Hydration status and diurnal trophic interactions shape microbial community function in desert biocrusts

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Received: 23 April 2017 – Discussion started: 19 June 2017
Revised: 2 October 2017 – Accepted: 21 October 2017 – Published: 1 December 2017

Abstract. Biological soil crusts (biocrusts) are self-organised thin assemblies of microbes, lichens, and mosses that are ubiquitous in arid regions and serve as important ecological and biogeochemical hotspots. Biocrust ecological function is intricately shaped by strong gradients of water, light, oxygen, and dynamics in the abundance and spatial organisation of the microbial community within a few millimetres of the soil surface. We report a mechanistic model that links the biophysical and chemical processes that shape the functioning of biocrust representative microbial communities that interact trophically and respond dynamically to cycles of hydration, light, and temperature. The model captures key features of carbon and nitrogen cycling within biocrusts, such as microbial activity and distribution (during early stages of biocrust establishment) under diurnal cycles and the associated dynamics of biogeochemical fluxes at different hydration conditions. The study offers new insights into the highly dynamic and localised processes performed by microbial communities within thin desert biocrusts.

1 Introduction

Large tracks of arid lands are often covered by thin biological soil crusts (hereafter, biocrusts) that, in the absence of significant vegetation cover, play an important role in arid ecosystems. Biocrusts serve as biodiversity “hotspots” (Belnap et al., 2016) and act as ecosystem engineers to promote the rehabilitation of eroded soils in arid lands (Bowker, 2007). The photoautotrophs inhabiting biocrusts support rich and diverse food webs and provide the main source of organic carbon covering over 70% of arid land surface area (about 30% of all terrestrial surfaces; Belnap and Lange, 2002; Mager, 2010). Biocrust microbial activity produces extracellular organic exudates that alter the immediate environment by supporting a stable structure and altering the water retention and transport properties of the biocrusts (Mazor et al., 1996; Belnap and Lange, 2002; Belnap, 2003; Rodríguez-Caballero et al., 2015). The resulting modification of local hydrological processes, such as infiltration run-off and water storage (Chamizo et al., 2012), enhances the capability of other organisms to cope with water scarcity (Chamizo et al., 2016; Faist et al., 2017). Furthermore, this water-regulating function of biocrusts also protects the soil surface against wind and water erosion (Belnap and Gillette, 1998; Warren, 2001).

Evidence suggests that biocrusts are a locally and globally important component of the ecosystem in terms of biogeochemical fluxes; arid land biocrusts affect global cycles of carbon and nitrogen (Weber et al., 2016b). Biocrusts regulate carbon dioxide efflux through soil by fixing $\sim 0.6 \text{ Pg C year}^{-1}$, which is about 9% of the net primary productivity of this ecosystem (Sancho et al., 2016; Elbert et al., 2012). Their contribution to nitrogen fixation from the atmosphere is even more significant, evaluated as about 26 Tg year$^{-1}$ and corresponding to about 40% of the global terrestrial biological nitrogen fixation (Elbert et al., 2012; Ciais et al., 2014). Although biocrust contribution to terrestrial nitrogen fixation is considerably high, arid land ecosystems remain largely nitrogen limited due to the substantial losses of nitrogen gas caused by abiotic (temperature, pH) and biotic (nitrification, denitrification) processes (Peterjohn and Schlesinger, 1990; McCalley and Sparks, 2009). The global emission of reactive nitrogen (such as NO, HONO) from biocrusts has been estimated at about 20% of the
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globally emitted reactive nitrogen compounds from natural soils (Weber et al., 2015).

Biocrusts are sensitive and highly vulnerable systems to anthropogenic and natural disturbances, leading to the erosion of the invaluable microbial community (Kuske et al., 2012). Natural recovery of biocrusts is a slow process (multiple decades) (Weber et al., 2016a), and the recovery rates may vary widely depending on precipitation, soil texture, or carbon content (Weber et al., 2016a). The recovery stage follows a general successional pattern beginning with surface soil colonisation by mobile Cyanobacteria such as Microcoleus vaginatus (Büdel et al., 2009; Zaady et al., 2010).

The settlement of photoautotrophic organisms is followed by other phototrophic, heterotrophic, and chemooautotrophic microorganisms, algae, fungi, mosses, and lichens (Pepe-Ranney et al., 2016). Most established biocrusts consist of microscopic and macroscopic organisms within the top few centimetres of the soil surface (around 5 mm thick for cyanobacterial crusts and up to 5 cm thick for moss crusts). A typical biocrust community consists of hundreds of species representing different levels of trophic interactions that enable an entire arid land ecosystem to function systemically (Bowker et al., 2010a, b).

The composition and structure of a biocrust are determined by several environmental factors. On a local scale, soil properties such as texture, nutrient level, and pH are the main determinants (Bowker et al., 2016). On a global or regional scale, the characteristics of a biocrust community differ with climatic regions (from cold to warm deserts), soil type, and crust age since last disturbance (Garcia-Pichel et al., 2013; Bowker et al., 2016). Regional climatic variables such as the amount of precipitation or the potential evapotranspiration influence the biomass of Cyanobacteria and other photoautotrophs that as a consequence define the community composition (Isichei, 1990; Hagemann et al., 2015; Barnard et al., 2015). Studies have shown that cyanobacterial crust distribution and activity are highly correlated with periods between rain events and soil water availability rather than the precipitation amount of a single rain event (Lange, 2003; Cable and Huxman, 2004; Büdel et al., 2009). Thus, the response of microbial activity to wetting events, such as precipitation, is a crucial factor in the ecology of biocrusts.

Notwithstanding the importance of these ecosystems, quantitative studies using mathematical or computational approaches are scarce and interrelations among the biological, physical, and chemical processes that underlie this sensitive ecosystem remain unclear. Statistical analyses have served as the main means to deduce the impacts of various environmental factors on observed biocrust response in the majority of field and laboratory studies (Barger et al., 2006; Grote et al., 2010; Bowker et al., 2010a; Castillo-Monroy et al., 2011; Maestre et al., 2013). Process-based models have also been developed for biocrusts of lichens and mosses (Porada et al., 2013, 2014, 2016, 2017). These studies estimate their contribution to the carbon uptake and nitrous oxide emissions on a global scale under various climatic conditions.

This study reports a mechanistic model for the early stages of biocrust formation and key biophysical and chemical processes. We construct a representation of hydrological processes within a biocrust and the trophic interactions among key members of a biocrust microbial community. The model includes a detailed account of the physical domain available for microbial life (simple rough surfaces) and the consequences of different hydration conditions on connectivity and the transport of nutrients, gas, temperature, and light. The model also considers dynamic chemical processes. The key ingredient in biocrust functioning is the highly dynamic and spatially self-organising microbial community. For simplicity, we considered four microbial groups: photoautotrophs (primarily Cyanobacteria), aerobic heterotrophs, anaerobic heterotrophs (denitrifiers), and chemoautotrophs (Garcia-Pichel and Belnap, 2002; Johnson et al., 2005, 2007; Abed et al., 2010, 2013; Nunes da Rocha et al., 2015) and consider their role in carbon and nitrogen cycling. Figure 1 summaries the model in terms of processes, variables, parameters, and simulated results in this work.

The organisation of this paper is as follows: we first introduce the key physical and chemical processes in the mechanistic model. Next, the biochemical feedback of microbial activity and its spatial organisation is investigated. The results of this model are compared with data obtained from laboratory experiments. Finally, we provide new insights into the ecological functions of unsaturated soil structures in established biocrusts in arid regions.

2 A mechanistic model of desert biocrusts

The study was motivated by an interest in biocrusts as a model ecosystem supporting a multispecies microbial community that interacts to a limited spatial extent under large environmental gradients (Bowker et al., 2014). We employ the individual-based modelling of microbial processes in the presence of sharp environmental gradients in resources and conditions. The model first addresses the physical domain and its dynamic characteristics that vary with hydration conditions. Chemical and biological processes are then introduced into the physical domain (associated primarily with the aqueous phase and its distribution).

2.1 The biocrust physical domain

We use a modified rough surface patch model (Št’oviček et al., 2017; Kim and Or, 2016) to represent the top millimetres to centimetres of soil where most biocrusts develop (see Fig. 2a). For the physical domain, we consider a vertical cross section of a biocrust that considers rough soil grains and the gas pathways between grains (described in 2-D but in a simplified fashion including 3-D features in a spatial
Figure 1. A summary of the desert biocrust model (DBM). The DBM includes various physical, chemical, and microbial processes occurring within biocrusts. Variables and parameters used in modelling are listed and the main assumptions for each process are summarised. In this work, we focused on how hydration conditions under diurnal cycles shape chemical and biological profiles within desert biocrusts.
2.2.1 Light irradiance on the surface

Light determines the photosynthetic activity of the phototrophs (e.g. Cyanobacteria) within the biocrust (Berner and Evenari, 1978; Davies et al., 2013). To represent light penetration and the diurnal day–night cycle, we express irradiance, \( I(z; t) \), as a function of depth \( z \) and time \( t \):

\[
I(z; t) = \begin{cases} 
I_0 (1 - \cos(\omega t + \phi_I)) e^{-z/\delta_p} & \text{day} \\
0 & \text{night}
\end{cases},
\]

where \( I_0 \) is maximum irradiance (at midday on the biocrust surface). Incident irradiance at the surface is given by the period \( P \equiv \frac{2\pi}{\omega} \) (24 h) and \( \phi_I = 0 \) (with \( t = 0 \) at sunrise, 06:00); \( \delta_p \) is the light characteristic penetration depth. The values of \( I_0 \) and \( \delta_p \) regulate the activity and spatial location for an optimised growth of phototrophs in the model. The sinusoidal function with \( I_0 \) and \( \omega \) can be changed with respect to the specific location of the biocrust and the season of the year (and even the local aspect and slope of the surface). The value of \( \delta_p \) varies from about \( 10^{-4} \) to \( 10^{-3} \) m depending on the amount of mineral or soil texture (grain size distribution) (Garcia-Pichel and Bebout, 1996). In this work, we chose a maximum irradiance of \( I_0 = 500 \mu\text{mol m}^{-2} \text{s}^{-1} \) corresponding to the light intensity of overcast sky (assuming that a biocrust shows activity when it is wet, i.e. during rainy days). The calculations consider a constant light penetration depth of 0.2 mm (Garcia-Pichel and Belnap, 1996), and only vertical penetration is considered in the current work. The resulting distribution of irradiance in the model is depicted in Fig. 2b.

2.2.2 Temperature dynamics

The profile of soil temperature varies with time and space following a periodic function coupled with light incidence. We consider ambient temperature as a sinusoidal function for the surface boundary condition (Phillips et al., 2011):

\[
T(z = 0, t) = \overline{T} + A_0 \sin(\omega t + \phi_T),
\]

where \( \overline{T} \) is the average temperature on the surface and \( A_0 \) is the diurnal amplitude. The period \( P \equiv \frac{2\pi}{\omega} \) is assumed to be 1 day and the phase is set to be 0, so the maximum temperature corresponds to the maximum of light intensity (midday). Considering a homogenous domain with uniform hydration...
status, an analytical solution for a 1-D heat equation with sinusoidal temperature boundary condition (Eq. 2) yields a dynamic description of a diurnal soil temperature profile (a similar solution is obtained for seasonal profiles):

\[ T(z, t) = \bar{T} + A_0 e^{-\frac{\pi^2}{d^2}} \sin \left( \omega t + \phi_T - \frac{z}{d} \right) , \]

where \( d \) is a characteristic damping depth of the domain given by \( d = \sqrt{\frac{a \rho_c}{\lambda}} \) and \( \alpha = \frac{1}{c_p} \) is thermal diffusivity (for details, see Sect. S1 in the Supplement). Thermal diffusivity is a function of hydration conditions; i.e. conductivity increases with wetness. Soil temperature distribution over depth during a diurnal cycle for wet and dry conditions is illustrated Fig. 2b.

2.3 Biocrust biogeochemical processes: mass transfer and inorganic C and N partitioning

The chemical environment of soil is strongly influenced by its microbial activity (especially at the main biocrust region). For example, the pH of biocrust is altered diurnally due to microbial respiration (release of protons and bicarbonates) and photosynthesis, with CO\(_2\) removal significantly modifying porewater pH. These, in turn, affect nutrient availability and mobility, CO\(_2\) dissolution rates, and the solubilisation of soil minerals (Belnap et al., 2002). The model includes certain essential chemical processes: diffusion, gas–liquid-phase partitioning, and acid–base dissociation that affect microbial activity within typical biocrusts.

2.3.1 Gas diffusion with the biocrust

Unsaturated conditions dominate microbial life in desert biocrusts and support unhindered gas diffusion most of the time. The gas diffusion coefficient is on the order of \( 10^{-6} \text{ m}^2\text{s}^{-1} \), which is about \( 10^4 \) times greater than that in the aqueous phase. In the largely aerated biocrust, the partial pressures of soil gas near the surface equilibrate with the atmospheric level almost instantly (it takes a few seconds to aerate soil at a depth of a few millimetres). In contrast, when the soil surface becomes especially wet, the aqueous-phase configuration may temporarily hinder gas diffusion and delay such instantaneous partial pressure equilibration. Thus, an understanding of water configuration within the domain is necessary. In unsaturated soils, water is held on the rough soil surface due to the capillary force (given by the Young–Laplace equation) and absorbed water film (van der Waals force). The abstract model (Fig. 2a) provides a means for calculating the proportion of water held at the given hydration conditions from pre-assigned soil properties. This yields the degree of saturation after normalisation with the volume of void space and the local gas/water content (proportion of gas to water) of each spatial element (a patch). We used these local properties for gas-phase invasion probability and local diffusion coefficients. When a patch at location \( r \) is connected to the atmosphere (invasion percolation), constant boundary conditions at the gas phase are assigned with respect to the atmospheric mixing ratio of each gaseous element instead of resolving gas diffusion at the near surface (assuming instant equilibration).

2.3.2 Mass transfer between gas and liquid

The mass transfer rate across the gas–liquid interface can be determined by using Fick’s law and the film model:

\[ \frac{dC}{dt} = -\frac{A_{lv}}{d_{tot}} \left( \frac{D_g}{1 - \Theta} - \frac{D_l}{\Theta} \right) (C^l - C^g) \]

\[ \equiv -k_{l\rightarrow g} (C^l - C^g) , \]

where \( C^l \) and \( C^g \) are substrate concentration in the liquid and gas phases, respectively; \( A_{lv} \text{ m}^2\text{m}^{-3} \) is the specific liquid–vapour interfacial area; \( d_{tot} \) is the effective thickness of soil pore space; \( D_g \) and \( D_l \) are diffusion coefficients for the liquid and gas phases; \( \Theta \) is the degree of saturation; and \( k_{l\rightarrow g} \) is the net mass transfer rate across the interface, which is a function of hydration conditions.

The proposed model allows us to calculate the specific liquid–vapour interfacial area and the effective thickness of void space (Kim and Or, 2016). For instance, the model estimates that \( A_{lv} \approx 10^3 \text{ m}^2\text{m}^{-3} \) and \( d_{tot} \approx 10^{-5} \text{ m} \) at \( \Theta = 0.5 \) (half saturation). These values are consistent with other studies that have used a pre-assigned water retention curve (Zand-Parsa and Sepaskhah, 2004). Considering that the gas diffusion coefficient is on the order of \( 10^{-6} \text{ m}^2\text{s}^{-1} \), the net mass transfer rate between the two phases is \( \sim 10^4 \text{ m}^2\text{s}^{-1} \) in aerated soils. This implies that the concentration at the liquid phase also equilibrates almost instantly to the concentration at the gas phase. Even when the soil is nearly saturated, the rate is \( \sim 1-10 \text{s}^{-1} \) (\( \Theta \rightarrow 1 \), \( \Theta \approx \Theta_0 \)). For comparison, studies on waste water treatment used the rate of \( \sim 6.9 \times 10^{-5} \text{s}^{-1} \) (Buhr and Miller, 1983; Yang, 2011). Thus, mass transfer between gas and liquid in unsaturated soils is assumed to be rapid, and the concentration at each phase is always at equilibrium following Henry’s law:

\[ C^{l*, g} = H_{cc}(T)C^{g*} , \]

where \( H_{cc}(T) \) is a dimensionless Henry’s constant at temperature \( T \); \( H_{cc} = H_{cc}^{S} e^{-\Delta_{solnH} R \left( \frac{1}{T} - \frac{1}{T^*} \right)} \) where \( \Delta_{solnH} \) is the enthalpy of solution, \( R \) is the gas constant, and “S” refers to the standard condition (\( T^* = 298.15 \text{ K} \)) (Sander, 1999).

2.3.3 Dissociation of chemical substances

To evaluate the available CO\(_2\) in soil biocrust water, one must consider the open-system behaviour of CO\(_2\) and the different species of dissolved inorganic carbon (DIC), carbonic acid (H\(_2\)CO\(_3\)), bicarbonate (HCO\(_3^-\)), and carbonate (CO\(_3^{2-}\)). The relative amounts of such DIC species can be determined by
the concentration of protons, or the pH of the solution. Considering that most desert soils are alkaline (Bresler et al., 2012) (implying the predominant DIC species to be bicarbonate), the determination of the amount of dissolved CO$_2$ in soil is essential to the growth and functioning of autotrophs and to distinguishing between abiotic and biotic processes for CO$_2$ efflux estimation. Assuming that other chemical species are inert, the model includes the following geochemical reactions focusing on carbon and nitrogen dynamics in soil.

\[
\begin{align*}
H_2O & \rightleftharpoons OH^- + H^+ \quad (R1) \\
CO_2(aq) + H_2O & \rightleftharpoons HCO_3^- + H^+ \quad (R2) \\
HCO_3^- & \rightleftharpoons CO_3^{2-} + H^+ \quad (R3) \\
NH_3(aq) + H^+ & \rightleftharpoons NH_4^+ \quad (R4) \\
HNO_2(aq) & \rightleftharpoons NO_2^- + H^+ \quad (R5) \\
CaCO_3^{0}(aq) & \rightleftharpoons Ca^{2+} + CO_3^{2-} \quad (R6)
\end{align*}
\]

Some mathematical models have introduced pH estimation for systems with phototrophs under light–dark cycles, such as algal ponds (Buhr and Miller, 1983; Yang, 2011; Gomez et al., 2014) and phototrophic biofilms (Wolf et al., 2007). The algal pond models invoke solution equilibrium and charge neutrality and employ differential algebraic equations to estimate pH, while the phototrophic biofilm models consider acid–base reactions with rate equations by proposing near-equilibrium kinetics. The unsaturated conditions in desert biocrusts with large air–liquid interfacial areas and high mass transfer rates require special treatment. We adopted a similar approach of kinetics with charge balance (Wolf et al., 2007). In addition, the range of geochemical reactions was extended by including nitrous acid (HONO) and nitrous oxide (N$_2$O) to investigate nitrogen-related gas emissions from biocrusts. In this work, calcium is considered to enable the evaluation of the biogenic precipitation of calcium carbonate in biocrust formation. All the kinetics are based on the local concentration of each substrate in porewater with an assumption of water activity 1.

The equilibrium gas-phase concentrations of O$_2$, CO$_2$, NH$_3$, N$_2$O, and HONO are considered in the model according to Henry’s law. Considered reactions for gas- and liquid-phase partitioning and precipitation are listed below.

\[
\begin{align*}
CO_2(aq) & \rightleftharpoons CO_2(g) \quad (R7) \\
NH_3(aq) & \rightleftharpoons NH_3(g)(volatilisation) \quad (R8) \\
HNO_2(aq) & \rightleftharpoons HNO_2(g) \quad (R9) \\
N_2O_3(aq) & \rightleftharpoons N_2O_3(g) \quad (R10) \\
CaCO_3^{0}(aq) & \rightleftharpoons CaCO_3(s)(precipitation) \quad (R11)
\end{align*}
\]

The values and detailed kinetic equations used in the model are summarised in Sect. S2.

### 2.4 Microbial community in desert biocrust ecosystem

Advances in molecular taxonomic techniques and DNA sequencing have greatly expanded our knowledge on microbial community structure and diversity in biocrusts. These data generally delineate the interplay between multilevel trophic interactions (Bowker et al., 2011; Nunes da Rocha et al., 2015; Pepe-Ranney et al., 2016) and surrounding environmental conditions (Caruso et al., 2011). Biocrusts host a complex community of diverse autotrophs and heterotrophs (hundreds of species including about 20 generic or subgeneric taxa of Cyanobacteria) (Bowker et al., 2010a, b; Garcia-Pichel et al., 2013). Considering biocrusts as independent and self-sufficient ecosystems, the intrinsic diversity found in this system should not come as a surprise. The incorporation of the natural microbial diversity found in biocrusts is beyond the present capabilities of most models. Hence, we opted for a representation of the main microbial actors for modelling the associated biogeochemical cycles in a cyanobacterial crust.

#### 2.4.1 Microbial community and trophic interactions

Four functional microbial groups are represented in the in silico microbial model of a desert biocrust: diazotrophic photoautotrophs (that are able to fix atmospheric carbon and nitrogen), aerobic heterotrophs, anaerobic heterotrophs (denitrifiers, strictly anaerobes using NO$_3^-$ as a terminal electron acceptor), and chemoautotrophs (nitrifiers). These groups are chosen to elucidate the interlinked functionality of C/N cycling in a biocrust microbial community. Thus, we considered the following substrates in the soil solution that support microbial activity: oxygen (O$_2$), dissolved inorganic carbon, DIC, (CO$_2$ / HCO$_3^-$), ammonium (NH$_4^+$), oxidised nitrogen species (NO$_3^-$, NO$_2^-$), and organic carbon (CH$_2$O as an elementary form of polyglucose). Here, phototrophically produced CH$_2$O is assumed to be the primary carbon source available, which can be transformed into extracellular polymeric substances (EPSs) depending on environmental conditions. Other chemical species, Ca$^{2+}$, CO$_3^{2-}$, N$_2$O, NH$_3$, and HNO$_2$, are included for the study of chemical reactions but are not directly utilised by these microbial species.

The four microbial groups interact based on prescribed stoichiometric relations (Table S5 in Sect. S4). These stoichiometric relations require the photoautotrophs to be classified into four subgroups (Wolf et al., 2007) using one inorganic carbon source and one inorganic nitrogen source during photosynthesis (i.e. CO$_2$ + NH$_4^+$, HCO$_3^-$ + NH$_4^+$, CO$_2$ + NO$_3^-$, and HCO$_3^-$ + NO$_3^-$). Aerobic heterotrophs use CH$_2$O as an electron donor and O$_2$ as an electron acceptor, and NH$_3$$_2$ as a nitrogen source. Anaerobic heterotrophs (denitrifiers) use CH$_2$O as an electron donor, NO$_3^-$ as an electron acceptor and a nitrogen source. As obligate anaerobes, their growth is inhibited by the presence of oxygen. Chemoautotrophs are described in two subgroups, considering two oxidation processes: firstly, ammonia to nitrite by ammonia oxidising bacteria (AOB) and, secondly, nitrate to nitrite by nitrite-oxidising bacteria (NOB).
By using Monod-type kinetics with limiting substrates, the growth rate of species $i$ with a limiting factor $j$ can be written as

$$\mu_i = \mu_{\text{max},i} \min[f^1_i, f^2_i, \ldots, f^j_i].$$  \hspace{1cm} (6)

where $\mu_{\text{max},i}$ is the maximum growth rate of species $i$ and the Monod factors are of two types, $f^j_i = \frac{C_j}{K_{s,j}^j + C_j}$ (when nutrient $j$ is a substrate for the growth) or $f^j_i = \frac{K_{i,j}^j}{K_{i,j}^j + C_i}$ (when nutrient $j$ is an inhibitor of growth; for details, see Fig. 3 and Table S7 in Sect. S4).

The proposed model describes the various roles of phototrophs (i.e. Cyanobacteria) within its growth dynamics by including the activity switch between photosynthesis and dark respiration, the regulation of the $C / N$ ratio via N$_2$ fixation by heterocysts, and the production of EPSs. By adapting their growth stoichiometry to the local environment, phototrophs in the model control the primary productivity of the entire system depending on the time of the day (photosynthesis, dark respiration), nutrient availability (unbalanced $C / N$ ratio), and hydration conditions (EPS production). A detailed description of the activity of phototrophs is provided in Sect. S3.

To evaluate the stoichiometries of heterotrophs and nitrifiers, microbial metabolic reactions are explicitly considered using the MbT-Tool (Metabolism based on Thermodynamics) (Araujo et al., 2016). Details on the stoichiometry for microbial growth in the model can be found in the Supplement: Table S5 in Sect. S4. A graphical summary of microbial growth and trophic interactions is given in Fig. 3.

### 2.4.2 Temperature-dependent microbial growth

Desert environments are often characterised by large diurnal temperature fluctuations (especially in hot deserts), which influence microbial activity. To consider these thermal effects, a temperature-dependent growth model using the Arrhenius equation is included in the model. Although temperature adaptation and growth adjustments may vary among microbial species, we opted for a simple representation in which all species are assumed to follow the same optimal temperature. The maximal growth rate for a cell at tempera-
2.4.3 pH feedback

Our model considers the spatial and temporal variations in pH values that could locally affect microbial activity. Unlike the narrow range of high pH regulating the activity of autotrophs (often limited by dissolved organic carbon), the activity of heterotrophs in the presence of dark respiration likely lowers pH when other substrates are not limited. Furthermore, nitrate accumulation can result in acidification of the soil domain when denitrification is absent. Considering that high acidity and alkalinity profoundly affects microbial growth through substrate binding and catalyse reactions, the feedback of microbial growth to local pH change is included in the model. The microbial feedback on biocrust pH can vary based on types of enzymes, the number of ionisable groups, and the organisms under consideration. In this work, a non-competitive inhibition model in the form of a Monod function is employed (Tan et al., 1998):

\[ f_T = \left[ \frac{T^S e^{-\frac{\Delta H^S}{R} \left( \frac{1}{T^S} - \frac{1}{T} \right)}}{1 + e^{-\frac{\Delta H^S}{R} \left( \frac{1}{T^S} - \frac{1}{T} \right)} + e^{-\frac{\Delta H^S}{R} \left( \frac{1}{T^S} - \frac{1}{T} \right)}} \right], \tag{7} \]

where \( T^S \) is a reference temperature (25 °C = 298 K) and \( \Delta H^S \) (cal mol\(^{-1}\)) is the activation enthalpy of the reaction. In this model, two inactivation regimes are considered: one of low temperature, denoted by \( L \), and another of high temperature, denoted by \( H \). The parameters included for enthalpies and inactivation regimes are given in Table S9 in Sect. S5.

2.4.4 Microbial growth rates

A key objective of our model is to determine the spatial organisation of a microbial community based on local gradients in conditions and resources. Several biocrust physico-chemical properties and environmental conditions determine the microbial growth rate following the diel cycles of light, temperature, and feedback of pH. As a result, the growth rate of individual cell \( i \), Eq. (17), is explicitly expressed as

\[ \mu_i(r,t) = \mu_{\text{max},i} f_T(r,t) f_{\text{pH}}(r,t) \cdot \min[f_{i1}^1(r,t), f_{i2}^2(r,t), \ldots]. \tag{9} \]

Here, substrates are described within their minimum function (mass limitation of electron donors and acceptors) unlike pH and temperature correction terms. We assume that \( f_T \) indicates the optimal temperature of enzymes and \( f_{\text{pH}} \) indicates the costs of the osmosis of protons; therefore, they act on the maximum growth rate directly.

2.5 Microbial EPS production

The importance of EPSs for microbial life in natural environments has been discussed in many review articles (Or et al., 2007; Flemming and Wingender, 2010; More et al., 2014). Especially in arid or semi-arid environments, the role of EPSs secreted by Cyanobacteria is crucial for microbial communities surviving within (and below) biocrusts (De Philippis and Vincenzini, 1998; Pereira et al., 2009; Mager and Thomas, 2011; Rossi et al., 2012; Colica et al., 2014; Rossi and De Philippis, 2015). The synthesis of EPSs contributes to the stability of soil structure and hydrated microenvironments in soil, making it a key ingredient of biocrust formation. EPSs also function as nutrient storage by immobilising nutrients (dust trapping or glycosidic bonds) and as a protective shield from adverse environments, such as UV radiation, antibiotic substances, and invading viruses. In this work, we focus on two key aspects of EPSs in biocrusts: modification of the diffusion process of substrates and its role as a nutrient reservoir (increase in soil C) (Or et al., 2007; Pereira et al., 2009; Mager and Thomas, 2011). The complete range of EPS effects on soil hydrology, such as the swelling of hydrated gel owing to its chemical composition and physical structure, are not considered in this study.

2.5.1 EPS production and transport properties

EPS production by Cyanobacteria in drylands varies with soil type, climatic conditions, hydration status, and other resources (Hu et al., 2002). Estimation of production rates and amounts remain challenging. It is generally accepted that EPS synthesis in cyanobacterial soil crusts is affected by changes in moisture availability and nitrogen level (Mager and Thomas, 2011). We thus coupled photosynthesis and \( \text{N}_2 \) fixation in the biocrust model. This approach allows for the computation of the net production of carbohydrates using dynamic stoichiometry. A certain proportion of carbohydrates produced is assumed to be transformed into EPSs depending on the local hydration conditions (for details, see Sect. S3.3).

The fraction of EPSs produced from photosynthetically fixed carbon is defined by the binding of extracellular carbohydrate residues to the polymeric matrix. The binding probability is written as a function of EPS concentration \( C_{\text{EPS}} \) and
the saturation degree of water $\Theta$ in the model:

$$f_p(C, \Theta) = \frac{1}{e^{\frac{C_{\text{EPS}}-C_{\text{EPS}}^*}{C_{\text{EPS}}^*}} + 1},$$

(10)

where $C_{\text{EPS}}^*$ is the gelation point for an EPS as a polymeric substance. The function describes how residual carbohydrates will not bind to polymeric substances as long as EPSs are in the form of a weak gel (reaching $C_{\text{EPS}}^*$). The degree of polymer binding is regulated by the saturation degree. For example, when the domain is wet, EPS hydrolysis will lower the binding probability of newly produced residual carbohydrates.

Many studies have suggested different physical models to describe the diffusion coefficient in EPSs (Masaro and Zhu, 1999). For our biocrust model, we adopted the simple diffusion model in gels proposed by (Phillies, 1987):

$$D = D_0 e^{-\alpha_d \nu^2},$$

(11)

where $\alpha_d$ and $\nu$ are scaling parameters that differ from substance to substance. It is shown that $\alpha_d$ depends on the diffusant’s molecular weight (in g L$^{-1}$) and $\nu \sim 0.5$ for a high-molecular-weight diffusant (macromolecules). Diffusion of carbohydrates and EPSs is governed by this equation in the model.

### 2.6 Diffusion reaction equation on the biocrust scale

Microbial activity and resource consumption are expressed as a set of diffusion reaction equations within the biocrust domain.

$$\frac{\partial C_j(r,t)}{\partial t} = \nabla \cdot \left( D_j(r,t) \nabla C_j(r,t) \right) - \frac{1}{V_w(r,t)} \sum_{i=1}^{N(r)} \mu_i(r,t)b_i(t) + S_j(r,t),$$

(12)

where $C_j(r,t)$ is the local concentration of substrate $j$, $D_j(r,t)$ is the local diffusion coefficient (including modification by EPSs), and $V_w(r,t)$ is the amount of water in a given patch at position $r$ and time $t$. The second term on the right-hand side is the reaction term to calculate the total substrate consumption and production in the patch. $N(r)$ is the total number of individual cells at $r$, $Y_{i,\text{net}}^j$ is the net yield of species $i$ on substrate $j$, $b_i(t)$ is the biomass, and $\mu_i(r,t)$ is the growth rate described in Eq. (20). The last term $S_j(r,t)$ is the source or sink term of substrate $j$ with respect to the mass transfer between the gas and liquid phases and the charge compensation from the principles of solution equilibrium and charge neutrality. These chemical processes are very fast compared to microbial reaction and diffusion processes. Thus, we implemented these terms as dynamic boundary conditions (keeping gaseous element solubility and local charge neutrality during one time step). For individual cells, the growth dynamics is written as

$$\frac{db_i(t)}{dt} = [\mu_i(r,t) - m_i]b_i(t),$$

(13)

where $\mu_i(r,t)$ is the growth rate from Eq. (20) and $m_i$ is the maintenance rate of cell $i$. Cell growth, division, locomotion, and death are described using individual-based modeling (Kim and Or, 2016; Kreft et al., 1998).

### 2.7 Evaluation of the proposed mechanistic desert biocrust model (DBM)

A pioneering study on microbial communities within desert biocrusts (Garcia-Pichel et al., 2003) has suggested a vertical stratification of microbial community members in which abundance (biomass) and composition (functional groups) vary with the depth of the biocrust. Observations by Garcia-Pichel et al. (2003) demonstrated the stratification as a result of vertical gradients in physico-chemical conditions such as light, oxygen, pH, and other nutrients. The vertical profiles of N$_2$ fixation and the potential NH$_4^+$ oxidation rates (Johnson et al., 2005) and chemical profiles (total ammonium, nitrate) of soil solutions within active biocrusts (Johnson et al., 2007), as well as the profiling of oxygen concentration after wetting (Abed et al., 2014), have been investigated as well. Recently, the effect of physical conditions on cyanobacterial activity was examined using X-ray microtomography (Raanan et al., 2016). These experimental data on microprofiles within biocrusts can be used for a comparison between the measurements and numerical simulations of chemical and biological components within saturated crusts. For the comparisons in a spatial context, the DBM quantifies the biological activity as a product of local growth rate, $\mu$, and biomass, $b_i$, of cells with a unit of $\mu_{\text{cell}} g_{\text{soil}}^{-1} h^{-1}$:

$$A_j(r,t) = \mu_i(r,t)b_i(r,t).$$

(14)

This activity measure is suitable to indicate the active pathways for the upregulation of functional genes (i.e. the spatial distribution of gene activity). From this activity distribution, it is possible to calculate the rates of microbial processes, such as carbon fixation, ammonia oxidation, and denitrification by simply multiplying the yields from the stoichiometry of each species.

In contrast to the generally dry state of biocrust, most of the detailed studies reported above were conducted using saturated biocrusts (a state that rarely occurs in the field). Data on unsaturated biocrusts are hindered due to the technical difficulty of using microsensors (Pedersen et al., 2015) and molecular analysis of microbial activity (Carini et al., 2016). Consequently, we are left with the undesired option of using detailed data from saturated biocrusts for model evaluation. The primary aim of this study is to establish confidence in the DBM for these rare conditions and extend the predictions to the more common case of unsaturated biocrusts.
The DBM was evaluated with respect to diurnal dynamics and the results are compared to experimental studies that measured certain traits (e.g. gaseous efflux), such as the following studies: Thomas et al. (2008); Rajeev et al. (2013); Darrouzet-Nardi et al. (2015); Weber et al. (2015). In this work, we focus on carbon dioxide efflux under fully saturated conditions (Rajeev et al., 2013). Although the gaseous fluxes are usually considered direct indicators of microbial activity, quantitatively speaking these macroscopic measures emerged from all possible biological, chemical, and physical interactions combined.

2.8 Physical domain and boundary conditions for nutrients

For a prescribed matric potential (constant hydration conditions), the corresponding water film thickness, aqueous habitat connectivity, diffusion properties, and specific surface area are obtained locally on a patch scale (about 100 µm) from the pre-assigned surface properties and local porosity of each patch (a spatial element that determines local patch property). We selected parameters that mimic the property of loamy sand \( \phi = 0.4, \Theta = 0.6, D = 2.65 \). By applying \( m \times n \) patches, the domain describes a thin strip of biocrust with periodic boundary conditions in the horizontal direction.

For boundary conditions of chemical substances, gaseous elements and dissolved elements are treated differently. Oxygen, carbon dioxide, ammonia, nitrous oxide, and nitrous acid in the gas phase are assigned based on the atmospheric composition from the literature (see Table S1 in Sect. S2). The mixing ratios of atmospheric components are kept constant at the top of the domain during simulations assuming zero diffusive boundary layer and maximised gas exchange between atmosphere and biocrusts. These gaseous compounds are transferred to the liquid phase by their own solubility based on Henry’s law (Sander, 1999). DIC, ammonia, and nitrous acid are partitioned with the principle of local charge neutrality at obtained pH values.

Model evaluation is based on the following components: we first present the steady-state distribution of geochemical variables within the biocrust domain. Next, we present the quasi-steady distribution of microbial functional groups within the biocrust under field capacity (relatively wet conditions). We then compare the model results for the saturated conditions under which sample experimental data are available.

3 Results

3.1 Steady state of geochemical traits within the biocrust (no biological activity)

The abiotic exchanges that affect local distributions of geochemical environments and traits are evaluated first. A steady state of a chemical domain is calculated in the absence of biological activity. We consider a biocrust following wetting at field capacity (corresponding to water saturation 0.6 for the entire domain) assuming that this condition describes wetted crusts after drainage (in contrast to a fully saturated crust with saturation degree 1). We focus on traits such as diffusion, gas–liquid partitioning, and acid–base calculation without microbial activity. The spatial variations in phase distributions within the simulation domain (vertical cross section of biocrust) and related attributes are depicted in Fig. 4. The results suggest that these relatively wet conditions may disrupt gas-phase connectivity to the atmosphere. Gas diffusion through the biocrust is determined by the connectedness of the gas phase according to percolation theory. For certain values of local gas content (below 0.2), the gas phase becomes disconnected, affecting O\(_2\) distribution. This implies that gas volumes not connected to the atmosphere may exist in isolated pockets within the soil domain. Thus, the local concentration of dissolved oxygen varies according to this atmospheric source and spatial heterogeneity (Fig. 4c). This also shows a correlation between the gas-phase configuration and the spatial heterogeneity of porewater pH; the higher the local gas content, the lower the pH values (activity of protons). This indicates that a higher mass transfer rate from the gas to the aqueous phase yields acidity since the dissolution of CO\(_2\) is very fast in unsaturated soils (large surface areas and thin water films). On the other hand, patches with high water content and limited gas-phase penetration show higher pH (around 8–9) as the model mimics alkaline soils with high cation content (about 10 µg g\(^{-1}\) calcium and the same amount of other non-reactive cations, as shown in Johnson et al., 2005). This implies that volume-averaged pH may not be representative of local soil porewater or water film pH in unsaturated soil, thereby affecting microbial activity locally and giving rise to processes not definable by average values.

3.2 Microbial activity effects on the biocrust chemical environment

The four microbial groups are introduced into the simulation domain (representing a cross section in desert biocrust) and allowed the system to stabilise under diurnal cycles. Phototrophs were initially inoculated in the domain in an exponentially decaying manner over the biocrust depth to reflect a natural organisation under light penetration, while other groups were inoculated uniformly in the domain. Only phototrophs were inoculated differently to reduce the computational time as phototrophs only thrive up to the depth at which light penetrates. This well-mixed inoculation pattern ensures that the spatial organisation of microbial populations within the crust was not affected by initial conditions. The initial population sizes were the same for all functional groups, about 4000 cells for the entire domain. After about five consecutive days (diurnal cycles), the total population and spatial distribution of microbial groups reached a quasi-steady state.
Figure 4. Aqueous-phase distribution affects diffusion pathways and geochemical conditions (no biological activity). A typical simulation result of a steady-state soil biocrust (up to 10 mm of depth) when biological activities are absent at standard ambient temperature ($T = 25^\circ$C). (a) A pre-assigned soil structure determines the local gas content and configuration of water at field capacity (the aqueous phase is complementary in these pore spaces). (b) The unsaturated soil permits the gas phase to penetrate over the biocrust depth along pathways (marked in blue) not blocked by the aqueous phase (marked in blue). The process is described by invasion percolation in this study. When the gas phase is connected to the atmosphere, partial pressures of gaseous compounds equilibrate to the atmospheric level as boundary conditions. Gas- and liquid-phase configurations determine the distribution of chemical species in the liquid phase. (c) The distribution of dissolved oxygen concentration and (d) localised soil porewater pH.

Figure 5. Diurnal distributions of chemical constituents in the desert biocrust. A typical result of simulated chemical profile within biocrusts at midday (top panel) and at midnight (bottom panel) at field capacity (wet but unsaturated). (a, e) The profile of dissolved oxygen is relatively stable during the day and night cycle. This implies that gas transport from the atmosphere is fast enough to override the consumption and production of the microbial community. (b, f) The profile of pH changes in contrast to that of oxygen. During the day, the top of the crust (within 2 mm) exhibits strong alkalisation, marked as blue in the figure. During the night, pH at the top goes back to a similar level below 2 mm. (c, g) Total ammonia nitrogen (TAN) increases during the day on the top of the crust due to microbial production ($N_2$ fixation) and decreases during the night through microbial consumption. (d, h) Nitrate distribution shows a tendency of cumulation below 4–5 mm without clear diurnal patterns.
Noticeable changes in the resulting chemical environments occurred due to microbial activities even though the physical environments and hydration conditions were assumed to be constant (held at relatively wet conditions corresponding to field capacity). Figure 5 depicts four spatially distributed chemical attributes, namely dissolved oxygen, pH, total ammonia nitrogen, and nitrate, for midday (top panels) and midnight (bottom panels). The chemical profiles delineate the diurnal cycles of microbial activity across the soil domain. For instance, the alkalisation of top crust (2 mm) was clearly shown together with the production of ammonium. This implies that phototrophic activity fixes inorganic carbon and produces ammonium to fix N₂ using heterocysts. However, the oxygen profile was relatively stable compared to other chemical substances although photosynthesis and dark respiration could introduce changes in the local concentration of dissolved oxygen. This is due to the unsaturated conditions on the top crust where gas transfer rates override the net reaction rate of oxygen within the profile. In addition, the nitrate profile exhibits the tendency of cumulation below 4–5 mm, implying that inhibited denitrification occurred under unsaturated conditions. The diurnal patterns of nitrate were not clear, unlike the profile under saturated biocrusts (see Fig. S4 in Sect. S6). In general, regardless of differences among various chemical species and diurnal cycles, the strong spatial heterogeneity was still significant within the domain shaped by gas–liquid configuration.

### 3.3 Vertical stratification of microbial functional groups

The dynamics of the biocrust chemical environments are not only due to general microbial activity, but specifically due to trophic interactions within the biocrust community (due to different substrate use by microbial groups). A typical simulation result of the DBM is given in Fig. 6 to represent the activities and interactions among biocrust microbiota under two distinctive phases: (1) during daytime with active photosynthesis (a–d) and (2) during nighttime with dark respiration (e–h). The results show the emergence of vertical stratification of each microbial process within the thin biocrust (10 mm). The biocrust community is highly active above 4 mm and only some aerobic activities appeared very sparse.
activities are segregated due to the strong alkalisation during photosynthesis and intense competition over ammonium with AOB (marked in dark blue). Weak activity of anaerobic bacteria is also found together with aerobes at a similar depth due to the need for organic carbon for their activity. Local anoxic conditions support their growth in certain regions (purple in Fig. 6a and c) due to the consumption of oxygen by other organisms, heterotrophs, and nitrifiers. Below 3 mm, anaerobic activity is not found because the oxygen consumption by aerobic organisms is too low to create local anoxic conditions. Chemoautotrophs appear sparse over the depth, and AOB and NOB (light blue) stay in proximity as they are in a mutualistic relation. AOB shows high activity within 2 mm during daytime, benefiting from the ammonium fixed by the heterocysts of phototrophs and the inorganic carbon produced by heterotrophs. Its growth is mainly limited by inorganic carbon used during photosynthesis. The activity of NOB is also high at the top crust due to nitrite production by AOB.

Generally during daytime, the activity of phototrophs enhances other microbial activity by fixing inorganic carbon and nitrogen (Fig. 6d). During nighttime, phototrophs switch their activity to dark respiration. Dark respiration by phototrophs drives intense competition for organic carbon and ammonium among individuals at the top of the domain. As the input of fixed carbon and nitrogen is absent, the depletion of ammonium at the top crust lowers the activity of most organisms (Fig. 6g). However, NOB shows slightly higher activity during the night below 3 mm, suggesting that during daytime they are outcompeted by other organisms owing to their high yield and low growth rate.

3.4 Fully saturated biocrusts: comparing model predictions with observations

Despite the focus of the desert biocrust model (DBM) on unsaturated conditions in desert systems, we had to rely on definitive experimental data from saturated biocrusts to evaluate the details of model performance (García-Pichel and Belnap, 1996; Johnson et al., 2007; Abed et al., 2013; Rajeev et al., 2013; Raanan et al., 2016). The simulation domain was saturated by simply applying near-zero matric potential and filling up all surface pores with water. Using the fully saturated domain with stable microbial community distribution, the model biocrust was then exposed to diurnal cycles of radiation and temperature.

The spatial distribution of microbial activity within a fully saturated biocrust is given in Fig. 7. Ten independent simulations were averaged to obtain the possible distribution of microbial processes. The potential activity of anaerobes peaks below 2 mm (in contrast to other aerobic organisms) due to the formation of an anoxic region (Fig. S4 in Sect. S6). At the top, microbial distribution is clearly stratified in the following order: phototrophs, nitrifiers, aerobic heterotrophs, denitrifiers. Unlike unsaturated biocrusts, the vertical stratifica-
The chemical environments of other substrates during daytime and nighttime are given in the Supplement: Fig. S4 in Sect. S6.

3.5 Diurnal cycles of gaseous efflux from saturated biocrusts

In addition to comparing processes within the crust (Figs. 7 and 8), we simulated gas efflux from the saturated biocrust and compared it with the measurements of Rajeev et al. (2013). Figure 9 depicts the efflux of three gas compounds of carbon and nitrogen, namely CO\(_2\), NH\(_3\), and N\(_2\)O. We represent uptake as negative gas efflux and positive for emissions. The diel cycles of CO\(_2\) efflux are plotted together with experimental data tracking the net carbon exchange between the biocrust and the atmosphere (Fig. 9a). Within the biocrust, carbon fixation and respiration occur simultaneously; the net CO\(_2\) efflux indicates a balance between
respiration (release) and photosynthesis (uptake). The simulation results are in qualitative agreement with experimental data, except the steep transitions after sunrise and gradual changes after sunset that are not captured properly. We attribute this to the simplified model (using Monod functions) of the onset of photosynthesis and dark respiration. Next we evaluate the daily patterns of ammonia volatilisation to represent nitrogen abiotic losses. The results in Fig. 9b show that ammonia volatilisation occurs mainly during daytime as the top of the biocrust turns alkaline (pH above 10). The total ammonia loss due to volatilisation was estimated to be about 500 nmol m$^{-2}$ day$^{-1}$, similar to reported values of 540–1000 nmol m$^{-2}$ day$^{-1}$ from intact biocrusts on the Colorado Plateau (Evans and Johansen, 1999; Barger et al., 2016). We then evaluate N$_2$O release from the biocrust (indicative of denitrification); the results in Fig. 9c show that immediately after wetting N$_2$O flux is high. We attribute this rapid release to the accumulation of NO$^+_3$ during unsaturated conditions. After 2 days, nitrate is exhausted and denitrification relies on the activity of NOB. Finally, we also considered the potential release of NO$_3^-$ from the soil solution in the form of nitrous acid HONO. However, the results show no such release in agreement with the observations of Weber et al. (2015), which are not presented in this paper.

4 Discussion

4.1 Spatial and temporal variations in local pH within unsaturated biocrusts

Soil pH has been recognised as a significant predictor of microbial community composition and diversity (Fierer and Jackson, 2006; Lauber et al., 2009). Furthermore, for alkaline or saline soils (typical desert soils), abiotic contributions to gaseous efflux may account for up to 40% of total CO$_2$ emissions (Ma et al., 2013). Thus, to separate biotic and abiotic contributions for gaseous efflux, reliable estimates of pH are needed. It is especially crucial when the main producer of the system, phototrophic microorganisms, depends on the accessibility of inorganic C and N. The proposed desert biocrust model (DBM) offers a distinct advantage in this respect, namely the localised (pore scale) representation of pH that integrates physico-chemical interactions and microbial activity. The simulated pH profile dynamics within wet biocrusts presented above (Fig. 8) have confirmed that
the activity of photoautotrophs alters local pH by depleting DIC during a diel cycle (consistent with observations).

The results of the DBM suggest strong spatial variations in local pH within the unsaturated biocrust although the overall (spatially averaged) soil pH indicates an alkaline soil (Fig. 4). In practice, however, the spatial distribution of local soil pH is difficult to measure and often requires the use of microelectrodes (Pedersen et al., 2015). Moreover, it has been argued that the use of microsensors is limited to near-saturated soils (McIntyre, 1966). The modelled spatial variations in local acidity are consistent with the uptake kinetics of nitrous acid in the gas phase on a wetted wall film (Hirokawa et al., 2008). The model results suggest that pH in thin water films may be lower than in bulk liquid due to the resistance of mass transfer from the gas to the bulk liquid phase (we use the term “bulk” to represent large water-filled pores within the biocrust). As liquid surface on the wall corresponded to acidity in thin water film in the model, this result may support model predictions and the importance of soil water configuration in shaping local pH within unsaturated soils.

The strong correlation between soil moisture retention and soil pH and their role in defining the microbial community structure (Lauber et al., 2009) might be attributed to local pH distribution in unsaturated soils. We speculate that the high abundance of Acidobacteria (at phylum level), known to grow well in acidic cultures (pH 3.5–6.5) as aerobic heterotrophs (Pankratov et al., 2008) in most soils (Jones et al., 2000), might offer more evidence of the importance of localised acidity in unsaturated soils. We note that such an acidity-related phylum was also found in biocrust communities (Steven et al., 2013).

### 4.2 Microbial community stratification within biocrusts

Spatial segregation along vertical gradients is a well-known feature of microbial communities in aquatic biofilms, microbial mats, and endolithic communities (Schramm et al., 2000; Paerl et al., 2000). Similar to the Winogradsky column, these microbial stratifications are driven by the distribution of electron acceptors and donors. Since the most favourable electron acceptor for aerobic organisms is oxygen, the low solubility of oxygen and the limited diffusion of dissolved oxygen play a pivotal role in the emergence of spatial stratification. Stratification within biocrusts is also observed in terms of the biomass of oxygenic phototrophs, aerobic heterotrophs (Garcia-Pichel et al., 2003), and community composition analysis based on 16S rRNA sequencing (Steven et al., 2013). The simulated results of our biocrust model agree with observations of vertical stratifications in the biocrust community (Figs. 6 and 7).

The DBM captures the key physico-chemical conditions essential for vertical stratifications. The steep gradient of oxygen on top of the fully saturated biocrust (Figs. 8 and S4) is caused by limited mass transfer from the atmosphere and rapid consumption of oxygen. During nighttime, the depletion of oxygen (below a few millimetres) is expected naturally because of the limited amount of oxygen input. The oxygen produced by phototrophs during daytime is immediately depleted by aerobic organisms in the domain. Clearly, such formation of an anoxic region within the crust benefits anaerobic activity at a few millimetres (Fig. 7). The creation of supersaturation closer to the surface also indicates slower diffusion than net production and consumption of oxygen. Experiments on biocrusts immersed in water indicated effervescing of (presumably) oxygen at the surface (Rajeev et al., 2013). This demonstrates that the net production of oxygen is higher than the diffusion of dissolved oxygen.

The vertical segregation of different microbial groups also indicates organic carbon and nitrogen diffusing from the phototrophs and becoming available to other microbial members, especially stratification among aerobic organisms. The dominance of nitrite oxidising bacteria (NOB) at the top 2 mm is largely due to ammonia volatilisation. The alkalisation of the top crust during daytime increases ammonia volatilisation, which is not beneficial for aerobic heterotrophs and ammonia oxidising bacteria (AOB). Therefore, their activity retreats to a few millimetres (Fig. 7). The creation of anaerobic activity at a few millimetres (Fig. 7). The creation of an anoxic region within the crust benefits anaerobic activity at a few millimetres (Fig. 7). The creation of an anoxic region within the crust benefits anaerobic activity at a few millimetres (Fig. 7). The creation of an anoxic region within the crust benefits anaerobic activity at a few millimetres (Fig. 7). Therefore, the aqueous-phase configuration within unsaturated biocrusts (also possibly extending to general unsaturated soils) shapes microbial activity unlike in aquatic microbial mats and similar saturated systems. This stable oxygen profile of unsaturated biocrusts is due to the mass transfer between gas and liquid, which is assumed to be very rapid in the model (instant equilibration by Henry’s law; see Sect. 2.3.2). However, in real biocrusts in natural fields, the exchange of gases with the atmosphere can be constrained even under unsaturated conditions (at a certain range) because of a dense layer of EPSs and a finer soil texture in the uppermost part within biocrusts. These factors can retard mass transfer by decreasing interfacial area under relatively wet conditions (finer soil texture) and by sustaining thick water films owing to the presence of EPSs. The current model allows us to assign a finer soil texture to the biocrust domain by using a low porosity or a high fractal dimension on the uppermost part. For the biocrusts loaded with a dense EPS layer, the model can be improved by relating the local EPS amount with the water film thickness at a given matric potential.
4.3 Complex trophic interactions of microbial community within biocrusts

The biocrust community exhibits highly dynamic and complex trophic interactions, such as commensalism surrounding organic carbon utilisation between phototrophs and heterotrophs, competition over nitrogen sources between aerobic heterotrophs and AOB, and cooperation between NOB and anaerobic denitrifiers. Temporally, the diel patterns of trophic interactions orchestrated by phototrophs drive the shift in activity distribution of microbial activity as has been shown from Namib Desert soil (Gunnigle et al., 2017). Spatially, these complex trophic interactions take place within thin biocrusts and yield emergent spatial distributions of microbial groups as depicted in Fig. 6. The remarkable concentration of such interactions within a few millimetres and the stratification of the activities of the various functional groups highlight the ecological sophistication and versatility of such fine-tuned desert ecosystems. Remarkably, opportunistic life forms are harboured within such biocrusts; for example, the presence of anaerobic heterotrophs at low numbers suggests the presence of local anoxic conditions even under mild unsaturated conditions (Ebrahimi and Or, 2015) and their rapid response to episodic wetting events (Št’ovíček et al., 2017).

4.4 Gaseous efflux from desert biocrusts

Motivated by the availability of definitive data, the DBM was applied to simulate diurnal changes in gas efflux from saturated biocrusts. The results were in good agreement with measured CO$_2$ efflux (Fig. 9). The model represents the diurnal cycles of other gas fluxes that may be sensitive to pH, such as ammonia volatilisation and HONO emission. Details of the geochemical environment shed light on the important role of local conditions (pH) in soil and biocrust microbial activity. For example, the activity of AOB in alkaline soils can be suppressed during daytime on the top crust as strong alkalisation leads to a loss of nitrogen compounds. On the other hand, NOB in acidic soils should experience the opposite; as the soil becomes more acidic, HONO emission would lead to nitrogen loss.

To realistically describe microbial life within unsaturated biocrusts or dry soils, the inclusion of gas-phase interactions is necessary. Most experiments on biocrusts were conducted under saturated conditions (presumably to induce a significant and measurable response); however, these responses occur during narrow climatic windows with high precipitation (Garcia-Pichel and Belnap, 1996, 2002; Johnson et al., 2007). Although we have shown gaseous efflux from saturated soils to compare with experimental results, the DBM is capable of quantifying gaseous efflux from unsaturated biocrusts by tracking gas and water distribution.

4.5 Assumptions and limitations of the desert biocrust model (DBM)

The proposed DBM makes numerous simplifications pertaining to the life and functions of a complex microbial community in biocrusts in arid and semi-arid regions. Regarding the key physical processes, we built a physical domain that contains small subregions represented as patches. A patch is a subsection within a small vertical cross section in the biocrust that represents soil surfaces with different properties that retain water films and transport nutrients and gas. This enables the consideration of spatial heterogeneity within a vertical 2-D cross section across a biocrust; however, lateral variations in biocrust properties in space are not considered here.

Key geochemical processes that are dominant in desert soils (and biocrusts) are considered in this model. For simplicity, we consider calcium as a buffer together with other non-diffusing background cations (assuming uniformly distributed non-reactive cations as a set point of pH). The effects of saline soil (also a common property of desert soils) on dissociation constants and its influence on soil pH are not considered. We also did not include the effect of EPSs (as organic matter) on the top of the biocrust. The role of EPSs as a gate for matter flux on desert soil surfaces, the interaction between pH alteration and microbial activity, and changes in the physical properties of soil (relation between EPS swelling ratio and pH) can be the next goals for a mechanistic model of biocrusts. Other important aspects regarding chemical processes include modifying the diffusion equation. In the current model, the possibility of electrokinetic flow is not included. A more detailed description of electromigration can be included by modifying the diffusion equation for ionic particles by using the Nernst Planck equation. However, as the input of carbon dioxide to the thin water film is faster than the aqueous diffusion of ionic particles, the occurrence of local pH variation owing to the configuration of the gas phase is still expected in unsaturated soil.

By far, the most simplified component in this model is the biological one related to microbial processes. The DBM represents a system containing an astonishing level of diversity with a small number of microbial functional groups. The interactions among these community members are regulated by simple stoichiometric relations that control microbial growth. Monod parameters are mostly taken from models for activated sludge (a system far removed from life in desert biocrusts) (Henze, 2000). Considering that a desert is a water-, carbon-, and nitrogen-limited system with abiotic stresses, the values of these parameters are likely to be different from those governing life in sludge systems. We note, however, that the proposed Monod growth parameters are affected by local environmental conditions, such as temperature, pH, and substrate concentrations. Yet, an understanding of half-saturation constants and ratios between growth rates among different microbial groups would be necessary for es-
establishing quantitative predictions by the DBM for real systems.

The members of the biocrust consortia were selected to focus on C and N cycling and the characteristics of arid environments. Recently, the role of heterotrophic diazotrophs, anaerobic ammonium oxidisers, and nitrate-reducing bacteria within biocrusts has been studied. Including these members might alter some of the expected rates that we presented in this study. Comparisons between crust models with their presence and absence can be one of the future applications. Furthermore, as the model describes a hydrated porous medium, the fully saturated domain is easily applicable to describe the microbial community of sediments or microbial mats. However, when it comes to modelling such systems, other groups, such as anaerobic phototrophs, sulfate- and iron-reducing bacteria, or methanogens, might need to be considered together with the proposed community of C / N cycling. This might be beneficial for a mechanistic understanding of the biogeochemistry of such systems.

The DBM can be further used to predict the gaseous efflux dynamics of wetting–drying cycles and C and N turnover rates during hydration events. As hydration events in arid and semi-arid areas are scarce, a mechanistic understanding of biocrust response to hydration would benefit estimations of its contribution to global biogeochemical cycles. For instance, high N loss via NO₃⁻ leaching, NH₃ volatilisation, and HONO emissions can be investigated with respect to N cycling in such an environment. Furthermore, short-term perturbations of hydration conditions on biocrusts can be another application of the model, such as short wet-up cycles or rapid evaporation at high temperatures. The physical roles of biocrusts in hydrological processes can also influence C and N cycling in arid areas. For instance, changes in infiltration properties and wind and water erosion are not considered in the current work on microbial communities in biocrust. However, on a larger scale, these physical changes in the domain can be further extended.

5 Summary and conclusions

In this study we develop a mechanistic model of a desert biocrust microbial community under the strong vertical resource gradients prevailing in surfaces of arid landscapes. The desert biocrust model (DBM) combines a detailed account of soil hydration for different soil properties, an individual-based description of microbial life, and chemical processes that affect the trophic interactions among the microbial groups as an ecologically functioning unit. Although simplified (as much as possible) it elucidates the role of soil structure in shaping gaseous–aqueous diffusion and substrate fluxes at the atmosphere–soil interface crucial for the microbial activity occurring therein.

The model results show the distribution and composition of microbial functional groups over vertical gradients of light, temperature, and substrates across a model biocrust. Furthermore, geochemical and physical processes of mass transfer at the gas–liquid interfacial area in soil matrix and kinetics for inorganic carbon and nitrogen fractionation underline the importance of modelling unsaturated soil that significantly deviates from other environments such as aquatic systems or saturated soils. The modified chemical environment displays the feedback of microbial activity from photosynthesis to CO₂ efflux from biocrusts. The local pH of soil water as a cumulative measure of local ionic species concentrations determines the availability of inorganic carbon and nitrogen or other minerals for microorganisms by controlling the solubility of chemical compounds and their degree of protonation. Although the model does not include individual differences in optimal pH for microbial activity, its results based on acid–base equilibrium predict the spatially and temporally organised activity of all functional groups. This self-organisation indicates one of the reasons why biocrusts can host a high abundance and diversity of microorganisms even under very harsh conditions like deserts. The DBM provides a means for a systematic and climatically driven evaluation of the critical role of microorganisms in desert ecosystems. The model offers predictive capabilities (within the limitations of the assumptions) for biocrust responses to climate change and their contribution to large-scale carbon and nitrogen cycles.

Code and data availability. All relevant simulation data are presented within the paper. Underlying data and MATLAB codes for the desert biocrust model can be obtained upon request from the corresponding author (minsu.kim@usys.ethz.ch).

The Supplement related to this article is available online at https://doi.org/10.5194/bg-14-5403-2017-supplement.

Author contributions. DO conceived the research. MK built the model and wrote the codes. MK and DO conducted the analyses and wrote the paper.

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

Special issue statement. This article is part of the special issue “Biological soil crusts and their role in biogeochemical processes and cycling”. It is not associated with a conference.
Acknowledgements. The authors thank Daniel Baumann for IT support. Minsu Kim thanks Samuel Bickel (ETHZ) and Iso Christl (ETHZ) for constructive comments on the model. This work was supported by a European Research Council (ERC) Advanced Grant (320499-SoilLife) and the SystemsX.ch (2013-158:MicroScapesX project).

Edited by: Anita Antoninka
Reviewed by: two anonymous referees

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M. Kim and D. Or: Modelling desert biological soil crusts

Biogeosciences, 14, 5403–5424, 2017


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