq. 22, Arens, 2013). Losses of occluded P are coupled to the losses of secondary minerals ($f_{\text{Clay}^{\text{sr}}}$) and therefore defined:

$$f_{\text{P}_{\text{y}}} = f_{\text{Clay}^{\text{sr}}} \frac{P_{\text{s}}}{\text{Clay}^{\text{s}}}.$$  \hspace{1cm} (16)

Although it is not clear what happens to the recalcitrant carbon formed in the soil ($C_{\text{y}}^{\text{s}}$), it is assumed that baseflow drives the losses of this pool and of the organic acids pool $C_{\text{c}}^{\text{v}}$,

$$f_{C_{\text{y}}} = f_{\text{H}_{2}\text{O}_{\text{l}}} \frac{C_{\text{y}}^{\text{s}}}{\text{H}_{2}\text{O}_{\text{l}}^{\text{s}}}$$ and

$$f_{C_{\text{c}}} = f_{\text{H}_{2}\text{O}_{\text{l}}} \frac{C_{\text{c}}^{\text{s}}}{\text{H}_{2}\text{O}_{\text{l}}^{\text{s}}}. \hspace{1cm} (17)$$

2.5 Simulation set-up

2.5.1 Climate forcing

The model was run using daily climate data from the NASA Land Surface and Hydrology archive (Sheffield et al., 2006) for the time period 1960–1989 rescaled a global rectangular grid with a resolution of 2.810 × 2.810 lon-lat degree. Daily means of temperature, precipitation, relative humidity, and longwave and shortwave radiation were used as input to model. The glacial cover of the last glacial maximum was used to define areas where regoliths are necessarily much younger than in the rest of the world. Areas that were covered by glacial lakes were treated in the same way as areas that were covered by ice (Arens, 2013).
2.5.2 Simulation scenarios

Model simulations were run in low resolution (2.810 × 2.810 lon-lat degree) for 150,000 yr. Simulation results are presented at high resolution (1.125 × 1.125 lon-lat degree) to derive a more spatially refined resolution to compare to observations were run only for 70,000 yr.

Figure 6 gives an overview of the processes implemented in the model and the possible feedbacks on P availability to vegetation. We will focus on quantifying the role of (a) biotically active P uptake (BAU) and (b) biotic de-occlusion of P (BD). A control scenario to test the enhancement of weathering through root exudation (W-P) could not be defined because P limitation with the proposed model structure cannot be decoupled from root exudation. Therefore, the W-P feedback cannot be quantified, but it is nonetheless included.

1. \( W-P \): describes the simplest dynamics of the P model. When P is limiting the new formation of structural biomass, carbohydrates accumulate and are released as root exudates. The exudates are quickly respired in the soil. This might enhance weathering in regions where \( \text{CO}_2 \) is the limitation to weathering as described in Arens (2013). This simulation is run only for 50,000 years, as the vegetation becomes very P limited.

2. \( BD \): biotic de-occlusion of P from the occluded pool. In this simulation root exudates interacted with secondary minerals to exchange C with P as described in Eq. (14). The possible feedbacks of root exudates on weathering cannot be excluded. This simulation was run for 50,000 yr.

3. \( BAU \): biotic active uptake of P from the available pool as described in Eq. (5). In this version we assessed the role of mycorrhizae in sustaining productivity at long time scales. W-P was also included. This simulation was run for 150,000 yr at low spatial resolution and for 70,000 yr at high spatial resolution.
4. **BD-BAU**: includes biotic de-occlusion and biotically active uptake of P. This simulation was run for 50,000 yr.

5. **Control**: in this simulation vegetation was provided with P depending on its needs in order to get the maximum biomass production rate. The model simulation reached the maximum BP to GPP rate given by the model \( p_f C_{co}^v = 0.6 \). In this simulation root exudation was set to 0 and the vegetation was unlimited by P. This simulation was run for one thousand years, as it was only used for controlling standing biomass and net and biomass productivity.

### 2.5.3 Assessment of vegetation feedbacks on P availability

Since the carbon dynamics of the model were changed with the coupling to the P cycle, comparison of our results with the results presented in (Porada et al., 2010) was not suitable. A scenario in which P dynamics were included, but the vegetation was unlimited by P, was defined as the biomass control scenario to which all other scenarios were compared. Standing biomass was calculated as 10 yr means at every 50,000 yr intervals. The scenario in which the C feedbacks on the P cycle could most effectively reduce P limitation (i.e. smallest difference in standing biomass to the control scenario) was selected for further simulations, using higher spatial resolution.

### 2.5.4 Model evaluation

After comparing the different scenarios, we choose the less P limited and closer to observations scenario to run the model for 70,000 yr at a high resolution \( (1.125 \times 1.125 \text{ lon-lat}) \). We compared our estimates with existing data and literature ranges. To evaluate our model’s performance, we compared the geographical patterns of soil P pools to the data-driven model estimated P soil pools by Yang et al. (2013a), and to results of the model study of Goll et al. (2012).

For the evaluation of temporal dynamics, the model was run for 150,000 yr at low resolution \( (2.810 \times 2.810 \text{ lon-lat}) \), to reduce the computational cost. Five year means
were calculated at 10 000; 50 000; and 150 000 yr for (a) the weathering fluxes, (b) C and P in vegetation biomass, (c) gross primary productivity, (d) biomass productivity, and (e) P availability in soil solution. We compared the model results with literature estimates and general insights gained from chronosequences.

3 Results

3.1 Evaluation of vegetation feedbacks on P availability

The control-P scenario was defined so that P was not limiting photosynthesis and biomass production so that biomass pools could reach their maximum value everywhere around the globe. Figure 7 shows the geographical biomass patterns without P limitation. The biomass reaches its maximum at about 24 kg C m\(^{-2}\) which is in accordance with satellite-derived estimates of total biomass (Saatchi et al., 2011).

To test the importance of the processes mediated by vegetation to enhance P availability we compare the control-P scenario with the other scenarios including gradually more processes. Figure 8 shows the differences in total vegetation biomass between the control-P scenario and the other 4 scenarios, BD, BAU, BD-BAU and W-P (defined in Sect. 2.5) in ecosystems after 50 000 yr of soil evolution. In general, we find less differences between the control-P and the scenarios accounting for BAU (shown in the two right panels). The simplest scenario considering only the vegetation feedback on weathering (W-P) indicates that, after 50 000 yr, vegetation is extremely P limited worldwide (top left panel). The scenario considering both, W-P, and biotic de-occlusion, BD, shows a very similar limitation on standing biomass (bottom left panel). BAU shows less overall limitation to P compared to the scenarios not including it. The maximum absolute differences, between W-P and BAU are about 10 kg C m\(^{-2}\).

Additionally, to explore on what time scale these processes respond, we looked at latitudinal averages of GPP and BP at 10 000 and 50 000 yr of soil evolution (Fig. 9). After 10 000 yr of soil evolution, the GPP of all model scenarios coincide. After 50 000 yr
W-P and BD scenarios slightly reduce their GPP near to the equator. All our model estimates are higher than GPP estimated by Beer et al. (2010), which was based on a global set of eddy covariance data (in the upper panels Fig. 8 referred as data). Biomass production is in all scenarios lower then the control-P scenario, indicating widespread P limitation. The highest difference of biomass productions is between the control and the W-P scenario. Only little differences in biomass production are observed in the tropics between the BD and W-P scenarios, suggesting that BD as it is included in the model does not really help vegetation to relieve P limitation. The scenarios that have the smallest differences to the control scenario are those considering BAU. After 10,000 yr BAU and BAU-BD overlap, and after 50,000 yr BAU and BAU-BD decrease with rest to 10,000 yr although BAU-BD is slightly less. The model inter-comparison for net primary productivity from Cramer et al. (1999), which usually is interpreted as BP, is included as a reference. The scenarios including BAU best fit this line.

Overall, mycorrhizae mediated P uptake is identified as a very important mechanism in terms of maintaining ecosystem biomass during longer time scales (over 50,000 yr). That is represented by the differences between right and left panels in Fig. 8. This process is important, not only in the tropics, but also in temperate regions. Therefore, BAU was included in further simulations. In contrast, including the P cycle without this carbon feedback, leads to a scenario with very high limitation to biomass production. Investing carbon in root exudates to release P occluded in secondary minerals is not very effective according to the model.

Our previous results regarding vegetation C dynamics highlighted the importance of mycorrhizal mediated P uptake and the inefficiency of biotic de-occlusion to supply P to old ecosystems. Regarding standing biomass patterns, the simulation including both, BD and BAU imposed the smallest constraint on standing biomass. However, the difference to the scenario including just BAU, was very small. The scenarios that included the BD feedback, gave unrealistic P occluded ranges and global patterns. Therefore, the BAU scenario was run for 70,000 yr at a higher resolution to compare spatial and temporal patterns of model results with observations.
3.2 Evaluation of modeled soil and vegetation P spatial patterns and temporal dynamics

Figure 10 shows the P status of soils and vegetation after 70,000 yr of simulation. Because, glaciation and volcanism restart the soil forming process and are not included in our model, our map cannot accurately reproduce present conditions across the globe. However, it presents general features of how climate, topography, and lithology might influence soil P status and vegetation productivity. Places that were glaciated during the last glacial maximum (about 21,000 yr ago) are omitted here. Because P in primary minerals in the bedrock was also considered, the total amount of apatite is not presented here. The occluded P pool is the largest P pool in soils. Our numbers are in the order of magnitude estimated by Yang et al. (2013a) and the patterns looks similar (see for comparison Appendix B Fig. A2). The second largest pool in the soil is P in litter and soil organic biomass (top-right panel). Our model estimates are five times smaller than the ones given by Yang et al. (2013a) (see for comparison Appendix B in Fig. A1). However, our upper range is higher than that reported for agricultural soils 25 g P m⁻² (Brady and Weil, 2008), and in general agreement with ranges and patterns from Goll et al. (2012). The modeled dynamics with respect to P in vegetation are within the observed ranges (Brady and Weil, 2008) and also in agreement with modeled estimates by Goll et al. (2012) (Fig. 2). Figure 11 shows the performance of vegetation including P dynamics. The total biomass patterns are in agreement with Saatchi et al. (2011) estimates. Although there are no big differences in the patterns of GPP and BP, the map of biomass productivity efficiency, here, is the result of P limitation.

Overall, we found reasonable agreement between our modeled P global patterns and independent data- and model-based estimates, in particular considering the substantial uncertainty associated with the data sets used here for comparison with our model results. Figure 12 shows how weathering, carbon to phosphorus ratio in vegetation and biomass productivity efficiency change in time. The idea of putting those plots together is also to see how patterns are correlated. The P weathering rates, calculated from our
model, are in the range of observed values (50 to 1000 g Pha$^{-1}$ a$^{-1}$) (Newman, 1995), after 10 000, 50 000 and 150 000 yr of soil formation (see Fig. 12 central panels). Generally, the overall weathering rates decreased in time, except for mountainous regions where weathering stayed relatively constant over time. The C to P ratio was correlated to the lithological classes (see Fig. 5 for comparison) and increased over time, as the soil aged. For example, at 10 000 yr the western part of the Amazon basin had lower weathering rates compared to those from the eastern part, but over time (i.e. after 150 000 yr) the weathering rate in the west decreased to a lesser extent than in the east, which resulted in eastern Amazon being more P limited (higher C : P ratio) compared to the west. Given that the soils of the Amazon basin are million years old the simulations run for 150 000 yr are the best data comparison of this area. Differences in the allocation of carbon and vegetation functions between Western and Eastern Amazon Basin have been observed (Quesada et al., 2012; Saatchi et al., 2011). Overall, we see how the ecosystems swift towards P limitation over time. After 10 000 yr of soil formation, most of the ecosystems have enough P for biomass synthesis and therefore have high BP to GPP ratio (refer to left panels). However, over time (after 100 000 yr), more ecosystems shift towards P limitation, reducing the overall biomass productivity efficiency.

### 3.3 Phosphorus constraints on productivity

In the introduction we explained how P availability in the soil constrains biomass production and that vegetation feedbacks on P availability can influence these dynamics (see Fig. 2). In the model we included the processes that make P available and also three vegetation feedbacks that could results in enhancing P availability for vegetation. Figure 13 shows how biomass productivity efficiency changes over time. We have plotted here GPP vs BP at three different time scales of 10 000, 50 000 and 150 000 yr to show how ecosystems get more constrained with time. Data for forest ecosystems from the study of Vicca et al. (2012) show a similar pattern, although biomass production
efficiency of the forest ecosystems is in general higher than our model efficiency. Nevertheless, our model illustrates how BPE decreases in time and how GPP is decoupled from BP (refer to Fig. 12 left panel).

Our results on BPE suggest that P availability limits BP in old lowland soils and therefore it modulates the carbon cycle, although other limiting nutrients surely also influence BPE other areas. Figures 12 and 13 show how BPE changes during soil evolution. As soils evolve, most ecosystems tend to be P limited and decrease their biomass synthesis, which agrees with observations (Wardle et al., 2004) and with modeling results (Menge et al., 2012).

4 Discussion

The model we present here is very simple and it cannot be applied to understand processes at the plant level. Instead, the model can be seen as proof-of-concept of how one can account for the interactions between carbon and P limitations at the soil-vegetation level, on ecological as well geological time scales. This work, therefore, represents a first step towards understanding how the P cycle constrains vegetation and how vegetation feedbacks affects P dynamics.

With this simple model we show that biomass productivity, most probably, depends to some extent on the feedback processes mediated by vegetation and, more importantly, on the long-term P cycle. Hence, it might be essential to consider long term processes of the P cycle and element synergies to constrain the magnitude of the terrestrial vegetation’s feedback on CO$_2$ concentrations in a process-based modeling framework. Our model shows reasonable patterns of P stocks in soils and vegetation.

Evaluating the P model at the global scale is challenging given the scarcity and uncertainty of observed data. Qualitatively, the patterns of the occluded P pool look similar to the statistically modeled data from Yang et al. (2013a). However, with respect to the more important pools like organic P in the soil, our model predicts much lower concentrations for the whole soil profile than the concentrations derived by Yang et al. (2013a)
for the top 50 cm of soil. The upper range of soil organic P in Yang (up to 587 g P m\(^{-2}\)) is five times larger than the upper ranges obtained by our model. We wonder about the processes that could generate the high values and high spatial variability of organic P forms in the Sahara region derived by Yang et al. (2013a) (see Fig. A1 Appendix B).

Given the current lack of vegetation cover, perhaps the organic P has been there for a long time, originating from lichen productivity (Porada et al., 2013), or is the remaining P from periods when the Sahel region was wetter and therefore vegetated (Scheffer et al., 2001). Perhaps running the model with climates of the past will lead to a more accurate representation of the P patterns in such regions. Nevertheless, compared to the study of Goll et al. (2012) and literature ranges (Brady and Weil, 2008), our estimates are at the right order of magnitude. Therefore, we think that given the simplicity of the model is a good representation of P distribution across Earth’s ecosystems.

The temporal dynamics of our model are consistent with data and understanding gained from chronosequences (Walker and Syers, 1976; Crews et al., 1995; Wardle et al., 2004). In regions with a low topographical gradient, P depletes during pedogenesis regardless of the inclusion of tectonic uplift in the model. This is accompanied by vegetation dynamics reaching a maximum productivity at intermediate P depletion stages and then declining, as Wardle et al. (2004) observed in chronosequences. However, our results suggest that, on long-time scales, ecosystems still have weathering inputs. This contrasts with the general assumption that old P depleted ecosystems merely depend on exogenous P sources (Jordan, 1982; Walker and Syers, 1976; Crews et al., 1995; Wardle et al., 2004).

This insight – steady-state weathering is limited by the tectonic uplift – provided by our model simulations is in agreement with the result obtained by our simple P model (Buendía et al., 2010) that does not account for the effects of soil CO\(_2\) and soil depth. Because tectonic uplift is defined by the potential physical transport of secondary minerals and hence by topographical gradients, weathering is only limited by eco-hydrological factors in mountain areas. That is, the maximum possible chemical
weathering rate is defined by the eco-hydrological conditions: regolith drainage, CO$_2$, and temperature (Arens, 2013).

This results in ecosystems with high tectonic uplift rates (mountain regions) having high P input lists and therefore a high BPE. However, those ecosystems, with a high topographical gradient, also have more losses of soil litter and organic biomass, which are driven by the runoff and the topographical gradient, are not accounted for here. Since P is transferred mainly between soil and vegetation, the losses of litter and organic biomass could result in P availability being lower in mountainous ecosystems than what our model represents without including these processes.

4.1 Model limitations

The model accounts for the chemical weathering of material and erosion of secondary minerals, but it does not account for the erosion of primary minerals. Mountain ranges influenced by tectonic processes, like the Andes or the Himalayas, erode large amounts of material that is not chemically weathered. The material is transported by runoff to rivers and deposited in flooded regions, where the weathering occurs. It is known that flooded plains in the central Amazon have a high concentration of Andean sediments. In the Amazon Basin, the flood-pulse concept was developed and has a big influence on the ecosystem's productivity in flooded regions (Wittmann et al., 2006; Junk et al., 2011). Atmospheric deposition of P, despite being considered as an important input to regions with low weathering, was not included in our model. Model generated data on atmospheric dust deposition are available on request Mahowald et al. (2005). However, the uncertainties associated with the data are quite high (Mahowald, personal communication). Additionally in regions like the Amazon basin the atmospheric deposition is mostly originating from local redistribution of biotic particles, such as pollen and leave fragments (Mahowald et al., 2005; Pauliquevis et al., 2012).

At the global scale the atmosphere redistributes P from one region to the other, and, therefore if one would like to include this process, one not only has to account for P deposition in some areas, but also for P removal in the other areas. Therefore, we
decided that we could omit this process. Once the soil processes contributing to the P dynamics are better understood, one could try to include these processes, related to wind erosion and dust deposition.

Terrestrial ecosystems have regular losses of litter and soil organic biomass due to run-off, animals, and erosion. Those losses cause losses of Ca, C and P. For simplicity, our model does not account for these processes, since parameterizing these processes might be elusive beyond local scales. However, losses of Ca, C and P due to erosion would have two contradictory consequences:

1. Soil respiration will potentially decrease since the organic matter will no longer be respired in the soil, but rather in the rivers, lakes, and oceans.

2. Including Ca losses may increase weathering in the regions where weathering is currently constrained by high concentration of Ca in the soil solution.

It would be interesting in future studies to include Ca dynamics in vegetation, to test whether that can indeed drive the systems to a steady-state with higher weathering rates.

4.2 Vegetation feedbacks on P availability

We only included three feedbacks; other feedbacks that might also enhance P availability could also be included. We start by discussing the three feedbacks that we have included in this model, and then we discuss other possible feedbacks and how they could be implemented in the model. Our results illustrate the importance of active P uptake by vegetation in maintaining long-term terrestrial productivity. Our model shows how this process could be important in temperate as well as tropical areas. This agrees with the observed spread of plants across the Earth, over 86% having mycorrhizal association (Lambers et al., 2008; Brundrett, 2009). About 20% to 40% of the NPP is invested into root exudation (Chapin et al., 2002). The plant allocation of carbon gained from photosynthesis to root exudates can vary from 5% to 85% (Allen et al., 19374)
2003). Hence, regions with low P availability will use more carbon in symbiosis, with less carbon available for biomass production. The results presented here also agree with our findings in Buendía et al. (2010), where only P dynamics were considered. Including the carbon cost of mycorrhizae symbiosis gave the means to implement the process at the global scale. However, assuming this process to be equally important across ecosystems and in time would result in an unrealistic high concentration of P in vegetation in areas with ample P supply.

Our study suggests that biotic de-occlusion of P is not a long-term solution to P limitation, as the size of the occluded P pool is ultimately constrained by P inputs to the soil. De-occlusion of P was included as an exchange between occluded P and carbon. Despite the high respiration cost associated with this process, in the long term, the P occluded pool is depleted, which is contrary to what we observe in old soils, where the largest P pool is the occluded. Therefore, biotic de-occlusion was assumed to be not relevant for long-term dynamics and not included in the final simulations. In the model, biotic de-occlusion occurs if there is root exudation. In nature, biotic de-occlusion is related to root clustering, which is a morphological adaptation that allows the plant to release carbohydrates. This adaptation is common only in places with very low P availability (e.g. Western Australia.) (Lambers et al., 2008).

Root exudation induces soil respiration and therefore increases the CO$_2$ in soils, which should result in enhanced weathering rates (Berner, 1992). However, under long-term P limitation, vegetation growth is reduced and therefore the overall productivity of the vegetation and autotrophic respiration by roots. With that total respiration and decomposition is reduced. This is because biotic enhancement of weathering also depends on respiration, decomposition, soil stabilization, and water dynamics (Taylor et al., 2009), which ultimately depend on vegetation coverage. Additionally, biotic enhancement of weathering is limited by the supply of primary minerals and thus on crustal uplift (Arens and Kleidon, 2011). Therefore, given our approach to chemical weathering, this feedback will act in the model only in places with high uplift rates. Uplift of minerals into the soil column is a function of soil depth and thereby erosion, which
modulates soil depth. We assume constant topography, which means that erosion is compensated by uplift.

Apart from the mechanisms that were implemented in this model and discussed here, there are other mechanisms that could enhance P availability. In the paragraphs that follow we focus on the effect of the microbial community, independent mineralization of P from organic forms, and animal transfer of P between ecosystems.

Microbial P uptake could be an important mechanism on annual time scales with long-term consequences, preserving P in the ecosystem. Microbial activity is stronger during the wet period, taking up the available P, and with this preventing it from leaching out of the system. When ecosystems dry, microbes die and P becomes available for plant uptake (Resende et al., 2010). Additionally, it has been suggested that microbial P mineralization, which potentially also benefits plants, can be a side-effect of microbial C acquisition (Spohn and Kuzyakov, 2013). Independent mineralization of P, which occurs through enzymatic hydrolysis of phosphate esters, catalyzed by the enzyme phosphohydrolase plays an important role in the P dynamics (Landeweert et al., 2001; McGill and Cole, 1981). In both cases, microbial dynamics could be an important mechanism for preserving P in the ecosystem and making it available for plants when needed. Therefore, the inclusion of microbial contributions to P-availability could be the next step forward. Runyan and D’Odorico (2012) presented an option to model this process in cerrados, which are savanna-like ecosystems in norther Brazil. The challenge is to include these processes at the global scale without site-specific parameterization.

Animals can play an important role in nutrient cycling, as well as for redistributing nutrients between ecosystems (Vanni, 2002; Wardle et al., 2009; de Mazancourt and Schwartz, 2010; Schmitz et al., 2010; Doughty et al., 2013). However, to model animal transport of nutrients between ecosystems information on population densities, animal sizes and how individual move/migrate in the ecosystems needs to be fed to the model (Doughty et al., 2013), this is problematic given the lack of information. To avoid this assumptions we suggest to tackle that question by constraining the P flux from river to terrestrial ecosystems through a tradeoff with carbon availability.
5 Summary and conclusions

We described P dynamics during pedogenesis and its interaction with terrestrial vegetation in a simple but process-based way. Our results strongly suggest that P limitation of biomass production can be an important driver of terrestrial carbon cycling. We did so by coupling a model representing soil evolution accounting for the effects of climate, topography, lithology and vegetation to a model considering P dynamics. P constraints to biomass production result in C to P flexibility in vegetation, which provided the means to infer how much C is allocated to non-structural pools below ground, in order to increase P uptake. Although there are more possible vegetation feedback processes influencing P availability, this manuscript considers only three of them: weathering, de-occlusion, and active P uptake mediated by mycorrhizal associations. Only the two last processes were evaluated as the weathering feedback could not be excluded from the simulations, and therefore could not be tested. We found that active uptake of P (BAU) is important at intermediate-to-old soil stages and across all world climate regions, except for mountain regions. De-occlusion of P, instead, turned out to be important only at intermediate soil stages. Taking into account these dynamics leads to depletion of P in occluded forms in old soils, which contradicts observations that particularly old soils have high concentrations of occluded P. Therefore, the process of biospheric de-occlusion was not included in the final simulations.

The temporal and spatial dynamics of the simulations accounting for BAU are in agreement with observations from chronosequences and spatial P patterns. This is remarkable considering the low amount of site-specific information fed into the model.

Hence, the model presented here can be seen as proof-of-concept on how to consider the interaction between carbon and nutrient limitation at soil-vegetation level, with that constraining vegetation productivity and thereby the global carbon cycle. Based on this approach one could include more vegetation feedback processes to evaluate if and to which degree P limitation is ameliorated.
P constraints to biomass production and their carbon mediated feedbacks to P availability results in geographic variability of biomass production efficiency (BPE). We showed how BPE decreases with soil evolution: it is ultimate highly correlated with weathering. Only model simulations, which considered mycorrhiza mediated active uptake of P, could explain observed biomass patterns, biomass production efficiency, and C : P ratios in vegetation. This indicates how important it is to account for long-term P dynamics and vegetation mediated carbon feedback mechanisms when exploring vegetation functioning and global carbon cycling with process-based models.

Appendix A

Model description

A1 Naming convention

We followed JESSY-SIMBA naming convention. State variables are conventionally named $A^b_c$, where $A$ is a chemical compound, e.g. P, or a state variable such as temperature $T$, $b$ is the state of matter, e.g. s for solid (see Table 1). Fluxes have names of the form $fA^{sa}_{bt}$, where $b$ is start and $t$ the end state (to consider phase change) and $s$ is the start and $a$ the end location. P root uptake from soil to vegetation is thus named $fP_{sv}^{d}$. Pools start with the element, fluxes start with $f$, parameters with $p$, and constants with $c$.

A2 Phosphorus balance equations

$$\frac{d}{dt}P^v_d = fP^{sv}_{d} + fP_{d}^{y} - fP^{v}_{do} - fP^{sv}_{d}$$

$$\frac{d}{dt}P^v_o = fC^{v}_{co} - fC^{v}_{o}$$
\[
\frac{d}{dt} P_y = f P_{dy} - f P_{dy} - f P_{dy} \\
\frac{d}{dt} P_s = f P_{md} + f P_{od} - f P_{dy} - f P_{dy} - f P_{dy} \\
\frac{d}{dt} P_o = f P_{vs} - f C_{og} \\
\frac{d}{dt} P_m = f P_{ms} - f P_{md}
\]

A3 Carbon balance equations

\[
\frac{d}{dt} C_c = f C_{cs} - f C_{cg} - f C_{cy} - f C_{cc} \\
\frac{d}{dt} C_v = f C_{cv} - f C_{cv} \\
\frac{d}{dt} C_o = f C_{ov} - f C_{ov} \\
\frac{d}{dt} C_c = f C_{ga} - f C_{co} - f C_{cy} - f C_{cg} \\
\frac{d}{dt} C_o = f C_{os} - f C_{og} \\
\frac{d}{dt} C_y = f C_{cy} - f C_{cy} - f C_{cy}
\]

A4 Water balance equation

\[
\frac{d}{dt} H_2O_1 = f H_2O_1^{as} - f H_2O_1^{sv} - f H_2O_1^{sa} - f H_2O_1^{sr} - f H_2O_1^{sb}
\]

The model inputs daily precipitation \((f H_2O_1^{as})\) and calculates, based on soil heat, vegetation coverage and vegetation water uptake \((f H_2O_1^{sv})\), soil evapotranspiration \((f H_2O_1^{sa})\), runoff \((f H_2O_1^{sr})\) and base flow \((f H_2O_1^{sb})\).
Calcium balance equations

\[ \frac{d}{dt} Ca_d = f Ca_{pd} - f Ca_{br} \]

\[ \frac{d}{dt} Ca_p = f Ca_{bs} - f Ca_{pb} \]

Appendix B

Comparison to other estimates

Comparisons of modeled P in litter and soil organic biomass with results from Yang et al. (2013a). We run the model considering biotic active P vegetation uptake during 70,000 yr. Present the plot next to each other for a better comparison.

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Table 1. Abbreviations of variable names. State variables are conventionally named \( A^c_b \), where A is a chemical compound, e.g. P, or a state variable such as temperature T, \( b \) is the state of matter, e.g. s for solid (see Table 1). Fluxes have names of the form \( fA^{sa}_{bt} \), where \( b \) is start and \( t \) the end state (to consider phase change) and \( s \) is the start and \( a \) the end location. P root uptake from soil to vegetation is thus named \( fP^{sv}_d \). Pools start with the element, fluxes start with \( f \), parameters with \( p \), and constants with \( c \).

<table>
<thead>
<tr>
<th>element</th>
<th>state</th>
<th>Abbreviation</th>
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<tbody>
<tr>
<td>water</td>
<td>solid</td>
<td>H(_2)O</td>
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<tr>
<td>phosphorus</td>
<td>liquid</td>
<td>P</td>
</tr>
<tr>
<td>calcium</td>
<td>gaseous</td>
<td>Ca</td>
</tr>
<tr>
<td>carbon</td>
<td>dissolved</td>
<td>C</td>
</tr>
<tr>
<td></td>
<td>actively metabolizing biomass</td>
<td>o</td>
</tr>
<tr>
<td></td>
<td>carbohydrates</td>
<td>c</td>
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<td></td>
<td>bound to primary mineral</td>
<td>m</td>
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<tr>
<td></td>
<td>bound to secondary mineral</td>
<td>y</td>
</tr>
<tr>
<td>reservoir</td>
<td>atmosphere</td>
<td>a</td>
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<td></td>
<td>vegetation</td>
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<td></td>
<td>soil</td>
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<td></td>
<td>river channel</td>
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<tr>
<td></td>
<td>bedrock</td>
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**Table 2.** Description of model variables relevant for the P cycle (part 1 of 3).

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<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
<th>Units</th>
</tr>
</thead>
<tbody>
<tr>
<td>pools</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$H_{2}O_{v}$</td>
<td>water in vegetation</td>
<td>m</td>
</tr>
<tr>
<td>$H_{2}O_{s}$</td>
<td>soil water</td>
<td>m</td>
</tr>
<tr>
<td>$H_{2}O_{as}$</td>
<td>frozen soil water</td>
<td>m</td>
</tr>
<tr>
<td>$H_{2}O_{s}$</td>
<td>snow pack</td>
<td>m</td>
</tr>
<tr>
<td>$C_{v}$</td>
<td>C in actively metabolizing vegetation biomass</td>
<td>kgCm$^{-2}$</td>
</tr>
<tr>
<td>$C_{c}$</td>
<td>C in carbohydrates in vegetation</td>
<td>kgCm$^{-2}$</td>
</tr>
<tr>
<td>$C_{s}$</td>
<td>C in litter and soil vegetation biomass</td>
<td>kgCm$^{-2}$</td>
</tr>
<tr>
<td>$C_{a}$</td>
<td>C in organic acids the soil</td>
<td>kgCm$^{-2}$</td>
</tr>
<tr>
<td>$C_{y}$</td>
<td>C in recalcitrant forms in the soil</td>
<td>kgCm$^{-2}$</td>
</tr>
<tr>
<td>$C_{g}$</td>
<td>total CO$_2$ in soil</td>
<td>molesCm$^{-2}$</td>
</tr>
<tr>
<td>$P_{m}$</td>
<td>P in bedrock</td>
<td>gPm$^{-2}$</td>
</tr>
<tr>
<td>$P_{s}$</td>
<td>P in primary minerals</td>
<td>gPm$^{-2}$</td>
</tr>
<tr>
<td>$P_{d}$</td>
<td>P in soil solution</td>
<td>gPm$^{-2}$</td>
</tr>
<tr>
<td>$P_{v}$</td>
<td>P available in vegetation</td>
<td>gPm$^{-2}$</td>
</tr>
<tr>
<td>$P_{o}$</td>
<td>P actively metabolizing vegetation biomass</td>
<td>gPm$^{-2}$</td>
</tr>
<tr>
<td>$P_{y}$</td>
<td>P in soil organic matter</td>
<td>gPm$^{-2}$</td>
</tr>
<tr>
<td>$P_{c}$</td>
<td>P occluded in secondary minerals</td>
<td>gPm$^{-2}$</td>
</tr>
<tr>
<td>$Ca_{p}$</td>
<td>Ca in primary minerals</td>
<td>kgCm$^{-2}$</td>
</tr>
<tr>
<td>$Clay_{s}$</td>
<td>Secondary minerals in the soil</td>
<td>kgm$^{-2}$</td>
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Table 3. Description of model variables relevant for the P cycle (part 2 of 3).

<table>
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<th>Description</th>
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</thead>
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<tr>
<td>$f_{\text{H}_2\text{O}}^{\text{as}}$</td>
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<td>ms$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{H}_2\text{O}}^{\text{sv}}$</td>
<td>root water uptake</td>
<td>ms$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{H}_2\text{O}}^{\text{va}}$</td>
<td>transpiration</td>
<td>ms$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{H}_2\text{O}}^{\text{sa}}$</td>
<td>evaporation</td>
<td>ms$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{H}_2\text{O}}^{\text{ac}}$</td>
<td>surface runoff</td>
<td>ms$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{H}_2\text{O}}^{\text{sc}}$</td>
<td>baseflow</td>
<td>ms$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{H}_2\text{O}}^{\text{sl}}$</td>
<td>snow melt</td>
<td>ms$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{av}}$</td>
<td>vegetation gross carbon uptake GPP</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{co}}$</td>
<td>synthesis of actively metabolizing vegetation biomass</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{cv}}$</td>
<td>total autotrophic respiration</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{ca}}$</td>
<td>leaf respiration</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{cg}}$</td>
<td>root respiration</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{cs}}$</td>
<td>litter production</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{sr}}$</td>
<td>organic acid root exudation</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{cy}}$</td>
<td>C fixation into secondary minerals</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{cg}}$</td>
<td>microbial respiration of organic acids</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{sf}}$</td>
<td>losses of recalcitrant carbon</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{ss}}$</td>
<td>soil respiration</td>
<td>kgC m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{as}}$</td>
<td>input of CO$_2$ though rain</td>
<td>moles m$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$f_{\text{C}}^{\text{sa}}$</td>
<td>soil CO$_2$ efflux</td>
<td>moles m$^{-2}$s$^{-1}$</td>
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Table 4. Description of model variables relevant for the P cycle (part 3 of 3).

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
<th>Units</th>
</tr>
</thead>
<tbody>
<tr>
<td>$fP_m^{bs}$</td>
<td>physical weathering</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$fP_m^s$</td>
<td>chemical weathering</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$fP_{md}$</td>
<td>occlusion</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$fP_{dy}$</td>
<td>litter production</td>
<td>gCm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{o}$</td>
<td>P_{d} root exudation</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{d}$</td>
<td>P vegetation uptake</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{do}$</td>
<td>P synthesis to actively metabolizing vegetation biomass</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{od}$</td>
<td>mineralization</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{sr}$</td>
<td>losses</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{yd}$</td>
<td>erosion of occluded P</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{y}$</td>
<td>biotic deocclusion</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{pd}$</td>
<td>Ca chemical weathering</td>
<td>kgCam$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{bs}$</td>
<td>Ca physical weathering</td>
<td>kgCam$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>$nP_{sr}$</td>
<td>Secondary minerals erosion</td>
<td>kgm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>potential</td>
<td>$aP^{sv}_{do}$ P active uptake function</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>fluxes</td>
<td>$aP^{sv}_{do}$ P potential synthesis to actively metabolizing vegetation biomass</td>
<td>gPm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td></td>
<td>$aC_{co}$ C potential synthesis to actively metabolizing vegetation biomass</td>
<td>kgCm$^{-2}$s$^{-1}$</td>
</tr>
<tr>
<td>states</td>
<td>LAI leaf area index</td>
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</tbody>
</table>

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Table 5. Description of model parameters. Parameters with the reference \(^a\) are set by personal assessment (see text) while parameters marked by \(^b\) are calibrated to values, which lead to reasonable model output, considering the data used to evaluate the model.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Description</th>
<th>Value</th>
<th>Units</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>ceopot</td>
<td>correction of pot. evaporation</td>
<td>1.30</td>
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<tr>
<td>(p\Delta Z_s)</td>
<td>soil depth</td>
<td>1.0</td>
<td>m</td>
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<tr>
<td>(\epsilon_{\text{lue}})</td>
<td>factor for light-use efficiency</td>
<td>120</td>
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</tr>
<tr>
<td>(\epsilon_{\text{wue}})</td>
<td>factor for water-use efficiency</td>
<td>(3.5 \times 10^{-10})</td>
<td></td>
<td></td>
</tr>
<tr>
<td>(p\tau_{C_v})</td>
<td>carbon residence time in vegetation</td>
<td>(3.1 \times 10^{-8})</td>
<td>s</td>
<td></td>
</tr>
<tr>
<td>(p\tau_{C_s})</td>
<td>soil carbon turnover time scale</td>
<td>(1.2 \times 10^{-9})</td>
<td>s</td>
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<td>(pq10ss)</td>
<td>Q10 value of litter decomposition</td>
<td>2.0</td>
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</tr>
<tr>
<td>(p\tau_{C_v}^{\text{c}})</td>
<td>starch residence time in vegetation</td>
<td>(1.08 \times 10^{-08})</td>
<td>s</td>
<td>(^b)</td>
</tr>
<tr>
<td>(p\tau_{P_v}^{\text{d}})</td>
<td>available residence time in vegetation</td>
<td>(1.08 \times 10^{-09})</td>
<td>s</td>
<td>(^a)</td>
</tr>
<tr>
<td>(p\tau_{P_v}^{\text{yd}})</td>
<td>P occlusion rate</td>
<td>0.00005</td>
<td></td>
<td>(^b)</td>
</tr>
<tr>
<td>(p\tau_{aP_v}^{\text{sd}})</td>
<td>active uptake scaling factor</td>
<td>(2.5 \times 10^{-7})</td>
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<td>(^a)</td>
</tr>
<tr>
<td>(p\text{My}<em>{C</em>{cg}})</td>
<td>mycorrhizae respiration rate</td>
<td>0.95</td>
<td></td>
<td>(^a)</td>
</tr>
<tr>
<td>(pBP_{C_{cg}})</td>
<td>growth respiration rate</td>
<td>0.25</td>
<td></td>
<td>(^a)</td>
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<tr>
<td>(p\text{maxf}<em>{P</em>{do}})</td>
<td>maximum P to biomass production rate</td>
<td>0.8</td>
<td></td>
<td>(^a)</td>
</tr>
<tr>
<td>(pP)</td>
<td>concentration of P in primary minerals</td>
<td>lithology</td>
<td>(kg \text{C m}^{-3})</td>
<td>(^a)</td>
</tr>
<tr>
<td>(pCa)</td>
<td>concentration of Ca in primary minerals</td>
<td>lithology</td>
<td>(kg \text{C m}^{-3})</td>
<td>Arens (2013)</td>
</tr>
<tr>
<td>(cCP_s^\text{y})</td>
<td>P de-occlusion exchange rate</td>
<td>(6 \text{ mol C : 1 mol P})</td>
<td></td>
<td>(^a)</td>
</tr>
<tr>
<td>(cCP_v^\text{y})</td>
<td>actively metabolizing biomass C to P ratio</td>
<td>2</td>
<td>(kg \text{C (g P)}^{-1})</td>
<td>(^a)</td>
</tr>
</tbody>
</table>
Fig. 1. A conceptual diagram showing the feedback processes by which vegetation can affect P cycling in the pedosphere and the C cycle. Plants with abundant P supply can sequester carbon from the atmosphere and store it in biomass (path 1), which results in an overall negative feedback on atmospheric CO$_2$ concentrations. P limited vegetation has a reduced capacity for producing biomass. Instead, it builds-up non-structural carbon reserves that can lead to increased root exudation into the soil (path 2). Increasing root exudates increases soil respiration, potentially leading to (A) enhanced rock weathering; (B) P de-occlusion from the secondary minerals; or (C) more efficient P uptake by means of symbiosis with root mycorrhizae.
Fig. 2. Theoretical relationship of P supply and biomass productivity (BP) efficiency. The efficiency of BP (i.e. the slope of the line) depends on the supply of P to the ecosystem and on the degree that vegetation can enhance P availability. The maximum efficiency of BP occurs when vegetation grows under sufficient P (case 1, solid black line). In the other case, P-availability limits BP (case 2, grey solid line). The carbon-energetic cost of obtaining P depends on vegetation-mediated processes within the soil: e.g., if root exudations results in P gain by vegetation; then the ecosystem can produce new biomass with this biotically acquired P, which then increases BP efficiency (dotted grey line).
The terrestrial P cycle

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Fig. 3. Diagram of the model structure (after Porada et al., 2010). The model uses daily climate data, topography and lithological classes as input. It has three main components, bedrock, soil and vegetation in which fluxes of primary minerals from the bedrock are calculated. Vegetation dynamics drives water, carbon, and phosphorus dynamics. The model calculates e.g., weathering rates, gross primary productivity, and biomass production. For an extensive list of the model outputs refer to Table 4.
Fig. 4. The C-P biogeochemical model: symbols in bold denote pools and symbols, in italic denote fluxes between pools. Black arrows show fluxes in the standard version of the model (W-P); blue arrows show P de-occlusion and the associated carbon flux to recalcitrant forms (BD); red arrows show P active uptake mediated by mycorrhizae and its associated carbon respiration (BAU). The model has a switch for turning on and off active P uptake (blue fluxes) and biotic de-occlusion of P (red fluxes). For definition of symbols refer to Table 4.
Fig. 5. Map of the lithological classes and their respective apatite concentrations (g Apatite m$^{-3}$). Lithological classes taken from Amiotte Suchet and Probst (1995), values for the concentration of apatite were calculated from Newman (1995).
Fig. 6. Diagram showing the modeled carbon feedbacks on the P availability. Simulation scenarios are defined with respect to these feedbacks. Dashed line represents the effects that can be turned off and pointed line represents the unrealistic Control-P.
Fig. 7. Carbon in terrestrial vegetation biomass of the control run. In this scenario the vegetation growth is constrained only by water, light, and temperature so that vegetation biomass reaches its maximum everywhere around the globe, while not accounting for P limitation. In this simulation biomass reaches 24 kg C m\(^{-2}\) in some tropical areas.
Fig. 8. Difference in the standing biomass of the control with the different feedback scenarios for the world after 50,000 yr of soil evolution. Biomass differences between the control simulation (see Fig. 7) and W-P (top-left), BD (bottom-left), BAU (top-right), BAU-BD (bottom right). W-P shows the greatest differences with the control scenario: more than 50% biomass reduction worldwide, the differences are high in temperate coastal areas (e.g., Chile, Guyana, and South Africa). BD (bottom-left) shows a very similar pattern to the W-P scenario, with only small differences in temperate regions. Biotic active uptake (BAU) (top-right panel) shows much smaller differences to the control, with the absolute greatest differences located in the tropics. BAU-BD (bottom right) shows a pattern almost identical to the BAU scenario, with only very little differences.
Fig. 9. GPP (top panels) and BP (bottom panels) zonal averages after 10,000 (left panels) and 50,000 (right panels) years of soil formation. Different scenarios considered are compared with data based estimated GPP (Beer et al., 2010) and model based estimated NPP (Cramer et al., 1999).
Fig. 10. P pools in the soil considering active P vegetation uptake after 70,000 simulated years of pedogenesis.
Fig. 11. Carbon fluxes and pools in vegetation considering active P vegetation uptake after 70,000 yr of soil evolution.
Fig. 12. Biomass production efficiency (BPE), P weathering rates, and C to P molar ratio in vegetation changes during pedogenesis. Top plots represent the dynamics in recent soils (after about 10,000 yr of soil evolution). The intermediate plots represent intermediate soil ages of about 50,000 yr and the bottom plots represent old soils about 150,000 yr of soil evolution. Biomass production is more efficient in temperate regions in both, young and old, soils. In tropical regions, BPE decreased in time, reaching a status of very low efficiency, particularly in central Africa and Western South America. P weathering rates (central panels), decrease everywhere with time. In the oldest soils, weathering is high in places at high elevations. Also locations with shale lithology produced relatively high weathering rates, compared to those with sandstone. The C to P molar ratio in vegetation increased in time. In tropical regions, the ratio was correlated with weathering rates.
Fig. 13. Gross primary productivity vs biomass production after 10,000 and 50,000 and 150,000 yr of soil evolution for the BAU scenario.
**Fig. A1.** Global maps of P in the litter and soil organic matter in the soils after upper panel data repotted from Yang et al. (2013a) and the lower panel is a modeling result after 70,000 yr soil evolution, taking into account the active P-uptake mediated by root exudation.
Fig. A2. Global maps of P in occluded forms upper plot repotted from Yang et al. (2013a) lower map model results after 70,000 years of soil evolution, taking into account the active P uptake.