



**1 Modelling the diurnal and seasonal dynamics of soil CO₂ exchange in a semiarid ecosystem with
 2 high plant-interspace heterogeneity**

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11
 12 **Abstract**

13 This study represents a first attempt to model the diurnal and seasonal dynamics of soil CO₂ exchange
 14 (F_s) in a dryland ecosystem with a high plant-interspace heterogeneity. The modelling used an
 15 integrated process-based approach, in which the CO₂ production, transport and surface exchanges (e.g.
 16 biocrust photosynthesis, respiration and photodegradation) are considered simultaneously. The model
 17 was parameterized and validated with multivariate data measured during year 2013-2014 in a semiarid
 18 shrubland ecosystem in Yanchi, northwestern China. We also investigated the sensitivity of simulated
 19 F_s to a set of stand-specific parameters and investigated the relative contribution of different flux
 20 components. The model explained reasonably well the two-year dynamics of F_s measured from a non-
 21 crusted and two lichen-crusted plots. Simulations showed that the temporal pattern of F_s could deviate
 22 from that of the total CO₂ production from rooting-zone soil. Such deviations could be explained by
 23 the variations of CO₂ dissolution and the CO₂ exchanges of biocrust during wetting-drying cycles, and
 24 the root uptake and transport of dissolved CO₂. Moreover, the F_s was spatially sensitive to the plant-
 25 interspace differences and the variations in root biomass, soil organic matter and pH. These results
 26 emphasized that, the processes beyond autotrophic and heterotrophic respirations and the
 27 heterogeneities of soil at plant-interspace can strongly affect the F_s dynamics and their climatic
 28 sensitivities. Such variability should be carefully considered in extrapolation of findings from
 29 chamber to ecosystem level and from seasonal to inter-annual scales. Based on this work, our model
 30 can serve as a useful tool to simulate F_s dynamics in dryland ecosystems.

31
 32 **Keyword:** ecosystem modelling; heterogeneity; inorganic carbon; semiarid shrub ecosystem; biocrust



33

34 1. Introduction

35 CO₂ exchange between soil and atmosphere constitutes a major C loss from terrestrial ecosystems
36 (Raich et al., 2002; Giardina et al., 2014). It also plays an important role in the feedbacks between
37 global carbon cycle and climate change (Rustad et al., 2000; Giardina et al., 2014; Karhu et al., 2014).
38 However, the contribution of soil CO₂ flux (F_s) in arid and semiarid (dryland) ecosystems to the
39 global C budget is less-studied (Castillo-Monroy et al., 2011; Gao et al., 2012; Jia et al., 2014),
40 although these areas cover over 40% of land surface and contribute notably to inter-annual variations
41 of terrestrial C sink (Poulter et al., 2014). The temperature dependency of biological CO₂ productions
42 (i.e. autotrophic respiration and heterotrophic respiration) serves a conventional basis for F_s modelling
43 in many terrestrial ecosystems (Raich and Tufekciogul, 2000; Ryan, 2005; Song et al., 2015). Soil
44 CO₂ flux of dryland ecosystems is also widely interpreted using temperature-response functions
45 modified by other environmental constraints, e.g. soil water content, abundances of substrates and
46 microbial activities (Curiel Yuste et al., 2007; Wang et al., 2014a, 2014b, 2015).

47 Although empirical models may reproduce the dynamics of soil CO₂ flux in specified space-time,
48 their lack of mechanistic descriptions represents a major difficulty in extrapolation under changing
49 environmental conditions (Fan et al., 2015). Soil CO₂ flux is a “bulk” exchange that comprises two
50 main sets of processes, i.e. the CO₂ productions and transport (Fang and Moncrieff, 1999; Fan et al.,
51 2015). Hence, models considering only autotrophic and heterotrophic respiration often fail to account
52 for the observed F_s dynamics (Austin and Vivanco, 2006). Gas transport processes are important
53 mechanisms regulating the magnitude and hysteretic feature of soil CO₂ efflux (Ma et al., 2013). A
54 substantial fraction of respired CO₂ may be transported to atmosphere via xylem, and can’t be
55 measured by techniques like soil respiration chambers (Bloemen et al., 2013; 2016). During wet period,
56 soil CO₂ efflux could decrease significantly by water clogging of soil pores, which restricts the
57 diffusion of O₂ and CO₂ gases (Šimunek and Suarez, 1993; Fang and Moncrieff, 1999). In dryland
58 soils, the interactions between CO₂ transport and water cycle could also be intensive, due to the
59 commonly high salinity/alkalinity of soils. Large inorganic C fluxes can be driven solely by
60 dissolution and infiltration of CO₂ and carbonates (Buysse et al., 2013; Ma et al., 2013; Fa et al.,
61 2014). Such inorganic transport may not only contribute to the hourly or diurnal soil CO₂ efflux (e.g.
62 Emmerich, 2003; Xie et al., 2009; Buysse et al., 2013), but also to the terrestrial CO₂ sinks at much
63 broader spatiotemporal scales (Schlesinger, 2009; Li et al., 2015).

64 Key processes contributing to CO₂ production in dryland soils also extend beyond autotrophic
65 respiration and heterotrophic respiration. Although biocrust organisms (lichens, mosses, bacteria,
66 fungi and microfauna) inhabit in the top few centimetres of the soil profile, they constitute up to 70%
67 of biomes in the plant-interspace (Belnap, 2003). These communities are able to uptake C from the
68 atmosphere (Belnap, 2003; Castillo-Monroy et al., 2011; Maestre et al., 2013), leading to largely



69 greater concentration of organic matters in the crusted layer than the sub-soils (Ciais et al., 2013).
70 Although crust organisms could maintain inactive under stresses (e.g. drought, Green and Proctor,
71 2016), their photosynthetic potential could be large (Zaady et al., 2000; Lange, 2003), even
72 comparable to temperate forests with closed canopies (e.g. Zaady et al., 2000). The net C uptake by
73 biocrust is highly sensitive to stresses like droughts, thermal extremes and excessive ultraviolet
74 radiation (Pointing and Belnap, 2012). Such variations can readily alter the crusted soils between
75 considerable CO₂ sinks and sources within a few hours (e.g. Bowling et al., 2011; Feng et al., 2014).
76 In addition, the accumulation of debris from crust and canopy further fuels photodegradation, which
77 represents an important abiotic C loss in arid conditions beside the biotic decomposition (e.g. Austin
78 and Vivanco, 2006). Photodegradation is likely to dominate the mineralization during dry daytime
79 period, when the radiation is strong and microbial activities are prohibited by low moisture content
80 and high temperature (e.g. Gliksman et al., 2016). On an annual basis, photodegradation could
81 consume more than 10% of soil organic matter (SOM) at surface (e.g. Austin and Vivanco, 2006;
82 Henry et al., 2008; Brandt et al., 2010). This could be the case even for the substrates (e.g. lignin) that
83 are difficult to degrade via biotic pathways (Henry et al., 2008).

84 The influences of multiple C processes (i.e. autotrophic and heterotrophic respiration, net C uptake
85 by biocrust, inorganic C fluxes and photodegradation) on soil CO₂ exchange are highly overlapped
86 and tightly related to the water-energy processes. In dryland ecosystems, patchy vegetation and large
87 fractions of interspace are common features (Domingo et al., 2000), and the water-thermal conditions
88 can vary considerably from plant cover to interspace even within a few meters (Rodríguez-Iturbe et al.,
89 2001; Caylor et al., 2008; Ma et al., 2011). The water-energy dynamics at the different surfaces are
90 linked by multiple advection processes both above- and below-ground (Gong et al., 2016). Due to the
91 complexity of water-energy processes, there may exist possibly high non-linearity of water-thermal
92 responses to the climatic variability (e.g. Phillips et al., 2011; Barron-Gafford et al., 2013). This will
93 also complicate the C responses and consequently affect the relationships between the CO₂ fluxes and
94 environmental controls (e.g. Jarvis et al., 2007; Song et al., 2015).

95 The global change is expected to increase annual mean air temperature considerably and change
96 precipitation regimes (Donat et al., 2016). Understanding the response of dryland ecosystems to such
97 changes requires mechanistic models that integrate the multiple biotic and abiotic mechanisms in soil
98 C cycling. So far, only a few models have coupled the biotic CO₂ productions with the transport of
99 gas and heat (Šimuněk and Suarez, 1993; Fang and Moncrieff, 1999; Phillips et al., 2011; Ma et al.,
100 2013; Fan et al., 2015). Nevertheless, none of those models have described the heterogeneous water-
101 energy processes in soil-vegetation-atmosphere continuum (SPAC), or the unconventional C fluxes
102 such as net C uptake by biocrust and photodegradation despite the importance of these processes in
103 arid and semiarid environments. Models by Porada et al. (2013) and Kinast et al. (2016) represent the
104 few existing work in this sense. However, both models focus on the patterns at the regional-scale with



105 very simplified ecosystem processes and neglect stand-scale heterogeneities of water-energy budget,
 106 and have not yet been validated by field measurements.

107 This study represents a first attempt to model the diurnal and seasonal dynamics of soil CO₂
 108 exchange (F_s) in a dryland ecosystem with a high plant-interspace heterogeneity. The modelling used
 109 an integrative process-based approach, in which the CO₂ production, transport and surface exchanges
 110 (e.g. biocrust photosynthesis, respiration and photodegradation) are considered simultaneously. The
 111 model was parameterized and validated by multi-variant data measured during year 2013-2014 in a
 112 semiarid shrubland ecosystem in Yanchi, northwestern China. By employing the model, we also
 113 investigated the sensitivity of simulated F_s to a set of stand-specific parameters and investigated the
 114 role of different flux components in regulating the F_s . The model development in this study is based a
 115 water-energy modelling by Gong et al. (2016).

116

117 2. Materials and methods

118 2.1 Outlines for the modelling

119 The process-based modelling was based on multi-variate data measured during year 2013-2014 in
 120 a semiarid shrubland ecosystem located at the southern edge of the Mu Us desert (37°42'1" N,
 121 107°13'7" E, 1560 m above sea level), Ningxia, northwestern China (see Wang et al., 2014a, 2015).
 122 The long-term mean temperature (1954–2004) is 8.1 °C, and the mean annual precipitation is 287 mm,
 123 most of which falls from July to September (Jia et al., 2014). The radiation and evaporation demand
 124 are high in this area, i.e. the annual incoming shortwave radiation is 1.4×10^5 J cm⁻² and the annual
 125 potential evaporation is 2024 mm. The vegetation is dominated by scattered crowns of *Artemisia*
 126 *ordosica* (Fig. 1a). The soil is highly alkaline (pH = 8.2). Biocrust (mainly lichens and algae) covers
 127 about 40% of interspace soil. The thickness of the crust layer was 0.5 – 2.5 cm (Gong et al., 2016).

128 The modelled ecosystem was subtracted as replications of “representative land units” (RLU, Fig. 1;
 129 Gong et al., 2016), which consist of the area covered by shrubs and the surrounding soil (interspace).
 130 Vertically, the model simulates the C flows over the soil profile and the water-energy transport from
 131 the lower boundary of rooting zone to a reference height in the boundary atmosphere. Horizontally,
 132 the SPAC processes at plant cover and the surrounding interspace are differentiated but related via
 133 advection and diffusion flows, as driven by the gradients of temperature, water potential and gas
 134 concentrations. The mineralization, uptake and transport of soil C and N are further regulated by
 135 water-energy conditions. Key processes and variables included in the FS modelling are shown in Fig.
 136 1(c).

137 The model includes a set of sub-models, which describe: (i) CO₂ dissolution, transport and efflux;
 138 (ii) Autotrophic and heterotrophic CO₂ productions in the soil profile; (iii) CO₂ uptake and emission
 139 by biocrust; (iv) Surface energy balance and soil temperature profile; and (v) Soil hydrology and



140 water balance. These sub-models are linked by multiple feedbacks to represent the coupling of C,
 141 water, vapor and energy transportations in the ecosystem. Sub-models (iv) - (v) have been developed
 142 and described in details in our previous work (Gong et al. 2016), which focused on (i) introducing the
 143 plant-interspace heterogeneity into water-energy modelling, and (ii) investigating the influences of
 144 such heterogeneity on the ecosystem water-energy budgets for a dryland ecosystem. Gong et al. (2016)
 145 also validated the model in regard to the diurnal to seasonal dynamics of radiation balance, surface
 146 energy balance, soil temperature and moisture content in the footprint area of an eddy-covariance (EC)
 147 site (details of measurement see Jia et al., 2014). In this work, we therefore focus on the development
 148 of sub-models (i) – (iii) and their parameterization and validation by F_S measurements, based on
 149 automatic respiration chambers from crust-covered and non-crusted soils. Based on the validated
 150 model, we also analyzed the model sensitivities to stand parameters and plant-interspace heterogeneity
 151 and investigated the relative contribution of different flux components to F_S .

152

153 2.2 Modelling approaches

154 2.2.1 Submodel (i): CO₂ transport, dissolution and efflux

155 For soil fraction x (see Fig. 1b for RLU settings), CO₂ exchange (F_S , upward positive) was the sum
 156 of CO₂ exchange by biocrust (F_B), photodegradation (F_P) and the total emission from soil under the
 157 biocrust layer (F_T):

$$158 F_{Sx} = F_{Bx} + F_{Tx} + F_{Px} \quad (1)$$

159 where F_B is the net balance between biocrust photosynthesis (P_B) and respiration (R_B), and $F_B = P_B -$
 160 R_B (see Section 2.2.3). F_T was modelled based on the mass-balance functions of PATCIS (Fang and
 161 Moncrieff, 1999), which combined major transport processes in both gaseous and liquid phases. To
 162 account for the plant-interspace heterogeneity, we expanded the original one-dimensional function to
 163 the two-dimensional space. For soil layer (x, i) and time step t , the CO₂ concentration and C flows
 164 were calculated as follows:

$$165 \frac{\partial C_{x,i}}{\partial t} = \frac{\partial}{\partial z} (F_{dg}^v + F_{ag}^v + F_{dw}^v + F_{aw}^v) + \frac{\partial}{\partial h} (F_{dg}^h + F_{ag}^h + F_{dw}^h + F_{aw}^h) + S_{x,i} \quad (2)$$

166 where superscripts v and h denote the vertical and horizontal directions, respectively (see also in Gong
 167 et al., 2016); C is the total CO₂ content; F_{dg} and F_{dw} are the CO₂ flows due to diffusion/dispersion via
 168 the gaseous and liquid phases; F_{ag} and F_{aw} are the flows in gaseous and liquid phases due to gas
 169 convection and water movement, and S is the net CO₂ sink of the layer. The calculation schemes of
 170 F_{dg} , F_{dw} , F_{ag} and F_{aw} have been described in detail by Fang and Moncrieff (1999). F_T is the total
 171 exchange of gaseous CO₂ between surface and topmost layer:

$$172 F_{Tx} = F_{dg_{x,1}}^v + F_{ag_{x,1}}^v + E_{x,1}^S C_{w_{x,1}} \quad (3)$$



where $E_{x,i}^S$ is the soil evaporation at section x (see Eq. (17) in Gong et al., 2016); C_w is the equivalent CO_2 concentrations in the solution of the topmost soil. For layer (x, i) , C_w is linked to the gaseous CO_2 concentrations (C_g):

$$C_{x,i} = C_g V_{x,i} (V_{x,i} - \theta_{x,i}) + C_w \theta_{x,i} \quad (4)$$

where V is the total porosity; and θ is soil water content.

C_g and C_w were further related via the dissolution-dissociation balance of CO_2 in soil solution, following Fang and Moncrieff (1999) and Ma et al (2013):

$$\text{CO}_2(g) + \text{H}_2\text{O}(l) \rightleftharpoons \text{H}_2\text{O}(l) + \text{CO}_2(aq) \quad K_H = P_C / \text{CO}_2^{aq} \quad (5)$$

$$\text{CO}_2(aq) + \text{H}_2\text{O}(l) \rightleftharpoons \text{H}_2\text{CO}_3 \quad K_0 = \text{CO}_2^{aq} / [\text{H}_2\text{CO}_3] \quad (6)$$

$$\text{H}_2\text{CO}_3 \rightleftharpoons [\text{H}^+] + [\text{HCO}_3^-] \quad K_1 = [\text{H}^+][\text{HCO}_3^-] / [\text{H}_2\text{CO}_3] \quad (7)$$

$$\text{HCO}_3^- \rightleftharpoons [\text{H}^+] + [\text{CO}_3^{2-}] \quad K_2 = [\text{H}^+][\text{CO}_3^{2-}] / [\text{HCO}_3^-] \quad (8)$$

where P_C is the partial pressure of CO_2 in pore air; K_H is Henry's Law constant; K_0 , K_1 and K_2 are equilibrium coefficients of dissolution, the first- and the second-order dissociation reaction for carbonic acid, respectively (for details see Fang and Moncrieff, 1999). The equilibrium $[\text{H}^+]$ was determined by the soil pH and the coefficients K_H , K_0 , K_1 and K_2 , which were functions of soil temperature in each soil layer (Fang and Moncrieff, 1999). C_w was calculated as the sum of CO_2^{aq} , H_2CO_3 , HCO_3^- and CO_3^{2-} .

190

2.2.2 Submodel (ii): autotrophic and heterotrophic CO_2 productions along the soil profile

For soil layer (x, i) , $S_{x,i}$ (Eq. 2) was calculated as the sum of autotrophic and heterotrophic CO_2 productions, and the dissolved CO_2 removed with the water uptaken by roots:

$$S_{x,i} = R_{s,x,i} + R_{a,x,i} - E_{x,i} C_w \quad (9)$$

where E is the transpirative uptake of water (Gong et al., 2016); R_s is the CO_2 production by heterotrophic SOM decomposition; R_a is the autotrophic respiration of the rhizosphere, which comprises maintenance respiration (R_m) and growth respiration (R_g):

$$R_{a,x,i} = R_{m,x,i} + R_{g,x,i} \quad (10)$$

To simulate R_s , we simplified the pool-type model of Gong et al (2013, 2014), which was originated from Smith et al (2010) for simulating coupled C and N cycling in organic soils. SOM pool in each soil layer was divided into debris (M_{deb} , i.e. litters from roots and biocrust), microbes (M_{mic}) and humus (M_{hum}), which are different in biochemical recalcitrance and N content. During decaying, mineralized masses transfer from M_{deb} and M_{mic} to more resistant form (i.e. M_{hum}), leading to a decrease in lability (e.g. Li et al., 1992). The mineralization of organic C followed first-order kinetics



and was constrained by multiple environmental multipliers, including temperature, water content and oxygen content (Šimunek and Suarez, 1993; Fang and Moncrieff, 1999):

$$m_{x,i}^r = M_{x,i}^r k_r f(Ts_{x,i}) f(\theta_{x,i}) f(O_{x,i}) dt \quad (11)$$

where superscript r denotes the type of SOM pool ($r=1$ for M_{deb} , $r=2$ for M_{mic} , and $r=3$ for M_{hum}); m is mineralized SOM during time step dt ; k is the decomposition constant; dt is time step; $f(Ts_{x,i})$, $f(\theta_{x,i})$ and $f(O_{x,i})$ are multiplier terms regarding the temperature, water content and oxygen restrictions, respectively. $f(O_{x,i})$ was calculated following Šimunek and Suarez (1993). $f(Ts_{x,i})$ and $f(\theta_{x,i})$ were reparameterized with respect to the site-specific conditions of plants and soil (see Section 2.3.4). The CO_2 production from mineralization was further regulated by the N-starvation of microbes following Smith et al. (2010):

$$Rs_{x,i} = r_E m_{x,i}^r \quad (12)$$

where r_E is the gas production rate ($r_E \in [0, 1]$), and $(1 - r_E)$ is the proportion of organic matters passed to the downstream SOM pools. The evolution of each SOM pool was calculated as below:

$$M_{x,i}^r = (1 - r_E) m_{x,i}^{r-1} - m_{x,i}^r + A_{x,i}^r dt \quad (13)$$

where A is the SOM input rate ($A=0$ for M_{mic} and M_{hum}); superscript $r-1$ denotes the source SOM pools.

$Rm_{x,i}$ was calculated in a similar way to $Rs_{x,i}$ (e.g. Chen et al., 1999; Fang and Moncrieff, 1999). $Rg_{x,i}$ was calculated as a fraction of photosynthetic assimilates, following Chen et al. (1999):

$$Rm_{x,i} = M_{x,i}^R k_R f(Ts_{x,i}) f(\theta_{x,i}) f(O_{x,i}) dt \quad (14)$$

$$Rg_{x,i} = k_g P_g f r_{x,i} \quad (15)$$

where M^R is the root biomass; k_R is the specific respiration rate of roots; k_g is the fraction of photosynthetic assimilate consumed by growth respiration; $f r_{x,i}$ is the mass fraction of roots in soil layer (x, i). P_g is the photosynthesis rate of plants. P_g was estimated using a modified Farquhar's leaf biochemical model (see Chen et al., 1999). This model simulates photosynthesis based on biochemical parameters (i.e., the maximum carboxylation velocity, V_{max} , and maximum rate of electron transport, J_{max}), foliage temperature (T_c) and stomatal conductance (g_s). The values of V_{max} and J_{max} were obtained from in situ measurements from the site (Jia et al., unpublished). T_c and g_s were given in the energy balance sub-model, which was detailed in Gong et al. (2016).

N content bonded in SOM mineralized and was added to soil layers simultaneously with decaying. The abundance of mineral N (i.e. NH_4^+ and NO_3^-) regulates the growth of microbial biomass and r_E following Smith et al. (2010) and Gong et al. (2014). Key processes governing the dynamics of mineral N pools included nitrification-denitrification (Smith et al., 2010), solvent transport with water flows (Gong et al., 2014) and the N uptake by root system. However, the plant growth was not modelled in this work and therefore, N_{upt} was calculated using the steady-state model of Yanai (1994),



238 based on the transpiration rate, surface area of fine roots and the diffusion of solvents from pore space
239 to root surface:

$$240 \quad N_{upt} = 2\pi r_o L \alpha C_o dt \quad (16)$$

241 where r_o is the fine root diameter; L is the root length, and $2\pi r_o L$ is the surface area of fine roots; α is
242 the nutrient absorbing power, which denotes the saturation degree of solute uptake system ($\alpha \in [0, 1]$);
243 C_o is the concentration of solvents at the root surface, and is a function of bulk concentration of
244 mineral N (N_{min}), inward radial velocity of water at the root surface ($v_o = E/(2\pi r_o L)$) and saturation
245 absorbing power α . Further details for calculations of α and C_o can be found in work of Yanai (1994).

246

247 2.2.3 Submodel (iii): CO₂ exchange of biocrust and photodegradation

248 Biocrusts are vertically layered systems that comprise topcrust (or, bio-rich layer) and underlying
249 subcrust (or, bio-poor layer), which are different in microstructure, microbial communities and C
250 functioning (Garcia-Pichel et al., 2016; Raanan et al., 2016). Topcrust is usually few-millimetre thick,
251 which allows the penetration of light and the development of photosynthetic microbes (Garcia-Pichel
252 et al., 2016). On the other hand, the subcrust has little photosynthetic-activity. We here focused
253 mainly on describing the C exchanges in the topcrust, but assumed the C processes in the subcrust
254 were similar to those in the underneath soil. We developed the following functions to describe the C
255 fixation and mass balance in the topcrust,

$$256 \quad F_{Ct} = P_{Ct} - R_{Ct} \quad (17)$$

257 where P_C is the bulk photosynthesis rate; and R_C is the bulk respiration rate. P_C and R_C were further
258 modelled as follows:

$$259 \quad P_{Ct} = \frac{\alpha_C A_{PAR} P_{Cm}}{\alpha_C A_{PAR} + P_{Cm}} \quad (18)$$

$$260 \quad R_{Ct} = M_{Ct} k_{cr} f_{RC}(T_{Ct}) f_{RC}(\theta_{Ct}) \quad (19)$$

261 where α_C is the apparent quantum yield, P_{Cm} is the maximal rate of photosynthesis, and was a function
262 of the moisture content (θ_{Ct}) and temperature (T_{Ct}) in topcrust; A_{PAR} is the photosynthetically active
263 radiation (PAR); M_{Ct} is the total C in the SOM of topcrust; k_{cr} is the respiration coefficient; $f(\theta_{Ct})$ and
264 $f(T_{Ct})$ are water and temperature multipliers. Here, we assumed no photosynthesis in subcrust. The
265 heterotrophic respiration (R_{Cs}) was calculated as was done for soil respiration (Eq. (11)) based on the
266 C storages ($M'_{x,I}$) and temperature and moisture content of crust layer (i.e. $T_{S_{x,I}}$ and $\theta_{x,I}$; see Eq. (29)
267 and Eq. (14) in Gong et al., 2016).

268 To consider different C losses and exchanges, and to calculate the C balance in topcrust and
269 subcrust, respectively, we considered the following matters. R_{Ct} includes the respirations from both



autotrophic (M_{CA}) and heterotrophic (M_{CH}) pools. When autotrophic organisms die, SOMs pass from M_{CA} to M_{CH} and influence the turnover processes. A variety of topcrust organisms can reach into subcrust (e.g. through rhizines, Aguilar et al., 2009) and export litters there. When the surface is gradually covered by deposits, topcrust organisms tend to move upward and recolonize at the new surface (e.g. Garcia-Pichel and Pringault, 2001; Jia et al., 2008), leaving old materials buried into the subcrust (Felde et al., 2014). On the other hand, the debris left to soil surface are exposed to photodegradation. Based on above, the C balance in topcrust and subcrust was calculated as following, assuming the partitioning of respiration between autotrophic and heterotrophic pools was proportional to their fractions:

$$M_{Ct} = M_{CA} + M_{CH} \quad (20)$$

$$\frac{dM_{CA}}{dt} = P_{Ct} - R_{Ct} \frac{M_{CA}}{M_{Ct}} - k_m M_{CA} - k_b M_{CA} \quad (21)$$

$$\frac{dM_{CH}}{dt} = k_m M_{CA} - R_{Ct} \frac{M_{CH}}{M_{Ct}} - k_b M_{CH} - F_P \quad (22)$$

$$\frac{dM_{Cs}}{dt} = k_b M_{Ct} - R_{Cs} \quad (23)$$

where k_m is the rate of C transfer (e.g. mortality) from autotrophic pool to heterotrophic pool; k_b is the rate of C transfer (e.g. burying) from topcrust to subcrust; F_P is the loss of SOM due to photodegradation.

Photodegradation tends to decrease surface litter masses in a near linear fashion with the time of exposure (Austin and Vivanco 2006; Vanderbilt et al., 2008). Considering the diurnal and seasonal variations of radiation, F_P was calculated as a function of surface SOM mass and solar radiation:

$$F_{Px} = M_{surf} k_p Rad_x \quad (24)$$

where Rad_x is the incident shortwave radiation at surface x (Gong et al., 2016); M_{surf} is the surface litter mass; and k_p is the photodegradation coefficient.

2.3 Model parameterization

2.3.1 Measurements of micrometeorology and soil CO₂ efflux

Meteorological variables were measured every 10 seconds and aggregated to half-hourly resolution during 2013-2014. The factors measured included the incoming and outgoing irradiances (PAR-LITE, Kipp and Zonen, the Netherlands), PAR (PAR-LITE, Kipp and Zonen, the Netherlands), air temperature and relative humidity (HMP155A, Vaisala, Finland). Rainfall was measured with a tipping bucket rain gauge (TE525WS, Campbell Scientific Inc., USA) mounted at a nearby site (1 km



away, see Wang et al., 2014a). The seasonal trends of the measured T_a and P can be found in Jia et al. (2016). No surface runoffs were observed at the site, indicating the horizontal redistribution of rainfall was mainly through subsurface flows.

Continuous measurements of F_s were conducted using an automated soil respiration system (model LI-8100A fitted with a LI-8150 multiplexer, LI-COR, Nebraska, USA). The system was on a fixed sand dune of typical size (Wang et al., 2014a), which was located about 1.5 km south from the EC tower described in Gong et al (2016). Three collars (20.3 cm in diameter and 10 cm in height, of which 7 cm inserted into the soil) were installed on average at 3m spacing in March 2012. One collar (C1) was set on a bare soil microsite with no presence of biocrust. Two other chambers (C2 and C3) were set on lichen-crusts soils. F_s was measured hourly from C1 and C2 by opaque chambers, whereas by transparent chamber from C3 to include the photosynthesis and photodegradation. Litters from the shrub canopies were cleared from the collars during weekly maintenance. Hourly T_s and θ at 10 cm depth were measured outside each chamber using the 8150–203 soil temperature sensor and ECH2O soil moisture sensor (LI-COR, Nebraska, USA), respectively. Root biomass was sampled near each collar (within 0.5 m) in July 2012, using a soil corer (5 cm in diameter) to a depth of 25 cm. The samples were mixed and sieved sequentially through 1, 0.5 and 0.25 mm meshes, and the living roots were picked by hands. The comparison of the three micro-sites is shown in Table 1. Methods used in data processing and quality control have been described earlier in details (see Wang et al., 2014a, 2015). The quality control led to gaps of 10 - 13% in the F_s dataset.

319

2.3.2 Parameterization of vegetation and soil texture

The parameterization schemes supporting the simulations of energy balance and soil hydrology in sub-model (i) - (v) have been described previously in detail by Gong et al. (2016). As the water-energy budget is sensitive to vegetation (i.e. canopy size, density and leaf area) and soil hydraulic properties (see Gong et al., 2016), we hereby revalued these parameters for the F_s site. Measurements based on four 5m×5m plots showed that the crown diameter D (86 ± 40 cm) and height H (47 ± 20 cm) at this site were similar to those measured from the eddy-covariance (EC) footprint by Gong et al. (2016). However, the shrub density was 50% greater, leading to higher shrub coverage (42%), shorter spacing distance L (40.2 cm) and greater foliage area. On the other hand, the subsoil at the F_s site is sandy and much coarser than that at the EC footprint. Therefore, we collected 12 soil cores from 10 cm depth, and measured saturated water content (θ_{sat}), bulk density and residual water content (θ_r) from each sample. Then, the samples were saturated, and covered and drained by gravity. We measured the water content after 2-hour and 24-hour draining, which roughly represented the matrix capillary water content (10 kPa) and field capacity (33 kPa) (Armer, 2011). The shape parameters n and α_n (see Eq. (26) in Gong et al. 2016) for the water-retention function were estimated from these values (Table 2).



336

337 2.3.3 Parameterization of soil C and N pools

338 The sizes and quality of soil C pools were parameterized based on a set of previous studies. The
 339 total SOC in the root-zone soil (i.e. 60cm depth, bulk density of 1.6 g cm^{-3}) was set to 1200 g m^{-2} ,
 340 based on the values reported from previous studies in Yanchi area (e.g. Qi et al., 2002; Chen and
 341 Duan, 2009; Zhang and Hou, 2012; Liu et al., 2015; Lai et al., 2015). The mass fraction of resistant
 342 SOM pool (M_{hum}) was set to 40 - 50 % of total SOM, following work by Lai et al. (2015). The vertical
 343 distribution of the SOM pools was described following Shi et al. (2013). At the ecosystem level, the
 344 total root biomass was calculated as proportional to the aboveground biomass (Xiao et al., 2005),
 345 which linearly related to the crown projection area ($0.5\pi D^2$, Zhang et al., 2008). The vertical profile of
 346 root biomass was parameterized following Li and Xiao (2007), and the root biomass was set to
 347 decrease with the distance from the center of a shrub crown (Zhang et al., 2008). The N content was
 348 parameterized following the measurement of Wang et al. (2015).

349 Based on the above settings, the specific decomposition rate of debris was estimated from the
 350 litterbag experiment of Lai et al. (2015), which showed a 16% decrease in the mass of fine-root litters
 351 during a 7-month period of year 2013 at the Yanchi site. The photodegradation coefficient (k_p) was
 352 calculated from the mass-loss rate reported by Austin and Vivanco (2006). M_{surf} was set to 33% of
 353 M_{CH} in topcrust, assuming the depth of light penetration was about 2 mm and C concentration was
 354 homogeneous in topcrust. The surface litter from canopy was not considered in this modelling, as the
 355 plant litters were cleaned from the collars during weekly maintenance. The specific respiration rate of
 356 roots (k_R), however, could be much greater during vegetative growing stage than other periods, e.g. at
 357 the defoliation stage (Fu et al., 2002; Wang et al., 2015). Here we linked k_R to the development of
 358 foliage in modelling using the approach of Curiel Yuste et al. (2004):

$$359 \quad k_R = k_{R0}(1 + n_R L_l / L_{max}) \quad (25)$$

360 where k_{R0} is the “base” respiration rate (Table 2); L_l is the green leaf area, which is a function of Julian
 361 day (Gong et al., 2016); L_{max} is the maximum L_l ; n_R is the maximum percentage of variability and is
 362 set to 100%.

363

364 2.3.4 Parameterization of the water-thermal sensitivity of soil CO₂ productions

365 Based on the empirical study of Wang et al. (2014a), the steady-state sensitivity of CO₂ productions
 366 to soil temperature and water content (i.e., $f(Ts)f(\theta)$, Eq. (11)) can be described as a logistic-power
 367 function:

$$368 \quad f(Ts)f(\theta) = f(Ts, \theta) = \{1 + \exp[a(b - Ts)]\}^{-1}(\theta/\theta_{sat})^c \quad (26)$$



where a , b and c are empirical parameters. This function represents the long-term water-thermal sensitivity of CO₂ productions over the growing seasonal, yielding an apparent temperature sensitivity Q_{10} of 1.5 for the emitted CO₂ (Wang et al. 2014a). However, this could underestimate the short-term sensitivities of CO₂ productions. The apparent Q_{10} could be much greater at the diurnal level than at the seasonal level (Wang et al., 2014a). In this work, we firstly calculated the “base” sensitivity using the long-term scheme (Eq. 26) with 1-day moving average of water-thermal conditions. Then the deviation of hourly sensitivity from “base” condition was adjusted by the short-term Q_{10} :

$$f(Ts)f(\theta) = f(Ts_{short}, \theta_{short}) + [f(Ts, \theta) - f(Ts_{short}, \theta_{short})]Q_{10}^{(Ts - Ts_{short})/10} \quad (27)$$

$$Q_{10} = \max[Q_{10}(Ts_{short}), Q_{10}(Ts_{short})] \quad (28)$$

$$Q_{10}(Ts_{short}) = -0.42 Ts_{short} + 12.4 \quad (29)$$

$$Q_{10}(Ts_{short}) = 18010 \theta_{short}^{3.721} + 1.604 \quad (30)$$

where Ts_{short} and θ_{short} are the 1-day moving averages of Ts and θ , respectively; $Q_{10}(Ts)$ and $Q_{10}(\theta)$ are the adjustment functions for short-term apparent Q_{10} , regarding the short-term Ts and θ .

Further non-linearity of soil respiration responses refers to the rain-pulse effect (or the “Birch effect”, Jarvis et al. 2007), that respiration pulses triggered by rewetting can be orders-of-magnitude greater than the value before rain event (Xu et al., 2004; Sponseller, 2007; Cable et al., 2013). Such response could be very rapid (e.g. within 1 hour to 1 day, Rey et al. 2005) and sensitive to even minor rainfalls. It also seems that the size and duration of a respiration pulse not only depend on the precipitation size, but also on the moisture conditions prior to the rainfall (Xu et al., 2004; Rey et al., 2005; Evans and Wallenstein, 2011). In this work, we multiplied a simple rain-pulse coefficient (f_{pulse}) to Eq. (26):

$$f_{pulse} = \max[1, (\theta/\theta_{72h})^{n_p}] \quad (31)$$

where θ is the 3-day moving average of soil moisture content; n_p is a shape parameter and was set to 2 in this study. θ_{72h} is the 72-hour moving average of θ .

393

2.3.5 Parameterization of biocrust photosynthesis and respiration

In sub-model (iii), Equations (17) - (19) were parameterized based on the experiment of Feng et al. (2014). In the experiment, 50 lichen (topcrust) samples of 0.5-0.7 cm thickness (100% coverage, average C content of 1048 $\mu\text{mol C cm}^{-3}$) were collected from a 20 m \times 20 m area. The samples were wetted and incubated under controlled T_C (i.e. 35°C, 27°C, 20°C, 15°C, and 10°C). These samples were divided into two groups to measure the net primary productivity (NPP) and dark respiration (Rd) separately. Gas exchanges and light response curve for each crust sample were measured using LI-6400 infrared gas analyzer equipped with an LI-6400-17 chamber and an LI-6400-18 light source (LI-



402 COR, Lincoln, NE, USA). Measurements were taken at ambient CO₂ values of 385 ± 35 ppm.
 403 Saturated topcrust samples were placed in a round tray and moved to the chamber. CO₂ exchange was
 404 measured during the drying of samples, until the CO₂ flux diminished. During drying, θ_{Ct} was
 405 measured every 20 min. For more details see Feng et al. (2014).

406 Fitting the measured Rd to T_{Ct} and θ_{Ct} (see Fig. 2a) obtained the multipliers in Eq. (19) as following:

$$407 \quad f_{RC}(T_{Ct})f_{RC}(\theta_{Ct}) = Q_{Ct}^{\frac{(T_{Ct}-20)}{10}}(a_{RC} + b_{RC}\theta_{Ct} + c_{RC}\theta_{Ct}^2) \quad (32)$$

408 where Q_{Ct} , a_{RC} , b_{RC} , c_{RC} are the fitted shape parameters (Table 2).

409 The parameterized Eq. (19) was then used to simulate the Rd for the NP samples, based on the
 410 correspondent T_{Ct} and θ_{Ct} of each measurement. P_{Cm} was determined by subtracting the simulated
 411 respiration rate from the NP measured under light-saturated conditions. Then P_{Cm} was fitted to T_{Ct}
 412 and θ_{Ct} as following (Fig. 2b):

$$413 \quad P_{Cm} = f_{Pt}(T_{Ct})f_{Pw}(\theta_{Ct}) \\ 414 \quad = (a_{Pt} + b_{Pt}T_{Ct} + c_{Pt}T_{Ct}^2 + d_{Pt}T_{Ct}^3)(-a_{Pw} + b_{Pw}\theta_{Ct} - c_{Pw}\theta_{Ct}^2 + d_{Pw}\theta_{Ct}^3) \quad (33)$$

415 where a_{Pt} , b_{Pt} , c_{Pt} , d_{Pt} , a_{Pw} , b_{Pw} , c_{Pw} , d_{Pw} are fitted shape parameters (Table 2).

416 It should be addressed that T_{Ct} and θ_{Ct} could change more rapidly than the mean conditions of the
 417 crust (i.e. $T_{Sx,l}$ and $\theta_{x,l}$). In this work, T_{Ct} was calculated from the surface temperature (T_s , see Eq. (13)
 418 in Gong et al., 2016) and $T_{Sx,l}$ by linear interpolation. The calculation of θ_{Ct} , on the other hand,
 419 depended on the drying-rewetting cycle. During drying phases, θ_{Ct} was interpolated linearly from $\theta_{x,l}$
 420 and surface moisture content (θ_x); whereas during wetting phases, the mass balance of water input P
 421 and evaporation loss ($E_{x,l}^s$, see Eq. (17) in Gong et al., 2016) was considered:

$$422 \quad T_{Ct} = \frac{T_s Z_{Ct} + T_{Sx,l} Z_{Sx,l}}{Z_{Ct} + Z_{Sx,l}} \quad (34)$$

$$423 \quad \theta_{Ct} = \max \left[\frac{\theta_x Z_{Ct} + \theta_{x,l} Z_{Sx,l}}{Z_{Ct} + Z_{Sx,l}}, \theta_{Ct} + \frac{P - E_{x,l}^s}{Z_{Ct}} \right] \quad (35)$$

424 where $Z_{Sx,l}$ is the thickness of the biocrust; and Z_{Ct} is the thickness of the topcrust. θ_x was calculated
 425 from the surface humidity and the water retention of the crust layer, using Eq. (25) – (26) by Gong et
 426 al. (2016).

427

428 2.3.6 Calculation of litter input to soil and SOC transport in biocrust

429 The litter falls added to each soil layer ($A_{x,i}^1$, Eq. (13)) were linked to the mortality of roots, which
 430 was calculated following Asaeda and Karunaratne (2000).



$$A_{x,i}^1 = k_{mo} Q_{mo}^{T_{S_{x,i}}-20} M_{x,i}^R \quad (36)$$

where k_{mo} is the optimal mortality rate at 20°C; Q_{mo} is the temperature sensitivity parameter (Asaeda and Karunaratne, 2000). Similarly, we attributed the C transport rate (A_{Cm}) from M_{CA} to M_{CH} mainly to the mortality of autotrophic organisms. We assumed that most mortality of crust organisms occurred during abrupt changes in wetness, as microbial communities may adapt slow moisture changes or remain inactive during drought (e.g. Roberson and Firestone, 1992; Reed et al., 2012; Coe et al., 2012; Garcia-Pichel et al., 2013; Maestre et al., 2013). Here, we introduced a water-content multiplier, $f_m(\theta_{Ct})$, to describe the impact of abrupt θ_{Ct} changes on k_m :

$$A_{Cm} = k_{mc} Q_{mo}^{T_{Ct}-20} f_m(\theta_{Ct}) M_{CA} \quad (37)$$

$$f_m(\theta_{Ct}) = \max[0.01, 1 - \min(\theta_{Ct}, \theta_{Ct7}) / \max(\theta_{Ct}, \theta_{Ct7})] \quad (38)$$

where k_{mc} is the optimal mortality rate at 20°C; Q_{mo} is the temperature sensitivity parameter (Asaeda and Karunaratne, 2000); θ_{Ct7} is the forward 7-day moving average of θ_{Ct} .

C transport from topcrust to subcrust was calculated as driven mainly by the sand deposition and burying of topcrust SOM. Assuming the C content in topcrust was homogeneous and the thickness Z_{Ct} was near-constant, the transport rate (k_b) was then proportional to the sand deposition rate:

$$k_b = \frac{k_{sand}}{\rho_{bulk} Z_{Ct}} \quad (39)$$

where ρ_{bulk} is the bulk density of soil; k_{sand} is the sand deposition rate in Yanchi area, which is a function of wind velocity (Li and Shirato, 2003):

2.4 Model validation and sensitivity analyses

2.4.1 Boundary conditions and initial values used in model simulations

In the model simulations, soil depth was set to 67.5 cm to cover the rooting zone (Gong et al., 2016), including the crust layer (2.5 cm) and sandy subsoil (65 cm, stratified into 5 cm layers). Water content measured at 70 cm depth was used as the lower boundary condition for hydrological simulation (Jia et al., 2014). The calculation of soil temperature extended to 170 cm depth with no-flow boundary, regarding the probably of strong heat exchange at the lower boundary of rooting zone (Gong et al., 2016). Zero-flow condition was set for the lower boundary of CO₂ and O₂ gases, whereas dissolved CO₂ was able to leech with seepage water. Based on presumed similarity of RLU structures, we assumed no-flux conditions for transports of water, heat, solvents and gases at outer boundary. In the simulation, we assumed instant gas transport via topcrust, whereas considered the CO₂ released by subcrust (R_{Cs}) was subject to the dissolving-transport processes. In this work, we aggregated the C



processes in subcrust with those in soil profile and set $F_B = F_{Ct}$ (Eq. (2)). The initial ratio of $M_{CA} : M_{CH}$ was set to 2:3. The C concentration of organic matters was set to 50%.

The model simulation employed half-hourly meteorological factors including the incoming shortwave radiation, incoming longwave radiation, PAR, T_a , relative humidity, wind speed and precipitation. The initial temperatures and soil moisture content for each soil layer were initialized following the work by Gong et al. (2016). Surface CO_2 concentration was set to 400 ppm. The initial gaseous CO_2 concentration was set to increase linearly with depth (5 ppm cm^{-1}). The initial CO_2 concentration in liquid form was then calculated based on Eq. (4) – Eq. (8). The initial content of mineral N content was set to 40 mg/g, which was within the range of the field observations. The two-dimensional transpirations of water, energy and gases along the soil profile were solved numerically using the Predict–Evaluate–Correct–Evaluate (PECE) method. In order to avoid undesired numerical oscillations, the transport of water, energy and gases were calculated at 5-min sub steps.

474

2.4.2 Model validation

First, we validated the modelling of soil temperature and moisture content for the F_S site (Test 0). The simulated hourly soil temperature and moisture content at 10 cm depth were compared to the measured values for each collar. The validation was based on the same meteorological data as used by Gong et al. (2016), who validated the model in regard to the diurnal to seasonal dynamics of radiation balance, surface energy balance, soil temperature and moisture content at the EC site.

The validity of the modelled F_S was examined using three separate tests. In Test 1, modelled F_S was validated for non-crust soils. In this case, F_T in Eq. (1) was the only term affecting F_S ($F_B=0$ and $F_P=0$), and the crust influences on C-water exchanges were excluded. The biocrust-related processes were considered in Test 2 and Test 3. Test 2 considered the dark respiration of biocrust (R_{Ct}), and set $F_B = R_{Ct}$ and $F_P=0$. Test 3 considered all the flux components (F_T , F_P and F_B). In these tests, different values of root biomass were assigned to the model regarding the different collar conditions (Table 1). In Test 1 – Test 3, half-hourly F_S were simulated and averaged to hourly values, and compared to the those measured from the collar C1 – C3, respectively. Linear regressions were used to compare the modelled and measured values. The biases (ζ) of the simulated values were calculated by subtracting the measured values from the modelled ones. Gap values in the measurements were omitted in the validation and the bias analyses.

492

2.4.3 Sensitivity analyses

By employing the validated model, we studied the sensitivity of simulated soil CO_2 efflux to a set of stand-specific parameters and compared the C fluxes at plant-cover and interspace soils (Test 4). This was done to find out how different flux components (i.e. P_{Ct} , R_{Ct} , F_P , F_T , R_a and R_s) to the soil CO_2



497 efflux. These simulations were based on model settings for C3 and climatic variables of year 2013-
498 2014. For the comparison purpose, same settings for soil C storages (650 gC m^{-2}), root biomass (200 g
499 m^{-2}) and biocrust were employed at both under-canopy and interspace areas.

500 Our previous work has already studied the sensitivities of modelled soil temperature and moisture
501 content to the variations in soil texture, water retention properties, vegetation parameters and plant-
502 interspace heterogeneities Gong et al. (2016). In this study, we further tested the sensitivity of the
503 modelled F_S and componential fluxes to the changes in a number of stand-specific variables and
504 parameters (Table 4).

505 We also investigated the model sensitivities for several newly defined parameters (i.e. n_R , n_p and f_m).
506 It was also studied the sensitivities of modelled C fluxes to the changes in soil temperature and
507 moisture content, which could be biased regarding the heterogeneities of soil texture, hydraulic
508 properties and vegetation covers (see e.g. Ma et al., 2011; Gong et al., 2016). We further on analysed
509 the model sensitivity to a set of biogeochemical parameters, including root biomass, total SOC ($M_{tot} =$
510 $M_{deb} + M_{mic} + M_{hum}$) in soil and topcrust (M_C), soil N content (N_{tot}), ratio of $M_{CA} : M_{CH}$, the
511 decomposability of debris (k_l), the rates of litter productions from roots (k_{mo}) and topcrust (k_{mc}) and
512 soil pH. Simulated F_S and componential fluxes at the interspace were also compared to the values
513 under no-change conditions.

514

515 3. Results

516 3.1 Model validity

517 Figure 3 shows the modelled hourly T_s and θ at 10 cm depth with the mean values measured from
518 the F_S site during year 2013. Based on the site-specific vegetation and soil texture parameters, our
519 model explained 97% of the variations in the measured hourly T_s . The model underestimated the
520 temperature mainly in summer time (i.e. day 150-250, Fig. 3a). The underestimation was most
521 pronouncing around the noontime in the diurnal cycle. The measured θ at 10 cm depth was much
522 lower at the F_S site than that shown by Gong et al. (2016) for the EC site (Fig. 3b). Such a difference
523 was in line with the coarser texture of the soil at the F_S site. The model underestimated mainly the soil
524 water content during the freezing season (Fig. 3b). During the ice-free period, it explained 83% of the
525 variations in the measured mean water contents at 10 cm depth.

526 The measured F_S showed large diurnal and seasonal variations regardless the existence of crust
527 cover (Fig. 4 and Fig. 5). Rain events clearly influenced the hourly F_S measured from the non-crust
528 surface C1 (Fig. 4a). The F_S dropped significantly from the pre-rainfall level even to near-zero, but
529 rebound rapidly and peaked after rain stopped. Our modelling reasonably reproduced the diurnal and
530 seasonal fluctuations of F_S . The model explained 87 and 83% of the variations in the hourly F_S
531 measured on the non-crust surface in year 2013 and 2014, respectively (Fig. 4a). The model mainly



underestimated the daytime F_S during the freezing seasons. During the ice-free periods, the model overestimated the efflux in early springs. The biases of modelling largely showed a diurnal pattern (Fig. 4c), that F_S was underestimated in noon hours (i.e. from 10 a.m. to 3 p.m.) but overestimated in the afternoon and evening. At the daily level, our model explained 94% of the variations in measured daily efflux during the two-year period (Fig. 4c).

Comparing to the non-crust surface (C1), the simulated F_S for the crusted surfaces (C2 and C3) exhibited greater deviations from the observations. At the hourly scale, our model explained 75 % (year 2013) and 68 % (year 2014) of the variations of measured F_S from C2 (Fig. 5a), and 68 % (year 2013) and 61 % (year 2014) of the variations of measured F_S from C3 (Fig. 5b). For the two-year period, the root-mean-square errors (RMSE) of the modelled hourly F_S were $0.25 \text{ umol m}^{-2} \text{ s}^{-1}$ and $0.35 \text{ umol m}^{-2} \text{ s}^{-1}$ for C2 and C3, respectively. The biases of the simulated F_S for C2 and C3 showed similar diurnal pattern as compared to C1, and the magnitudes of biases ($|\zeta|$) were greater during the rainfall period (i.e. from the start of raining to 24 hours after end of rainfall) than the inter-rainfall period (Fig. 6). Nevertheless, at the daily scale, the model explained 91% (C2, Fig. 5c) and 86% (C3, Fig. 5d) of the variations in the measured F_S during the two-year period. There were no significant systematic deviations between the measured and the modelled daily values, as indicated by regression slopes close to 1 and intercepts close to 0 (Figs. 4 and 5).

549

3.2 Model sensitivity

Relative contribution of component fluxes to F_S

If root biomass and SOC were homogeneous at plant cover and interspace, the C loss at interspace was faster than under-canopy on an annual basis (Test 4, Table 3). In both areas, R_s was a major contributor to the total CO_2 produced in root-zone soil and F_T dominated the effluxes (F_S) during the two-year period. The simulated NPP of topcrust were $18.2 \text{ gC m}^{-2} \text{ year}^{-1}$ and $31.1 \text{ gC m}^{-2} \text{ year}^{-1}$ at under-canopy and interspace, respectively. At hourly scale, the net C uptake by topcrust could be comparable to F_T after rewetting (Fig. 5A). However, at annual scale, the C losses via respiration and photodegradation accounted for 90% of the GPP, leading to a near-zero contribution of topcrust to F_S during the two year period (i.e. $< 5 \text{ gC m}^{-2} \text{ year}^{-1}$).

Test 4 further showed mismatched trends of F_T and the root-zone CO_2 production (R_P). The annual F_T was 17 and 15% smaller than R_P at under-canopy and interspace, respectively. Such a gap was mainly due to the root uptake and transport of dissolved CO_2 (i.e. $36 \text{ gC m}^{-2} \text{ year}^{-1}$) whereas the CO_2 loss via seepages or pore-mediated horizontal flows were limited (i.e. $7.4 \text{ gC m}^{-2} \text{ year}^{-1}$). Moreover, the temporal patterns of F_T and R_P were largely inconsistent with respect to the wetting-drying cycles (Fig. 7a). Comparing to R_P , the responses of F_T to rainfall were largely lagged and smoothed (Fig. 7b – 7d), disregard the size of rain events. R_P increased rapidly following the rewetting of soil. On the other hand, F_T firstly depressed during rainfall then increased after rain ceased. In all the examples, F_T



568 exceeded R_P within 48 hours after the ending of rain events. At the annual level, the total R_P was
 569 larger during wetting period (i.e. raining days plus 1 day after rainfall) than the rest days of the year
 570 (i.e. drying period), whereas the total F_T was greater during the drying period (Fig. 5e).

571 *Sensitivity of modelled F_S to site-specific parameters*

572 In general, the modelled F_S and the component fluxes were more sensitive to $\pm 2^\circ\text{C}$ in T_s or $\pm 10\%$
 573 in θ , comparing to the effects of $\pm 10\%$ or $\pm 20\%$ in the other parameters (Table 4). Varying θ by 10%
 574 produced greater impacts on the simulated R_P and crust-related fluxes (i.e. P_{Ct} , R_{Ct} and F_P), as
 575 compared to changing T_s by $\pm 2^\circ\text{C}$. Increasing θ by 10% enhanced the simulated P_{Ct} and NP_{Ct} of the
 576 crust by 41 and 28%, and doubled the annual C sequestration by topcrust. However, such changes in
 577 crust C uptake had minor contribution to F_S , as it amounted for only 2.0% of the total efflux.

578 Adjustment of the newly introduced parameters N_P and f_m by $\pm 20\%$ produced limited influences on
 579 the modelled F_S and the component fluxes (Table 4). The model was also robust to the adjustment
 580 of several crust-related parameters, i.e. k_{mc} , M_{Ct} and M_{CA} : M_{CH} . Comparing to $\pm 10\%$ in root biomass,
 581 varying N_R by 20% led to similar responses in the simulated F_A but much weaker responses in F_T and
 582 F_S . A 10% variation in root biomass changed the annual F_T and F_S by about 7%, and such effects
 583 were 100% greater than that of 10% changes in the total SOC (M_{tot}).

584

585 **4. Discussions**

586 **4.1 Model performance**

587 *Validity of the F_S modelling at non-crust soil*

588 Our model reasonably well captured the measured dynamics of soil temperature during year 2013,
 589 based on revaluing the vegetation and soil hydraulic parameters for the F_S site. The model mainly
 590 underestimated the midday temperatures at the collars. In addition, the model underestimated the soil
 591 moisture content during the winter freezing period, probably because the impacts of solvents on the
 592 thermal conductivity and freezing point of soil water (Viterbo et al., 1999) were not included (see
 593 Gong et al., 2016). This could have contributed to the biases in the simulated T_s and F_S for freezing
 594 period. However, the influences of such biases on annual F_S was marginal, due to the very low
 595 emissions during the winter period (see also Liu et al., 2016). During ice-free season, the simulated
 596 soil water content agreed well with the measured means, and the biases in the modelled temperature
 597 and moisture content were less than the spatial variations observed in the area (e.g. Wang et al., 2015).
 598 Therefore, our model could be able to reproduce near-realistic trends for the water-energy conditions
 599 at the site. Based on the modelled water-energy dynamics, the model well described the seasonal
 600 variations of F_S measured from the non-crust soil (Test 1) during a two-year period. The model was
 601 able to capture the strong variability of hourly/daily F_S in wetting-drying cycles, and the performance
 602 was generally better than those based on empirical methods (e.g. Wang et al., 2014a, 2014b). In this



603 sense, we assume that our model have included the main mechanisms controlling the F_s dynamics in
604 the non-crusted soil system.

605 The uncertainties in the modelling, however, may exist in several aspects. *Firstly*, the RLU was a
606 statistical simplification to the target ecosystem at footprint scale (Gong et al., 2016), and may not
607 fully capture the spatially explicitly of soil environment and biogeochemistry at the scale of FS
608 measurements. For example, the model assumed Poisson probability of mutual shading (Bégué et al.,
609 1994), and the probability of shading increased continuously with solar zenith (Gong et al., 2016).
610 However, for explicit space-time, the chance of being sunlit or shaded is more likely to be binary,
611 which possibly explain the underestimation of net radiation (Gong et al., 2016) and the collar
612 temperature around midday, affecting the simulated F_s (see Fig. 3b). Moreover, field observations had
613 considerable spatial variations in soil temperature, water content and biogeochemistry (e.g. pH, litter
614 quality and root biomass) within a distance of 3-5 meters. Such variations could well exceed a
615 magnitude of 10 %, and even over 100 % (e.g. Zhang et al., 2008; Feng et al., 2013; Wang et al.,
616 2015). Therefore, the variation of F_s driven by the spatiality of soil factors could be greater than the
617 responses to ± 2 °C in soil temperature or ± 10 % in soil water content. Therefore, future modelling
618 may need to consider spatially explicit settings, in order to further minimize the gaps between model
619 settings and the reality, and to improve the reliability of simulated F_s responses to changing climatic
620 conditions.

621 *Secondly*, the uncertainties in inorganic C processes could strongly affect the accuracy of modelling,
622 as indicated by the high sensitivity of simulated F_s to soil pH. In this model, the calculation of CO₂
623 transport was based on gaseous and liquid phases, whereas the solid phase were not involved. This is
624 likely to underestimate the dissolved CO₂ and its fluctuations, regarding the high lime content (2300–
625 5400 kg ha⁻¹) in the soil (Feng et al., 2013; Wang et al., 2015). Based on soil samples of similar lime
626 content (2700 kg ha⁻¹), Buysee et al. (2013) showed that neglecting the inorganic C exchanges of the
627 CaCO₃-H₂O-CO₂ system could underestimate F_s during the heating phase of a day, but overestimate
628 F_s during the cooling phase. Such phenomenon was very similar to the diurnal pattern of biases found
629 in our modelling (Fig. 3d). Therefore, further improvement on the modelling may need to consider the
630 solid phase as well. However, the effect of solids could be complicated, as it may differ by the mineral
631 type and degree of exposure to pore air and liquid (Buysee et al., 2013; Fa et al., 2014). Liu et al.
632 (2015) suggested that a considerable fraction of CO₂ absorbed by minerals could even be stored for
633 prolonged period with high biochemical stability. In this sense, the functioning of the CO₂ buffering
634 system may require site-specific parameterizations, in order to improve the model performance at
635 hourly level.

636 *In addition*, the current model still lacked mechanistic descriptions on the growth of plants.
637 Comparing to many other ecosystems, drylands feature high root-shoot ratio (Jackson et al., 1996) but
638 low SOC storages. Changes in plant physiology and growth can readily influence root metabolisms
639 and labile SOC pools, thus modify the climatic sensitivities of F_s (Wang et al., 2015). On the other



hand, the parameterization of soil respiration employed constant scheme throughout the years. However, the large fluctuation of diurnal and seasonal temperature may drive the microbial communities to shift between warm-adapted to cold-adapted (Van Gestel et al., 2013), which largely enhances winter respiration and its sensitivity to freeze-thaw cycles (Van Gestel et al., 2013; Liu et al., 2016). Both the biotic controls are mixed with the legacy effects of the climatic variability over annual and inter-annual courses (Sala et al., 2012; Jia et al., 2016; Shen et al., 2016), and could affect the C-water simulations cumulatively through the feedbacks between biomass accumulation and soil biogeochemistry (Bradford et al., 2016). This may explain the decrease of model validity from year 2013 to 2014 (Fig. 3, Fig. 4). Therefore, the dynamics of plants and microbial communities are required in future modelling, in order to improve the F_S simulations regarding inter-annual and long-term periods.

651

652 ***Validity of the F_S modelling at lichen-crusts soils***

Comparing to the non-crusts soil, the adding of biocrust increased the model uncertainty particularly at the hourly scale (e.g. Fig. 4). The biases of modelled F_S for the lichen-crusts soil mainly exist at hourly level after rewetting. Nevertheless, the model reasonably explained the seasonal dynamics of the measured F_S during 2013-2014, and the RMSE of the modelled hourly F_S was one-order smaller than the seasonal variation of the measured values. This indicated that the model could serve a quantitative tool to simulate seasonal and inter-annual F_S from the lichen-crusts soils in dryland ecosystems similar to the site studied here.

The uncertainty of the F_S modelling for the lichen-crusts soils was partly due to the simulated subsoil processes, as indicated by the similar diurnal patterns of biases across Test 1-3 (Fig. 3, Fig. 5). Other sources of uncertainty may be due to the simulated crust processes, and especially due to the biases of the estimated water content in topcrust (Table 4). The water content in the very thin layer of topcrust can be highly dynamical during wetting-drying cycles. Hence, it is challenging to track the photosynthesis or respiration peaks based on the hourly simulations of micrometeorology and soil hydrology. Moreover, nocturnal water inputs (e.g. dewfalls) often occur at stable atmospheric conditions. These rewetting events are important to the metabolisms of crust organisms (e.g. Liu et al., 2006). However, they are hard to be quantified precisely by EC measurement, or models derived from EC data. These uncertainties further affect together with the biases in the modelling of water-energy balances (Gong et al., 2016). Due to the lack of data on the crust moisture dynamics, it is still difficult to analyse the extent to which such uncertainties could have influenced the model validity.

The challenges to reproduce realistic trends of biocrust C uptake may also relate to the presumed homogeneity of structures of the crust layers and the consistency of C-water processes throughout the simulation. In reality, there may not be clear boundaries between topcrust and subcrust, and even topcrust itself may contain significant variations in microstructure and communities even within one



centimetre (Williams et al., 2012; Raanan et al., 2016). Furthermore, the C sequestration of biocrust not only relies on instantaneous environmental factors such as radiation, temperature and water content (Feng et al., 2014), but also depends on the dynamics of microbial communities and their interactions (Belnap, 2003; Pointing and Belnap, 2012; Maestre et al., 2015). The changes in microbial communities, on the one hand, can influence the C functioning of biocrusts directly (Feng et al., 2014; Maestre et al., 2015). On the other hand, such changes affect surface albedo (Chamizo et al., 2012), nocturnal water inputs (Liu et al., 2006), soil aggregate structure and pore forming (Williams et al., 2012; Felde et al., 2014), which ultimately feedback to microniches and water-gas fluxes (Garcia-Pichel et al., 2016). So far, many questions remain unanswered about the mechanisms that control the colonization, adaption and succession of microbial communities and the structure-function of biocrust (Pointing and Belnap, 2012). Further knowledge on these mechanisms will be helpful to improve the modelling of crust C functioning in response to climate change and extreme climatic events.

688

689 4.2 Effects of plant-interspace heterogeneity and soil C processes on soil CO₂ effluxes

Clumped distributions of foliage and biomass are critical features for the adaptation and functioning of vegetation in arid and semiarid environments. Previous studies have mainly emphasized the effects of shrub pattern on ecohydrology (e.g. Rongo et al., 2006; Gong et al., 2016) and enrichment of fine sediments and nutrient, known as “resource island” effects (Reynolds et al., 1999; Rietkerk et al., 2004). Our simulations showed that the presence of shrub canopy could also influence the soil C exchanges. Comparing to the interspace, the presence of shrub cover reduced the simulated F_S by 13% annually (Test 4). As the soil SOC and root biomass were set to be the same at under canopy and interspace in the simulation, such a decrease in F_S was probably due to the cooling effect of canopy (Gong et al., 2016) on soil. Such an effect was close to the modelled responses of F_S to ± 2 °C in soil temperature or $\pm 10\%$ in soil water content. As the density of roots and litter production are commonly larger under canopy than interspace (e.g. Zhang et al., 2008), the lower respiration rate under canopy tends to facilitate the accumulation of biomass and organic matters in under-canopy soils and feedback to functioning of “resource islands” during prolonged periods. In this context, the different C functioning at plant cover and interspace shall not be neglected in studies on the climatic sensitivity of C dynamics in dryland ecosystems.

Our modelling provide a way to separate the multiple soil C processes and investigate their roles in regulating F_S dynamics in dryland ecosystems. So far, efforts to quantify the soil C loss in terrestrial ecosystems often consider soil C efflux as a synonym of respired CO₂. However, based on this work cautions must be taken when extrapolating the F_S responses from the chamber to ecosystem scale and from short-term to long-term periods. Our simulations reckoned that a considerable fraction of CO₂ produced could be removed by root uptake and leave the volume measured by the respiration chamber. Bloemen et al. (2016) showed that the CO₂ concentration in root xylems could be higher than in soil



712 solutions. This implies that such a “missing source” might be even greater than the model estimation,
 713 although our knowledge is still limited about the efficiency of the removal and the diffusion/release of
 714 CO₂ during the transport (Bloemen et al., 2016). On the other hand, the soil processes other than
 715 autotrophic and heterotrophic respiration could significantly modify the F_s responses to climatic
 716 variability. Our simulation highlighted decoupled CO₂ productions and emissions during the wetting-
 717 drying cycle, as regulated by the CO₂ transport in soil profile. The simulated CO₂ productions in soil
 718 profile were much greater than effluxes during rain pulses (e.g. Fig. 7). This indicated that, the limited
 719 FS during rewetting was mainly due to the increased dissolution of CO₂, rather than the reduced
 720 respiration rates by low O₂ supply (e.g. Fang and Moncrieff, 1999). This finding is further supported
 721 by the measurement of Maier et al. (2011), which showed that 40% of the respired CO₂ could be
 722 stored temporally in soil pore-space after rainfalls. The dissolved CO₂ then released gradually with the
 723 evaporation of pore water, leading to lagged responses of efflux as compared to respirations.
 724 Regarding that a major fraction of CO₂ was produced during the wetting periods (Fig. 5e), such a
 725 lagging effect should be carefully examined when analysing the climatic sensitivity of F_s .

726 Accounting for the effects of biocrust organisms is also important to accurately estimate C
 727 exchanges in dryland ecosystems (Maestre and Cortina, 2003; Castillo-Monroy et al., 2011). Existing
 728 studies on the C functioning of biocrust have mainly focused on measuring the net C exchanges under
 729 laboratory conditions or from field (e.g., Maestre and Cortina 2003; Wilske et al., 2008; Bowling et
 730 al., 2011). However, the contributions of biocrusts as C sink or source remain largely unknown
 731 (Castillo-Monroy et al., 2011), due to the difficulty to separate the CO₂ exchanges of crust organisms
 732 from the background respirations (Castillo-Monroy et al., 2011; Sancho et al., 2016). In this sense, use
 733 of our mechanistic model may provide more insights on the roles of biocrusts in soil C cycling and
 734 effluxes. our simulation study showed that the C exchanges of biocrust was largely masked by the
 735 background effluxes from root-zone soil. The C uptake by biocrust turned the soil from a net CO₂
 736 source to sink during large rewetting events only, when the background emission was restricted. After
 737 rain stopped, the sinks diminished quickly (e.g. within 1 day, Fig. 5b – 5c), not only due to the
 738 decreased photosynthesis with drying, but also the increased CO₂ emission from the soil underneath.
 739 Based on the climatic variables of a two-year period (Test 4), the simulated NPP of the topcrust was
 740 31 g C m⁻² year⁻¹ at the interspace conditions. Considering a 30% coverage of lichens over the
 741 sampling area (Feng et al., 2014), the interspace-level NPP was 9.3 g C m⁻² year⁻¹. This value was
 742 largely greater than the lab-based estimation for the site (Feng et al., 2014), but it was in range of the
 743 values reported from several other dryland ecosystems (i.e. 5–3 - 29 g C m⁻² year⁻¹, Sancho et al.,
 744 2016).

745 Our simulations also suggested that photodegradation might offset about 48 % of the CO₂
 746 photosynthesized by biocrust and reduced the net sequestration to about 5 gC m⁻² year⁻¹. It could
 747 explain the much higher F_s measured from the transparent chamber (C3) than the opaque chamber
 748 (C2) during dry daytime periods (e.g. Fig. 8). It should be also noticed that the litter from shrub



canopy was not included in the measurement nor modelling. Also, the interactions between photodegradation and biotic decaying were not considered either. Therefore, the contribution of photodegradation to soil C balance could be greater than our estimation at the ecosystem level (see e.g. Gliksman et al., 2016). Future studies are therefore required to clarify the role of photodegradation in regulating the C turnover and sequestration of biocrusts in arid and semiarid ecosystems.

5. Conclusions

This work represents a first attempt to integrate the CO₂ production, transport and surface exchanges (e.g. biocrust photosynthesis, respiration and photodegradation) in F_s modelling for dryland ecosystems of high plant-interspace heterogeneities. Our model simulated reasonably well the F_s dynamics measured from non-crust and lichen-crust soil collars during year 2013-2014, although introducing the gas exchanges of lichen crust into modelling decreased the model performance at the hourly scale. Our model could thus be used to simulate the seasonal and annual F_s in dryland soils similar to our site. However, further development of model may still be required on several aspects, e.g. by including: i) the spatial-explicit schemes for surface conditions and soil biogeochemistry; ii) influences of lime and solids on CO₂ transport; iii) growth dynamics of plants; iv) high-resolution dynamics of surface water-thermal conditions and v) the dynamics of microstructure and microbial communities of biocrusts.

Our modelling work also highlighted that, the plant-interspace heterogeneity and complexity of soil C processes could affect largely the soil CO₂ efflux. The presence of plant cover tended to decrease the CO₂ production from root-zone soil probably due to the cooling effect of canopy. Moreover, the transport processes of inorganic C and the metabolisms of biocrusts strongly modified the CO₂ efflux, and these influences are closely linked to soil hydrology. The CO₂ emission from root-zone soil also delayed by increased CO₂ dissolution after rewetting. In addition, an ineligible fraction of respired CO₂ could be removed via lateral flows and root uptakes, and be “missing” from volumes under respiration chambers. During rewetting, the lichen-crust soil could shift from net CO₂ source to sink, due to the activated photosynthesis of lichens and the restricted CO₂ emissions from subsoil. Whereas after rain events, the NPP of lichens could be easily masked by the background C emissions from the soil profile. Based on our modelling, the annual NPP was 9.3 gC m⁻² by topcrust at interspace. However, the net C sequestration by topcrust could be marginal, if the photodegradation is accounted.

To conclude, our work suggests that the complexity and plant-interspace heterogeneities of soil C processes affect largely the soil CO₂ efflux dynamics and their climatic sensitivities, which should be carefully considered in extrapolation of findings from chamber to ecosystem level and from seasonal to inter-annual scales. Our model can also serve as a useful tool to simulate the soil CO₂ efflux dynamics in dryland ecosystems.



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1067 Tables

1068

1069 Table 1. Configuration of soil collars used in this study

Collars	C1	C2	C3
Surface type	Non-crusted	Lichen-crusted	Lichen-crusted
Chamber type	Opaque	Opaque	Transparent
Root biomass (g m^{-3})	420	106	92
Gap of data (%)	12.9	10.5	9.85
Annual C efflux (gC m^{-2}) ^a	259	194	192

1070 ^a The values were calculated from the measured hourly FS data excluding data gaps.

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1072

1073 Table 2. Parameters for soil water retention and C turnover

Parameter	Equation	Unit	Value
α_h	- ^a	-	0.0355 ^b
n	- ^a	-	1.5215 ^b
k_1	(11)	$\text{g g}^{-1} \text{day}^{-1}$	0.01 ^c
k_2	(11)	$\text{g g}^{-1} \text{day}^{-1}$	0.08 ^d
k_3	(11)	$\text{g g}^{-1} \text{day}^{-1}$	0.001 ^d
k_g	(15)	g g^{-1}	0.15 ^e
k_{cr}	(19)	$\text{g g}^{-1} \text{s}^{-1}$	0.0014 ^f
k_{R0}	(25)	$\text{g g}^{-1} \text{day}^{-1}$	0.002 ^e
a	(26)	-	0.1 ^g
b	(26)	-	24 ^g
c	(26)	-	0.89 ^g
Q_{Cr}	(32)	-	1.585 ^f
a_{RC}	(32)	-	-0.0525 ^f
b_{RC}	(32)	-	2.602 ^f
c_{RC}	(32)	-	-1.653 ^f
a_{Pt}	(33)	-	0.9837 ^f
b_{Pt}	(33)	-	-0.1385 ^f
c_{Pt}	(33)	-	0.0095 ^f
d_{Pt}	(33)	-	-1.6318E-4 ^f
a_{Pw}	(33)	-	-0.3501 ^f
b_{Pw}	(33)	-	5.5884 ^f
c_{Pw}	(33)	-	-7.1783 ^f
d_{Pw}	(33)	-	2.6837 ^f

1074 ^a See Eq. (26) in Gong et al. (2016). Sources of parameter values: ^b This study, see section 2.3.2; ^c Lai et al.

1075 (2015); ^d Gong et al. (2014); ^e Chen et al. (1999); ^f This study, see section 2.3.4 and Fig. 2; ^g Wang et al., 2014a.

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Table 3. Simulated annual F_S ($\text{gC m}^{-2} \text{ year}^{-1}$) and its componential fluxes ($\text{gC m}^{-2} \text{ year}^{-1}$) at plant cover and interspace

Surface type	F_S	F_T	R_P^a	R_a	P_{Ct}	F_{Ct}	F_P	F_{Cnet}^b
Interspace	244	249	295	113	54.6	31.1	26.1	5.0
Plant covered	214	218	263	108	36.3	18.2	14.6	3.6

^a R_P is the total respired CO_2 in root-zone soil and is the sum of autotrophic respiration ($R_a = \sum_i R_{a_i}$, see Eq. (10)) and heterotrophic respiration ($R_s = \sum_i R_{s_i}$, see Eq. (12)); ^b F_{Cnet} is the net CO_2 exchanges of topcrust and $F_{Cnet} = F_{Ct} - F_P$, see Eq. (17) – Eq. (18).

Table 4. Sensitivity of simulated F_S and its componential fluxes to changes in parameter values

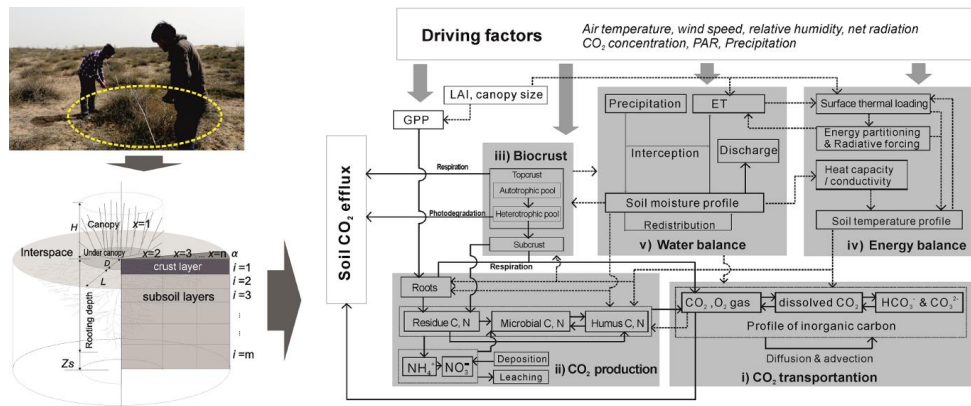
Change of parameter	F_S	F_T	R_P	R_a	P_{Ct}	F_{Ct}	F_P	F_{Cnet}
$n_R + 20\%$	+3.3	+3.2	+2.7	+7.9	/	/	/	/
$n_R - 20\%$	-2.9	-2.8	-3.4	-8.8	/	/	/	/
$n_P + 20\%$	+1.6	+1.6	+1.0	/	/	/	/	/
$n_P - 20\%$	/	/	-1.4	/	/	/	/	/
$f_m + 20\%$	/	/	/	/	+2.9	+3.8	+3.4	+6.0
$f_m - 20\%$	/	/	/	/	+1.2	/	-5.7	+30
$T_S + 2\text{ }^\circ\text{C}$	+9.5 ^a	+9.6	+7.1	+11	+4.9	+3.9	+1.5	+16
$T_S - 2\text{ }^\circ\text{C}$	-9.0	-9.2	-8.1	-11	-1.3	-2.9	/ ^b	-20
$\theta + 10\%$	+3.6	+5.6	+7.5	+14	+41	+28	+14	+102
$\theta - 10\%$	-5.0	-5.6	-8.1	-14	-16	-13	-8.4	-34
$M_{tot} + 10\%$	+2.9	+2.8	+2.0	/	/	/	/	/
$M_{tot} - 10\%$	-2.5	-2.4	-3.1	/	/	/	/	/
$M^R + 10\%$	+7.0	+6.8	+6.8	+8.8	/	/	/	/
$M^R - 10\%$	-7.0	-6.8	-7.1	-8.9	/	/	/	/
$N_{tot} + 10\%$	/	/	/	/	/	/	/	/
$N_{tot} - 10\%$	/	/	/	/	/	/	/	/
$k_I + 10\%$	+2.9	+2.8	+2.4	/	/	/	/	/
$k_I - 10\%$	-2.5	-2.4	-3.1	/	/	/	/	/
$k_{mo} + 10\%$	+4.1	+4.0	+3.4	/	/	/	/	/
$k_{mo} - 10\%$	-3.3	-3.2	-3.7	/	/	/	/	/
$k_{mc} + 10\%$	/	/	/	/	/	/	+1.5	-8.0
$k_{mc} - 10\%$	/	/	/	/	/	/	-2.3	+8.0
$M_{Ct} + 10\%$	/	/	/	/	/	/	/	/
$M_{Ct} - 10\%$	/	/	/	/	/	/	/	/
$M_{CA}:M_{CH} + 10\%$	/	/	/	/	/	/	/	/
$M_{CA}:M_{CH} - 10\%$	/	/	/	/	/	/	/	/
pH +5 %	-8.6	-8.4	/	/	/	/	/	/
pH -5 %	+7.0	+6.8	/	/	/	/	/	/

^a Percentage (%) of changes in the C flux after manipulation of parameter values, as compared to the “base” conditions (i.e. Test 4; see Table 3). All the tests were based on the interspace conditions. ^b The change in the simulated C flux was smaller than 1 %.



1088 Figures

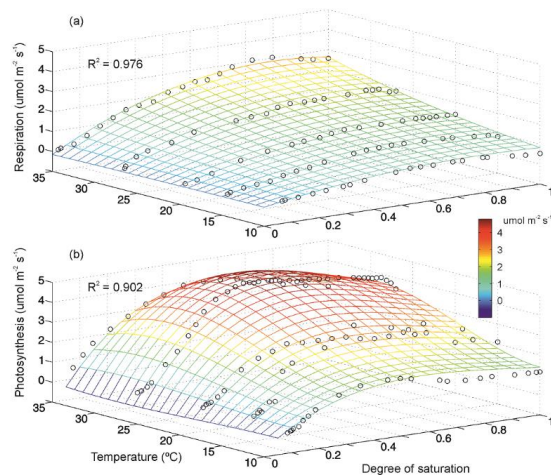
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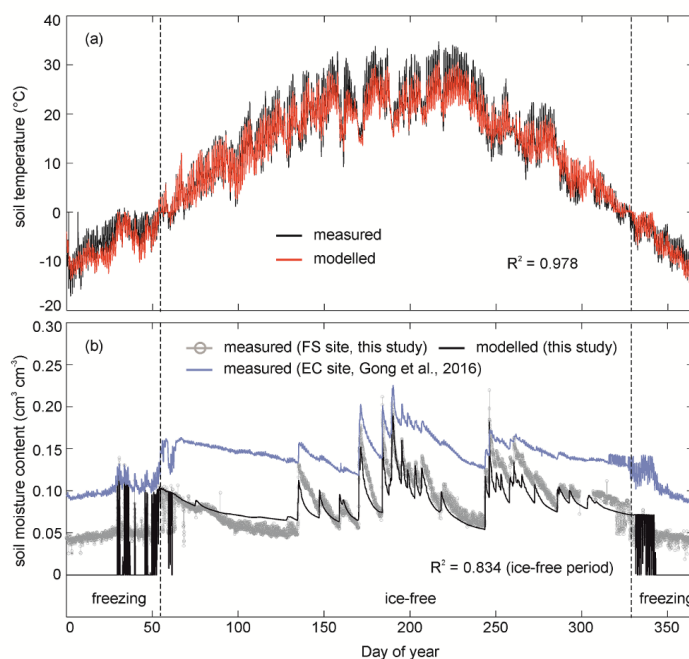
1091 Figure 1. Layout for the setting of representative land unit (RLU, as adopted from Gong et al., 2016)
1092 and conceptual framework of process-based modelling

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1094

1095 Figure 2. Measured and fitted bulk respiration (a) and photosynthesis (b) of the lichen topcrust as
1096 functions of temperature and water content.

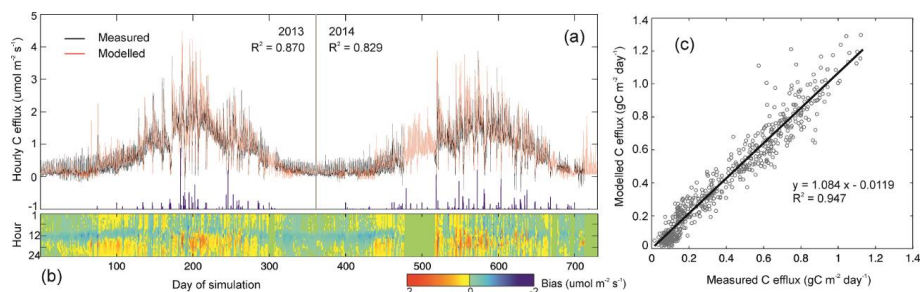


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1098 Figure 3. Measured and modelled soil temperature (a) and soil moisture content (b) at 10 cm depth for
 1099 FS site, and as compared to the EC site in year 2013 by Gong et al. (2016).

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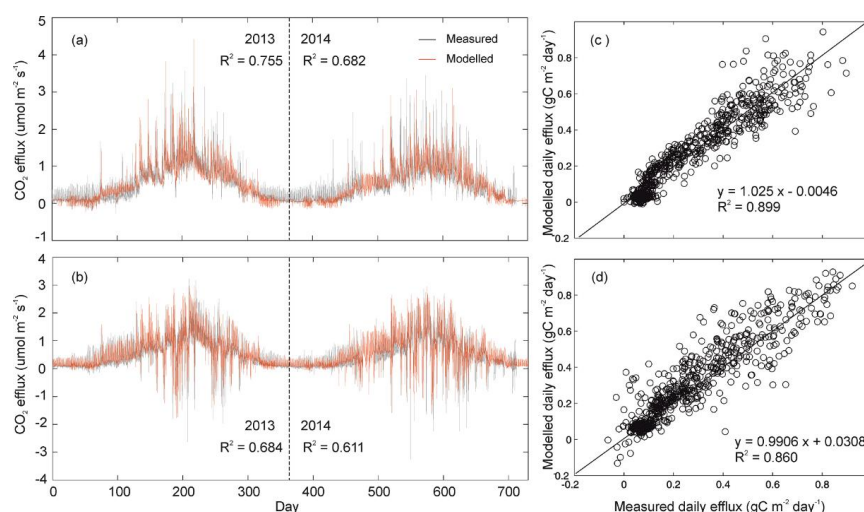
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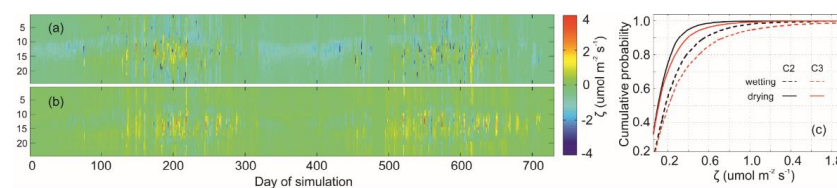
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1103 Figure 4. Measured and modelled hourly FS for non-crustified soil (a), the temporal pattern of the bias of
 1104 simulated hourly FS (b) and the comparison of measured and modelled daily FS (c) during 2013-2014.

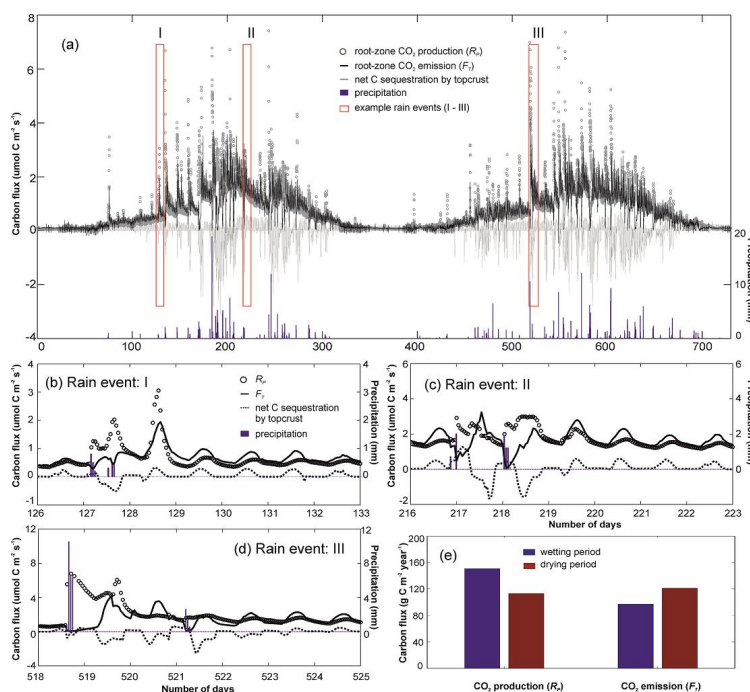
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 1107 Figure 5. Measured and modelled F_s of lichen-crusted soils for opaque (a, c) and transparent
 1108 chambers (b, d) at hourly (a, b) and daily (c, d) scales during 2013-2014.



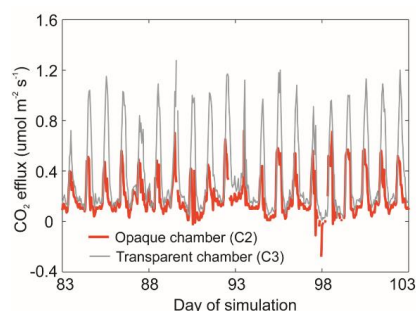
1110
 1111 Figure 6. Diurnal patterns of biases in the simulated hourly F_s for lichen-crusted soils using opaque (a)
 1112 and transparent chambers (b), and the cumulative probability of the biases during wetting and drying
 1113 periods (c) during 2013-2014. The wetting period included the raining days and a 1-day period after
 1114 each rainfall. The drying period included the rest time of the years other than the wetting period.



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1117 Figure 7. Simulated F_S and CO₂ exchanges by biocrust and root-zone soil (a), the simulated CO₂
 1118 fluxes before and after example rain events of 2.3 mm (b), 7.6 mm (c) and 12.8 mm (d) sizes, and the
 1119 comparison of F_T and R_R during wetting and drying periods during 2013-2014. The wetting period
 1120 included the raining days and a 1-day period after each rainfall. The drying period included the rest
 1121 time of the years other than the wetting period.

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1123

1124 Figure 8. Comparison of the measured F_S from lichen-crusted surfaces using opaque and transparent
 1125 chambers during a dry period (day 83-103) in spring 2013.