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TITLE

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SUMMARY **(320 Words)**

1   The priming effect (PE) is a key mechanism contributing to the carbon balance of the soil ecosystem. Almost 100 years of research since its discovery in 1926 have led to a rich body of scientific publications to identify the drivers and mechanisms involved. A few review articles have summarised the acquired knowledge; the last major one was published in 2010. Since then, knowledge on the soil microbial communities involved in PE and in PE + C sequestration mechanisms has been considerably renewed.

2   This article reviews current knowledge on soil PE to state to what extent new insights may improve our ability to understand and predict the evolution of soil C stocks. We propose a framework to unify the different concepts and terms that have emerged from the international scientific community on this topic, report recent discoveries, and identify key research needs.

3   Seventy percent of the studies on the soil PE were published in the last 10 years, illustrating a renewed interest for PE, probably linked to the increased concern about the importance of soil carbon for climate change and food security issues. Among all the drivers and mechanisms proposed along with the different studies to explain PE, some are named differently but actually refer to the same object. This overall introduces "artificial" complexity for the mechanistic understanding of PE, and we propose a common, shared terminology. Despite the remaining knowledge gaps, consistent progress has been achieved to decipher the abiotic mechanisms underlying PE, together with the role of enzymes and the identity of the microbial actors involved. However, including PE into mechanistic models of SOM dynamics remains challenging as long as the mechanisms are not fully understood. In the meantime, empirical alternatives are available that reproduce observations accurately when calibration is robust.

4   Based on the current state of knowledge, we propose different scenarios depicting to what extent PE may impact ecosystem services under climate change conditions.

KEY WORDS

Soil, Priming effect, Carbon cycle, minerals, enzymes, microbial actors, models, ecosystem services

1   INTRODUCTION

Soils host one of the main carbon (C) stores of our planet, around three times more than the atmosphere does (2,400 *vs*. 800 GtC) (Friedlingstein et al., 2019). Most of this C is stored in organic matter whose turnover controls soil fertility through the release of nutrients for plants. Soil Organic Matter (SOM) also releases huge amounts of CO_2 into the atmosphere, 7-fold more than anthropogenic emissions (Schlesinger and Andrews, 2000). Given the huge size of the soil organic C pool, any change in the soil C content and its dynamics at a global scale – even if tiny – is associated to great impacts in terms of food security and climate change (Dignac et al., 2017; Minasny et al., 2017). Consequently, preserving and even increasing organic carbon in soils is of prime importance to meet the food security challenge and mitigate greenhouse gas emissions (Minasny et al., 2017). This requires getting insights into the determining mechanisms of the soil carbon balance – i.e. mineralisation *vs.* sequestration.

The priming effect (PE) is a key mechanism contributing to the ecosystem carbon balance (Fontaine et al., 2007; Guenet et al., 2018). It corresponds to a change in the rate of SOM-C mineralisation by soil microorganisms resulting from the input of fresh organic matter (FOM) (Bingeman et al., 1953). The definition was later extended to almost any variation of microbial activities after an addition of molecule (e.g., toxic compounds; Kuzyakov et al. 2000), but the present review will focus on ~~C related~~ PE as previously defined, and more precisely on the plant induced priming effect (PIPE) which includes the rhizosphere priming effect (RPE) but also more generally all priming induced by plant originated carbon inputs (Box 1). The PE occurs everywhere, but it has mainly been studied in soils (Bengtsson et al., 2018; see supporting information). It has long been viewed as a net soil C loss since the most frequent effect reported in the literature is a stimulation of native SOM mineralisation, up to 400 % according to the type of environment and/or soil conditions (Guttière et al., 2021; Chen et al., 2019; Perveen et al., 2019, Blagodatskaya & Kuzyakov, 2008). However, it can also be negative, with up to 50% reductions in soil carbon emissions (Lyu et al., 2018; Chen et al., 2019). As the PE is a complex process, its impact on global C fluxes is hard to grasp and should be considered as a whole, integrating feedbacks between the different and sometimes contrasting effects it causes: for example, the increase of SOM mineralisation has negative consequences on climate change, and the nutrients released from organic matter may be leached down the soil profile; but when the released nutrients are taken up by plants, the PE indirectly contributes to plant growth and additional C sequestration in the plant-soil system. Thus, the effect of PE variation on the net soil C balance is not trivial, particularly in the context of climate change because C inputs may globally increase through enhanced plant

73 productivity, while soil C mineralisation may increase too due to higher temperature (Nishina et al.,
74 2014). Consequently, better knowledge of the mechanisms involved in PE is of fundamental
75 importance to improve our ability to control and predict the evolution of soil C stocks (Guenet et al.,
76 2018; Chen et al., 2019), in agreement with the 2019 IPCC report on land (Shukla et al. 2019) that
77 pointed out the need to better understand how PE will respond to climate change.

78

79 Since the pioneering studies that evidenced it (Löhnis, 1926) and due to its importance for C cycling,
80 the PE has prompted a large body of research work to identify the drivers and mechanisms involved
81 (Kuzyakov et al., 2000; Blagodatskaya & Kuzyakov, 2008). A bibliometric analysis of the scientific
82 articles published in the field during the last 50 years shows however that 70% of all the articles are
83 recent, *i.e.* published during the last ten years, and mainly motivated by the rising concern about the
84 importance of soil carbon for climate change and food security issues. We provide as supplementary
85 information a brief report of the main outputs of this analysis highlighting the dynamics of publication,
86 the extent of the associated scientific network and the topical drivers of the publications
87 (Supplementary information available from Zenodo: https://doi.org/10.5281/zenodo.6326056).

88

89 Based on these studies, it is clear today that the PE involves different microbial processes sometimes
90 driven in opposite ways by environmental parameters. Soil properties, climate, land use (Bastida et al.,
91 2019; Siles et al., 2022), but also the nature of the substrates entering the soil (*e.g.*, simple compounds,
92 complex compounds, plant residues or living roots) and the way they are added (*e.g.* continuous *vs.*
93 puse) all impact the intensity of priming effect. A few reviews have been published to summarise
94 knowledge in the field. The last major one was published in 2010 (Kuzyakov, 2010). However, in the
95 last ten years, studies aimed at characterising the microbial communities involved in soil processes
96 such as PE have benefited from considerable technical progress providing deeper insights into
97 microbial diversity and activity (Bouchez et al., 2016). In the same time span, the understanding of C
98 sequestration mechanisms has undergone a paradigm change (Lehmann & Kleber, 2015). Therefore,
99 it appeared important to update knowledge on PE mechanisms by integrating the findings of the last
100 10 years, which was the first objective of the present review. The second objective was to unify the
101 different concepts and terms that emerged from the international literature on this topic. Finally, we
102 identified and focused on specific points that need to be more deeply examined to evaluate and
103 predict the size and the impact of the different mechanisms across space and over time.

104
105 2   UNIFYING THE DIFFERENT GENERATION MECHANISMS.

106

107 A large number of mechanisms have been proposed to explain PE generation (Table 1). Some of them

108 are similar or even equivalent but have been named differently. This section (i) aims to clarify the

109 different denominations, (ii) groups them according to their effect on C fluxes, and (iii) integrates them

110 on a time scale. Box 1 gathers the terminology associated to the different pools of organic matter and

111 types of PE on which we based our discussion.

112

113   2.1   Let's clarify the denomination of the different mechanisms.

114 The PE is usually measured as the difference between the CO_2-C released from an amended soil

115 (excluding CO_2-C originating from the amendment itself) and from a non-amended control soil over a

116 certain time period, leading to a "positive" or "negative" PE (Kuzyakov et al., 2000).

117 Negative PE have been attributed to "preferential FOM substrate utilization" by regular native SOM

118 decomposers (Cheng 1999). SOM decomposers can also switch from SOM decomposition to the

119 decomposition of FOM-decomposer necromass after intense predation (Cui et al. 2020). This process

120 has been called "necromass recycling" and also leads to reduced SOM decomposition and

121 mineralisation over time. Viral lysis has been suggested to generate more microbial necromass than

122 predation by protists or nematodes (Kuzyakov and Mason-Jones, 2018), but nobody has yet published

123 on the link between bacteriophages and PE.

124

125 Positive PE processes can be called "apparent" PE when they only result from the replacement of

126 microbial cell C by C from the new substrate (labile fraction). Apparent PE have been described under

127 3 complementary processes (triggering effect, microbial activation, and pool substitution), depending

128 on the amount of labile FOM added compared to the total microbial biomass (Blagodatskaya and

129 Kuzyakov, 2008 – see Table 1).

130

131 Real positive PE have been attributed to multiple processes. Those called "stoichiometric

132 decomposition" (Chen et al., 2014), and "mechanism 1" (Fontaine et al., 2003) refer to an acceleration

133 of SOM decomposition generated by the increase in extracellular enzymes produced by FOM

134 decomposers (Table 1). This implies that certain FOM and SOM compounds have chemical similarities

135 (*e.g*., a plant origin). Therefore, FOM should be at least partly polymerised (such as in residues, litter

136 or dead roots). Both processes occur when N is not limiting, but SOM catabolites benefit SOM

137 decomposers in "Mechanism 1", and FOM decomposers in "stoichiometric decomposition". Processes

138 called "N-mining" (Chen et al., 2014) and "mechanism 2" (Fontaine et al., 2003) result from the energy

139 brought by FOM catabolites, for SOM decomposers to break down biochemically recalcitrant SOM for

140 nutrients. For both processes, the targeted SOM pool was supposed to be also of plant origin. One

141 process has been called "co-metabolism" in different papers and corresponds to the assimilation of
142 FOM and SOM catabolites by either FOM or SOM decomposers, without being aimed at N acquisition
143 (Kuzyakov, 2000; Yu et al., 2018). Therefore, in terms of carbon, this process overlaps with previously-
144 described mechanisms (either Mechanism 1 or Mechanism 2 and nutrient mining) and thus is not
145 discriminant.
146
147 A new mechanism of positive PE has been proposed recently to explain the rhizosphere PE (Keiluweit
148 et al., 2015). Roots can exude organic acids, which cause the release of labile organic compounds from
149 protective associations with minerals, and then these compounds can be decomposed by microbes.
150 As a large part of this mineral-associated organic matter (MAOM) is derived from microbial necromass
151 – with a low C:N ratio – through the so-called "entombing effect" (Liang et al., 2017), many nutrients,
152 historically considered as a relatively passive reservoir, can be remobilised by this process, that we
153 called "Abiotically mediated PE" (see section 3).
154
155   2.2   Gathering mechanisms according to their effect on C fluxes
156 Two important parameters have to be considered to evaluate the impact of the different PE generation
157 mechanisms on the dynamics and the fate of primed C.
158
159 Firstly, native SOM is a mixture of different pools characterised by different turnover rates depending
160 on their biochemical nature and their accessibility to enzymes. The organic matter that enters the soil
161 is mainly of vegetal origin. Plant tissues are composed of molecules easily (hemicellulose, cellulose) or
162 less easily (lignins, polyphenols, phytates) broken down by microbial extracellular enzymes, hence fast
163 to medium-long turnover rates (a few days to a few years – Derrien et al., 2006). In contrast, the SOM
164 pools showing the lowest turnover rate (centuries) are composed of compounds from both plant and
165 microbial sources, protected from enzymes by mineral association (Lehman & Kleber, 2015). Carbon
166 primed from a SOM pool with a fast turnover rate should impact long-term carbon sequestration less
167 than C remobilised from a pool with a long residence time.
168
169 Secondly, PE is often evaluated based on the change in CO_2 mineralisation rate, but respiration can be
170 accompanied by incorporation of C into cell biomass (growth) (see Geyer et al., 2016 for a review). The
171 efficiency of carbon incorporation into biomass (*i.e.*, carbon use efficiency, CUE) is calculated as the
172 amount of microbial biomass produced *per* unit of resource consumed. The CUE of cells can vary
173 between 0 (no growth but alive) to 0.85 (growth but still needing some energy for cell growth and
174 maintenance) depending on their physiology and life strategy. The link between life strategy and CUE
175 depends on the physiological status of microbes. R-strategists are actually less efficient than K-

strategists in using their assimilated C during a resting phase, due to a costlier basal metabolism. But as these populations have higher maximal growth rates, during the growth phase when substrate is not limiting, a higher proportion of C is allocated to biomass instead of respiration. This leads to a globally higher CUE than that of K-strategists (Geyer et al., 2016). As PE generation mechanisms are consequences of organic matter decomposition, they are more related to growth phases of different populations than to their resting phases. Thus we consider that during a PE event, r-strategists have high CUE while K-strategists have low CUE. Therefore, primed SOM assimilated by high-CUE populations is likely converted into microbial biomass rather than mineralised into CO_2, and further stabilized via the so-called "entombing effect" (Lehman & Kleber, 2015; Liang et al., 2017). As a result, high-CUE populations increase the chance of moving C from a fast- or a medium-turnover SOM pool to a slow-turnover pool.

We classified the different processes proposed in the literature according to both previously-described parameters: the turnover rate of targeted SOM pool and the CUE of microbial populations assimilating its derived C (Table1). It is interesting to note that, to date, no study has revealed the SOM pool impacted by the negative PE.

       2.3   Time scale integration of PE generation mechanisms.

When FOM enters the soil, its nitrogen content will help cells synthesize new enzymes for further FOM or SOM depolymerization. This phase of microbial turnover (apparent PE) first concerns high-CUE microbial populations and low-CUE populations depending on the size of the labile fraction of FOM (Figure 1). If the labile FOM pool exceeds the total microbial biomass and under non limiting N, SOM-decomposers will prefer the FOM substrate (negative PE) decreasing the regular SOM-C flux to the atmosphere. If the labile FOM pool is exhausted before the available N, FOM decomposer exoenzymes, liberated against the polymerized part of FOM will also attack SOM of vegetal origin. High-CUE populations will be more competitive than low-CUE populations under such conditions and will increase their biomass (stoichiometric PE). High-CUE population necromass will be rapidly recycled by low-CUE populations or stabilised by mineral association (Figure 1). When the soil or FOM-available N is exhausted, FOM catabolites will bring some energy to microbial populations able to catabolise recalcitrant SOM for N (N-mining), or to dissociate N-rich SOM from minerals, both population types having a low CUE. Nitrogen remobilised from SOM increases the soil nitrogen status and allows for a next phase driven by high-CUE populations (Fang et al., 2018; Vain et al., 2021).

3   ABIOTICALLY MEDIATED PE: CONCEPTS AND MECHANISMS

210   3.1   Mineral-associated organic matter (MAOM)

211 In models of organo-mineral associations, mineral phases involved in organo-mineral associations can

212 be classified into three main categories: well-crystallised minerals, short-range order phases, and

213 metallic monomers. Each category is associated with a distinct conceptual model of organo-mineral

214 association (Figure 2):

215   - *Adsorption*: organic molecules are adsorbed onto the mineral surface of well-crystallised minerals

216   (*e.g.*, smectite, vermiculite or goethite) (Kleber et al., 2007).

217   - *Coprecipitation*: organic molecules are coprecipitated with short-range order phases or

218   amorphous small polymers of Al, Fe and Si (Basile-Doelsch et al., 2015; Tamrat et al., 2019). Co-

219   precipitation is thought to be the dominant model as regards the amount of stabilised C

220   (Rasmussen et al., 2018).

221   - *Complexation and bridging*: organic molecules are bound to monomers (single ions). The main

222   metals involved in complexation are Al and Fe in acidic soils. At higher pH, Ca dominates the

223   organo-mineral associations by cation bridging (Rowley et al., 2018). Cations surrounded by

224   organic molecules form a tertiary structure described as "supramolecular aggregates" (Sutton &

225   Sposito, 2005).

226 The existence of these different types of organo-mineral models should be taken into account when

227 studying MAOM in order to understand PE mechanisms.

228

229 In each type of organo-mineral association model described above, an additional degree of complexity

230 comes from the binding energies. Regardless of the model, the five following types of organic-inorganic

231 bonds can be present: covalent bonding, electrostatic cation bridges, hydrophobic interactions, Van

232 der Waals forces, and H bonding (von Lutzow et al., 2006). The binding strength varies very strongly

233 depending on the type of bond (Newcomb et al., 2017). Therefore, the energy required to break these

234 bonds and destabilise organic molecules is also a very important parameter in the control of the

235 MAOM PE. However, while this parameter seems thermodynamically easy to conceptualise, its

236 consideration in PE experiments is still in its infancy (Li et al., 2021).

237

238 The mineral-associated organic matter is less accessible but not fully inaccessible. The bonds between

239 organic molecules and inorganic phases slow down OM degradation. Incubation studies have shown

240 that MAOM is more difficult to mineralise than free OM, whatever the model (adsorption,

241 coprecipitation or complexation) (*e.g.* Jones & Edwards, 1998; Saidy et al., 2015). The proportion of C

242 mineralised during MAOM incubations reached a plateau which, depending on the compounds, barely

243 reached the same proportion as in the control where OM was not associated with mineral phases.

244 These experiments showed that (i) part of MAOM remained accessible to microorganisms but

245   mineralisation was slowed down, and (ii) depending on the nature of the organic compounds and the
246   nature of the mineral (*i.e.*, depending on the nature of the chemical bond), part of MOAM was not
247   mineralised by microorganisms. Thus, rather than use a binary terminology tending to say that OM is
248   stabilised or not in organo-mineral associations, a stabilisation gradient concept should be preferred
249   and considered as a key parameter driving the potential of abiotically mediated PE.

250   3.2.   Priming effect of mineral-associated organic matter
251   The above-mentioned experiments did not take the full complexity of soil functioning into account.
252   When the physicochemical conditions of the soil solution are modified (changes in pH, redox
253   conditions, ionic strength, composition), the speciation of the mineral phases and organic molecules
254   can be modified, leading to an altered configuration of organo-mineral associations. During these
255   changes, if OM becomes more easily accessible to microbial enzymes, accelerated decomposition (*i.e.*,
256   a PE) may be observed.
257
258   The biotic impact on the physico-chemical parameters of MAOM leads to a direct and indirect PE
259   pathways. Roots and microorganisms play an essential role in controlling the physicochemical
260   conditions of the soil solution (Figure 2). Multiple factors impact these conditions. For example,
261   vascular plant roots, fungi and bacteria release low-molecular-weight organic acids. These are mainly
262   carboxylic acids like formic, oxalic, and acetic acid, but also aliphatic and aromatic acids (Clarholm et
263   al., 2015 and references therein). Microorganisms may decrease the pH locally in microsites
264   (Bonneville et al., 2011). Roots also release some easily metabolisable simple sugars (*e.g.*, glucose).
265   They can stimulate microbial activity and growth, which may in turn lower the pH and the redox
266   potential (Mabuza et al., 2020). These various examples show that the factors driving the
267   destabilisation of MAOM are multiple and may be dependent on each other. The factors directly
268   controlled by plant functioning are referred to as "direct (non-microbial) pathways". Conversely, in
269   "indirect (microbial) pathways" root exudates mobilise MAOM indirectly through a sugar-mediated
270   pathway dependent on microbial intermediaries (Keiluweit et al., 2015; Jilling 2021; Finley et al., 2018;
271   Li et al., 2021). Whether the pathway is direct or indirect, whether it is located in the rhizosphere or in
272   the residuesphere, it appears that the functioning of the biotic compartment (roots, bacteria and fungi)
273   is a source of low-molecular-mass organic acids, and pH and Eh variations, all of which are likely to
274   modify the bonding equilibrium of the organo-mineral associations and can be accompanied by an
275   abiotically mediated PE.
276

The conceptual mechanisms of abiotically mediated PE include three types of reactions releasing mineral-bound C and making the compounds accessible to the extracellular enzymes of microorganisms (Figure 2):

- *Mineral dissolution* can be promoted by exudates. Ligand-promoted dissolution is accelerated by strong ligands such as organic acids (Wang et al., 2014; Li et al., 2018). Protonation of a mineral surface can also weaken metal-oxygen bonds and dissolve minerals (Stumm, 1997). Reductive dissolution can be driven by organic reducing agents such as phenols and aromatic acids (Vermeire et al., 2019). However, as Al is not very sensitive to redox variation, these redox processes mainly impact Fe-bearing mineral phases.

- *Replacement* is promoted by strong ligands such as organic acids, which sorb onto mineral surfaces and displace or release previously sorbed organic compounds from minerals (Li et al., 2021). Other metabolites are also implied in the release of MAOM through displacement of weakly bound compounds (Leineman et al., 2018; Mikutta et al., 2019; Jiang et al., 2021).

- *Direct depolymerisation* of mineral-bound organic compounds by extracellular enzymes can also occur on weakly bound OM, bypassing the desorption step (Wang et al., 2020).

4   ROLE OF ENZYMES IN PE

    4.1   Enzymes involved in priming effect mechanisms

The type of enzyme involved in PE depends on (i) the chemical composition of the added substrate, (ii) the chemical composition of the native SOM, (iii) the nutrient availability, and (iv) the functional requirements of microorganisms, which differ amongst microbial communities. When the environment is not nutrient limited, FOM input is expected to increase the activity and biomass of high-CUE microbial communities presenting a high growth yield due to an increased activity of the FOM depolymerizing enzymes (Fanin et al., 2014; Fanin & Bertrand, 2016; Malik et al., 2019). For example, additions of plant residues stimulate plant cell-wall-degrading hydrolytic enzymes, and the resulting PE is directly related to enhanced degradation of plant-derived residues of the native SOM (Chen et al., 2014). Increased activity of microbial communities can also enhance the degradation of native SOM as a result of co-metabolism, *i.e.*, hydrolase activity yields enough energy and monomers to produce a panel of other enzymes degrading FOM, or/and the non-complex and accessible SOM (Blagodatskaya & Kuzyakov, 2008; Pascault et al., 2013).

When the environment is nutrient limited, microbial communities shift from high to low CUE, notably because nutrient limitation affects the community metabolism and reduces growth yield, while it favours investment in resource acquisition (Malik et al., 2020; Fanin et al., 2020). In this context, another type of co-metabolism directed to N-containing SOM is induced when N is not available or

becomes exhausted at the late stages of decomposition. It is characterised by a high expression level of enzymes degrading N-containing compounds (Moorhead & Sinsabaugh, 2006; Chen et al., 2014), including hydrolytic chitinases and oxidases, which release N associated to complex aromatic compounds (Derrien et al., 2014). The resulting PE is directly related to enhanced degradation of N-rich SOM compounds that differ chemically from those of the added substrate.

4.2 Energy cost and return on investment

The activity of enzymes involved in PE implies benefits and costs. The major benefit from enzyme activity is the release of mineral nutrients and organic monomers and oligomers that can be taken up and assimilated by microbes. On the other hand, the costs of enzyme production include the C and nutrients necessary to build the backbone structure of the enzyme itself, but also the metabolic energy required for protein synthesis and enzyme excretion (Allison et al. 2014). Because extracellular enzymes have a short lifespan, a return on investment is necessary to maintain microbial growth, especially when resources are scarce in the environment. Nevertheless, the impact of the energy balance associated to enzymes involved in PE is still unknown.

A critical factor influencing the energy balance related to enzymes is the distance between the substrate and the enzyme producer. While some enzymes are directly attached to the plasma membrane of microbial cells or grouped into multi-enzyme complexes (named 'membrane enzymes' hereafter), like the cellulosomes dedicated to cellulose degradation, other enzymes are excreted far away from the cells (named 'exoenzymes' hereafter). When microbes are relatively near their substrates, the monomers produced by membrane enzymes are directly recognised by cell-wall receptors and transported into microbial cells, and this optimises resource uptake from the soil solution. The benefits of enzyme production often exceed the costs thanks to a high return on investment, and allow for microbial growth (Schimel & Weintraub, 2003). By contrast, exoenzymes are sent far away from microbial cells, up to a few mm (Gaillard et al., 1999; Poll et al., 2006; Védère et al., 2020). A significant portion of labile C released through enzymatic action may be lost to the enzyme producer (*i.e.*, leached to deeper soil horizons, immobilised in soil aggregates and/or taken up by other microorganisms) (Allison, 2005; Burns et al., 2013). As a consequence, microbes may invest in chemical defenses (i.e., antibiotics) or produce more enzymes to optimize enzyme foraging (Hiscox & Boddy, 2017).

For these reasons, the degradation of substrates by membrane enzymes may potentially have a stronger positive feedback on SOM decomposition and biomass growth (and thus on PE) than exoenzymes. Therefore, we propose that the distance between microbes and their substrates as well

as substrate accessibility to enzymes should be considered as primary factors driving the magnitude of PE across various soil types. This could be tested by implementing random mutagenesis, gene complementation and functional assays (as performed by Picard et al., 2021) to identify how any mutation affecting the production of membrane-cellulosomes or exo-cellulases of a microbe community inoculated on a sterile soil impacts SOM decomposition and microbial growth after the addition of cellulose. Batch experiments with spatially localized applications of decomposers and of fresh cellulose on a sterile soil (as performed by Pinhero et al., 2015) could also allow the investigation of the effect of varying distances between substrate and decomposers on PE intensity.

4.3   PE induced by enzymes sorbed to minerals

Exoenzymes circulating in soil pores exhibit a strong affinity for mineral surfaces and can sorb onto them (Olagoke et al., 2019; Olagoke et al., 2020). This mechanism protects them from degradation and accounts for their persistence in the soil. Mineral-associated exoenzymes are suspected to contribute to the depolymerisation of freshly added substrate, even if their activity is diminished due to changes in conformation after binding to mineral surfaces. They are beneficial to the cost:benefit ratio of the PE because the released monomers are cost-free for the microorganisms that assimilate them.

The monomers taken up into microbe cells are either directed towards microbial biosynthesis, or oxidised into CO_2 by endoenzymes, as reported by CUE. Under aerobic conditions, the aerobic oxidation of monomers also generates energy in the form of adenosine triphosphate. Endoenzymes may be involved in a PE when they are released from dead microbial cells, attached to mineral surfaces, and maintained their activity outside the cell, with electron transfer to O_2 potentially ensured by soil particles due to their electric charge (Maire et al., 2013; Trevors et al., 1982).

5   MICROBIAL ACTORS OF PE

5.1   Actors proposed in the literature

Using molecular approaches, various studies have linked certain PE generation mechanisms to specific bacterial or fungal taxonomic groups listed in Table 2.

Bacterial taxa have been much better documented than fungal ones, probably because of methodological limitations. While a hyphal section is at the microbial scale, its prospected soil volume exceeds the cubic meter. Therefore, current molecular techniques are better adapted to single cell microbial populations (Ranjard et al., 2003). Yet, saprotrophic fungi are believed to have a preponderant role in PE generation, especially in the N-mining process. Their hyphal physiognomy allows them to co-metabolise FOM and SOM very easily, and they appear to be more competitive than bacteria for cellulose decomposition and in turn use cellulose C as an energy source for N-mining. In

382   addition, some fungi have strong enzymatic capacities to break down recalcitrant compounds and

383   access trapped N (Fontaine et al., 2011).

384

385   Kingdom or phylum levels do not seem appropriate to associate taxa to specific PE generation

386   mechanisms. Bacterial and fungal taxa have been assigned to all the different PE generation

387   mechanisms (Table 2). Bacteria are roughly considered as r-strategists and fungi as K-strategists, but

388   the reality is more complex (Ho et al., 2017; Tardy et al., 2015). The genus level shows a better

389   correspondence except for *Bacillus*, which gathers both true oligotrophic (*i.e.*, K-strategist) and

390   copiotrophic (*i.e.*, r-strategist) species (Morrissey et al., 2017).

391

392   Some interpretations have suffered from a lack of knowledge on poorly described phyla like the

393   Gemmatimonadetes, classified as oligotrophic by some authors (Pascault et al., 2013) but copiotrophic

394   by others (Yu et al., 2020). Increasing efforts in culturomics (Sarhana et al. 2019) and also in large-scale

395   studies of soil microbial diversity will provide precious information to confer functional attributes to

396   microbial taxa (Karimi et al., 2018).

397

398   The r-to-K or oligotroph-to-copiotroph concepts refer to a continuum, and many central taxa can

399   develop a certain physiological flexibility depending on (i) environmental conditions, and (ii)

400   competition or complementarity with the other taxa present in each assemblage. These taxa can

401   therefore participate to different PE generation mechanisms according to local conditions (Bernard et

402   al., 2009).

403

404   5.2.   A trait-based approach

405   Proposing taxa associated with the different PE generation mechanisms requires to directly focus on

406   functional traits linked to the different PE generation mechanisms described so far. Data retrieved

407   from the application of -omics techniques to environmental samples and recent developments in

408   culturomics have provided and will continue to provide precious information about the genes and

409   enzymes associated with particular taxa.

410

411   Different bacterial genomic traits have been proposed to be related to maximum growth rates, and

412   therefore to maximal CUE (linked to stoichiometric decomposition PE), like a high copy number of the

413   ribosomal operon (Klappenbach et al., 2000), a small genome size (Saifuddin et al., 2019), vicinity of

414   the ribosomal operon and the chromosomal origin of replication (Viera-Silva & Rocha, 2010), and the

415   translation efficiency, (Weissman et al., 2021). Some genera sharing those traits are *Burkholderia* sp.,

416   *Flavobacterium* sp. and some *Bacillus* sp. (Fierer et al., 2017). The highest maximal fungal growth rates

are shared by some well-known saprotrophic fungi that solely use simple sugars and produce spores (Ho et al., 2017), like *Acremonium, Alternaria, Aureobasidium, Cladosporium, Cephalotrichum, Mucor, Rhizopus,* and *Thysanophora*. Decomposition of polymers like cellulose and chitin is widespread and shared by copiotrophic bacterial and fungal species. Genes associated to simple sugar metabolism and ß-glucosidase activity were recently shown to drive PE in forest soils (Ren et al., 2021).

Obligate low-CUE bacteria should be characterised by opposite genomic traits, as adapted to oligotrophy by adopting a small size, a high adsorption capacity of substrate monomers. They should not produce depolymerising enzymes because maintaining these genes is costly (Lynch & Marinov 2015). Therefore, key species implied in nutrient-mining PE should first share a high diversity of genes coding for acquisition systems of poorly accessible substrates. White- and brown-rot saprotrophic fungi and certain bacterial species (*Nocardia* sp., *Streptomyces viridosporum* and other sp., *Amycolatopsis* sp., *Xanthomonas* sp., *Rhodococcus jostii., Novosphingobium* sp. and *Sphingomonas* sp.) share various peroxidases and laccases (Brown & Chang, 2014). All these bacterial genera have between 2 and 6 copies of the ribosomal operon in their genomes (https://rrndb.umms.med.umich.edu), suggesting that they can be physiologically flexible and adapt their CUE to the quality of the substrates they are decomposing. This is also the case of bacterial and fungal species able to weather minerals by producing ligands or chelating agents and who might be good candidates for the abiotically mediated PE, like *Arthrobacter, Janthinobacterium, Leifsonia, Polaromonas* and *Burkholderia* (see Uroz et al., 2015 for a review). As phosphorus and organic matter compete for sorption sites on minerals (Spohn et al. 2022), P-solubilising bacteria could also have an effect on SOM sorption. *Bacillus subtilis* can desorb P from gibbsite by mineral adhesion of the cells themselves (Hong et al., 2016).

6   HOW TO INTRODUCE PE INTO MODELS

Priming is difficult to model, like any other observed but not fully understood phenomenon. To develop a mechanistic model, equations describing the phenomenon are needed, and the very first step is obviously an in-depth understanding of observations. With priming, the situation is somehow slightly uncomfortable for modellers because the debate about the drivers of priming is still ongoing (Chen et al., 2014). Nevertheless, the ubiquity of priming and its potential importance in the soil C balance in a climate change context (Terrer et al., 2021) urge modellers to include priming in their models. We can basically split the modelling objective related to priming into two categories, *i.e.*, (i) understanding the phenomenon, and (ii) quantifying it because observations have shown that priming plays a major role in the C balance of ecosystems. Both objectives can be considered as

452   independent or linked depending on the authors (Blagodatsky et al., 2010; Guenet et al., 2018; Neill

453   & Gignoux, 2006; Perveen et al., 2014). These categories are often associated with different spatio-

454   temporal scales (Figure 3): most of the models aim at understanding priming at small scales and over

455   short time periods, whereas models aimed at quantifying the importance of priming on the SOM

456   balance are developed at larger scales and over longer time periods.

457

458   Priming being defined as a change in the SOM mineralisation rate induced by FOM additions

459   (Kuzyakov et al., 2000), we can consider that classic models describing SOM mineralisation using first-

460   order kinetics – *e.g.*, CENTURY or RothC (Coleman et al., 1997; Parton et al., 1987) – implicitly include

461   priming in their decomposition parameter values. Nevertheless, this is associated with implicit

462   assumptions that the relative importance of priming is fixed in space and over time. In particular, an

463   important implicit assumption of first-order kinetic models is that the microbial community is stable.

464   This hypothesis is not verified in a context of global change (climate, CO_2 + land use) (Carney et al.,

465   2007), and priming will play a major role on future C balances (Terrer et al., 2021). Therefore, it is

466   necessary to understand and quantify priming to represent it in models instead of just considering it

467   implicitly.

468

469   To represent priming following the classic definition, the FOM and SOM compartments must interact

470   directly or indirectly. This can be done simply by implementing SOM rate modifiers depending on

471   FOM (Guenet et al., 2013) or by changing the SOM CUE according to FOM availability without

472   modifying the decomposition rate (Zhang et al., 2018). Priming can also be represented by models

473   that consider explicit microbial mechanisms, with the effect of FOM on SOM considered through

474   microbial biomass (Fontaine & Barot, 2005; Huang et al., 2018; Perveen et al., 2014; Wieder et al.,

475   2015). These different views on priming representation in models should not be seen as

476   contradictory or opposed, but complementary. It would be unrealistic to consider that a complete

477   representation of priming will soon be included in large-scale models (Eyring et al., 2016); simple and

478   semi-empirical approaches are needed. However, it would also be quite unrealistic to believe that

479   semi-empirical approaches will help to develop a deep understanding of priming. In our opinion, the

480   main difficulty in both cases is related to the parameters used in the equation. Increasing model

481   complexity is most often associated with an increased number of parameters and reduced linear

482   responses of the models (Huang et al., 2021). Therefore, the more complex a model is, the higher

483   the number of possible model trajectories. To avoid unrealistic model behaviors, it is of prime

484   importance to constrain the model parameters using observations. Complex models associated with

485   short time periods and small scales can be constrained by observations coming from incubations

486   under controlled conditions (Neill & Guenet, 2010). Nevertheless, applying complex models and

487   finding suitable parameter values for different soil types, climates, plant covers, etc. is an important
488   challenge that requires a large dataset of incubations like those recently published by Bastida et al.
489   (2019). Even though such large datasets of incubations are now published in single studies or in meta-
490   analyses (Luo et al., 2016), laboratory incubations are still carried out in controlled environments,
491   and upscaling the parameters obtained at this scale to the ecosystem or even the global level is not
492   straightforward. At the end of the calibration process, upscaling may increase model prediction
493   uncertainties compared to less mechanistic models because the uncertainties associated with
494   parameter estimation and upscaling are large. Semi-empirical approaches with a limited number of
495   parameters are an intermediate approach that make it possible to add priming explicitly and to study
496   its effect at larger scales (Guenet et al., 2016; Perveen et al., 2014; Sulman et al., 2014). These models
497   do not catch the entire complexity of the priming response, but they are robust enough to estimate
498   the importance of priming on the SOM balance at large scales.

499

500   Different models may involve different, equally accurate or equally uncertain approaches. Therefore,
501   uncertainty remains concerning our prediction capacities. To solve this problem, a promising
502   approach consists in working in close collaboration with experimenters to set up trials aimed at
503   better understanding priming, developing the associated equations and estimating the parameters.
504   But current models can also be compared in the short term. This approach has been fruitful in many
505   fields such as climate sciences (IPCC 2013), biogeochemistry applied to large scales (Ito et al., 2020),
506   and SOM dynamics (Farina et al., 2021). Therefore, using model intercomparison exercises is an
507   interesting alternative to provide robust estimations. This is particularly important when such
508   estimations are key information for stakeholders and policy makers and robust estimations by soil
509   scientists are expected in the near future (Shukla et al., 2019).

510

511   Another difficulty in including priming into large-scale models representing plant growth and
512   phenology, water cycles and energy fluxes (Cramer et al., 2001; Sitch et al., 2003) is related to the
513   feedbacks between all the model (Goll et al., 2014; Zaehle et al., 2010; Zhu et al., 2019). A promising
514   research avenue aiming to incorporate complex models with reduced feedbacks but not yet
515   implemented in soil science to our knowledge is the use of a model emulator (Fer et al., 2018). Such
516   approach can help large-scale models to consider priming based on a complex mechanistic approach.
517   Representing priming in models is still a challenge with a large diversity of approaches that should
518   be considered as complementary and able to tackle highly different situations, from a deep
519   understanding of priming at small scales to the quantification of the importance of priming in the
520   global C budget.

521

522   7   CONSEQUENCES OF PE ON ECOSYSTEM SERVICES UNDER GLOBAL CHANGE CONDITIONS
523

524   Due to various technical constraints related to the measurement of PE such as isotopic labelling of
525   contributed OM, PE studies are often conducted in the laboratory in soil incubations with pure
526   chemical compounds (e.g., glucose) making any extrapolation to the ecosystem scale very
527   complicated. However, some studies have quantified PE and its impacts on C and N fluxes in soil-plant
528   systems (see review of Cheng et al., 2014), often involving heavy infrastructure in order to guarantee
529   the reliability of measurements (Cros et al., 2019). These studies show a massive change in SOM
530   decomposition rate in their rhizosphere induced by plants, from a 50% reduction to a 280% increase
531   (Cheng et al., 2014; Huo et al., 2017). This plant induced priming effect (PIPE) results from the
532   combination of several rhizosphere processes such as plant C inputs (litter rhizodepositions) as well as
533   root uptake of nutrients and water intensifying drying-rewetting cycles and breakage of soil aggregates
534   (Dijkstra et al., 2021). Most if not all ecosystem services and disservices related to SOM dynamics are
535   impacted by the PIPE. For example, the PIPE has been suggested to increase disservices such as soil C
536   loss (Carney et al., 2007; Henneron et al., 2020a) and nitrate leaching (Liu et al., 2008), as well as
537   services such as soluble nutrient supply to plants sustaining greater plant production and CO_2 fixation
538   (Henneron et al., 2020a; 2020b).

539

540   Understanding the response of the PIPE and SOM dynamics to current global environmental changes
541   (atmospheric CO_2 increase, global warming and land use changes) is crucial to predict the long-term
542   evolution of the related (dis)services. The task appears very complex because global change can act on
543   priming processes in multiple ways. For example, global warming can impact the intensity of PIPE
544   through its effects on soil enzyme activity, soil microbial populations (Fanin *et al*. in this Special Issue)
545   and plant communities, and in turn determine the quantity and quality of plant C inputs into soils. This
546   complexity is increased by the fact that other processes than PIPE control SOM dynamics (*e.g.*, SOM
547   formation) and have their own response to global change. We suggest a framework that reduces this
548   complexity to three scenarios describing the effect of GC on PIPE, SOM dynamics and four ecosystem
549   (dis)services (Figure 4). This framework is based on the assumptions that (i) plants can drive SOM
550   dynamics to meet their soluble nutrient needs, and (ii) global change can degrade this effect.

551

552   It is increasingly recognised that plants can modulate the PIPE and SOM dynamics to adjust the
553   microbial supply of soluble nutrients to their demand. In a common garden experiment comparing 12
554   grassland plant species, SOM mineralisation rates have been found to be adjusted to the demand of
555   each species (Henneron et al., 2020b). Enhanced plant photosynthesis and nutrient demand under
556   elevated CO_2 increase the PIPE, soil organic N mineralisation and plant N uptake in many different

16

557   contexts (de Graaff et al., 2009; Kuzyakov, 2019). In contrast, decreased plant photosynthesis (plant

558   demand) in response to plant shading/cutting induces reduced SOM mineralisation (Shahzad et al.,

559   2012; Tang et al., 2019). In temperate ecosystems, the mineralisation/immobilisation ratio changes

560   throughout the season in line with plant demands; immobilisation dominates in winter (low demand),

561   whereas mineralisation dominates in spring (high demand) (Bardgett et al., 2005; Yokobe et al., 2018).

562   Plants control the storage and release of C and nutrients from SOM through their C inputs and soluble

563   nutrient uptake, and this alters the activity of high-CUE and low-CUE microbes (Figure 5). Based on this

564   knowledge on plant control of SOM dynamics, our framework suggests three scenarios of ecosystem

565   response to global change. These scenarios should be considered as attempts to summarise current

566   knowledge, to make it easier to include PE in models and identify priorities for further investigations

567   on the effects of global change on ecosystem C and nutrient cycles.

568

569   ***Scenario 1: Control by plants is maintained – Global change affects plant production***. The effects of

570   GC on plant-soil systems are moderate in that plants keep their ability to adjust SOM dynamics to their

571   nutrient needs. Therefore, the ecosystem response to global change is determined by the plant

572   response to global change. More specifically, plant production is modified by global change, while

573   plant biomass remains unchanged. For example, in grasslands under elevated CO_2, primary production

574   is strongly stimulated but the amounts of C and nutrients stored in plant biomass barely change (Terrer

575   et al., 2021) as a consequence of the low storage capacity of herbaceous plants (compared to trees)

576   and continuous grazing. Our framework suggests that the change in primary production modifies the

577   turnover rate of SOM, not the SOM stock. The higher (or lower) plant nutrient demand is expected to

578   increase (or decrease) PIPE and SOM mineralisation. However, an equivalent amount of organic

579   nutrients is expected to return to the soil through litter fall and rhizodeposition if no biomass is

580   exported; this replenishes the SOM stock during periods of low plant demand (Figure 5). In sum, the

581   soil C and ecosystem C stocks as well as nutrient losses remain unchanged, while plant production is

582   changed, meaning there is no feedback effect on global change.

583

584   ***Scenario 2: Control by plants is maintained – Global change affects standing biomass.*** The sole

585   difference with scenario 1 is that global change also affects the standing plant biomass, as in forests

586   submitted to elevated CO_2 or moderate warming (+ 1 or 2 °C) where primary production and tree

587   biomass often increase (Melillo et al., 2002; Terrer et al., 2021). The increase or decrease in plant

588   biomass induces a parallel and opposite change in the SOM stock (a decrease or an increase,

589   respectively). Given that plants drive the PIPE and SOM dynamics as a function of their nutrient

590   demand, SOM behaves as a nutrient source or sink according to whether plant biomass accumulates

591   or loses nutrients. This nutrient redistribution between plant biomass and SOM affects the ecosystem

592 C stock since these two compartments have different carbon:nutrient ratios (*e.g*., redistribution of
593 nutrients towards plant biomass increases ecosystem C). In sum, plant production, the soil C stock and
594 the ecosystem are modified, while nutrient loss remains unchanged. In other words, an increase in
595 plant biomass leads to negative feedback and therefore attenuation of climate change, whereas a
596 decrease in plant biomass leads to a runaway positive feedback on climate change.

597

598 ***Scenario 3: Global change degrades plant control of SOM dynamics.*** This scenario considers more
599 extreme impacts of global change on plant-soil systems leading to a partial or total degradation of the
600 control of SOM dynamics by plants. For example, extreme climatic events such as droughts or
601 heatwaves can block plant nutrient uptake while soil microbes continue to mineralise SOM and release
602 soluble N that accumulates in soils (Deng et al., 2021). Degradation of plant control of the PIPE and
603 SOM dynamics leads to periods of excess soluble nutrients on the short term; these excess nutrients
604 increase nutrient loss and environmental pollution, and periods of nutrient deficiency limit plant
605 production. On the long term, the increased nutrient loss contributes to progressive soil
606 impoverishment and ecosystem decline (loss of plant biomass and production, lower ecosystem C
607 stock). This scenario leads to a runaway positive feedback on global change (climate change and
608 disruption of nutrient cycles)

609

610 Several experimental studies support our scenarios. The higher plant C input to soil in response to
611 elevated CO_2 intensifies PIPE and SOM turnover (Scenario 1) (Kuzyakov, 2019; Perveen et al., 2014).
612 Moreover, a recent meta-analysis (Terrer et al., 2021) showed that the effect of elevated CO_2 on SOM
613 stocks was best explained by a negative relationship with plant biomass (Scenario 2): when plant
614 biomass is strongly stimulated by elevated CO_2, SOM storage declines; conversely, when biomass
615 decreases, SOM storage increases. Although plant-soil interactions play a major role in the response
616 of the PIPE to global change, it is important to note that some factors of global change may act on the
617 PIPE and related services independently of soil-plant interactions. For example, moderate atmospheric
618 N depositions have been shown to decrease the microbial production of oxidative enzymes (and the
619 intensity of PIPE leading to SOM accumulation (Hagedorn et al., 2003; Ramirez et al., 2012; Perveen et
620 al., 2014; Liu et al., 2018).
621 In sum, we propose that global change can deeply impact PIPE, with consequences ranging from a
622 change in plant production to the progressive decline of ecosystems and their services. The
623 consequences of global change on ecosystem services would mostly depend on the ability of plants to
624 keep driving the PIPE and SOM dynamics in a global change context.

625

626 8   CONCLUSION AND PERSPECTIVES

627

After more than 70 years of research on PE, the way it will respond to climate change and impact the flux of C to the atmosphere is still unclear. Research has been strongly intensified these last ten years but a strong global international coordination is still lacking. PE are generated by multiple mechanisms, but many studies continue to focus on the global flux without considering the underlying processes. Various processes have been proposed throughout the literature, following context-dependant nomenclature, thus introducing some confusion in subsequent work. Moreover, while the environmental drivers of these mechanisms are roughly known, their respective output on the C cycle merit to be more deeply studied. We have proposed a classification of the different mechanisms according to their putative impact on the soil C balance between sequestration and mineralization. This classification is based on two factors which have to be taken into account: the putative turnover rate of the primed SOM pool and the putative CUE of the microbial populations respiring the primed C. Future studies have to verify these assumptions as some SOM pools, especially the MAOM, are likely characterised by a wider range of turnover rates than previously thought, depending on the organo-mineral association model and the chemical bonds involved. While wide metagenomic and culturomic studies will continue to increase our knowledge on functional attributes of poorly described microbial populations, the CUE of populations respiring the primed C cannot be extrapolated from the sole identity of present taxa. Actually, many key actors are probably characterized by a wide physiological flexibility, adjusting their CUE and driving different PE generation mechanisms following environmental conditions. Increasing our knowledge on C fluxes driven by the different PE mechanisms will help to introduce the PE into models of SOM dynamics, thus predicting its behavior in response to climate change. Developing *in situ* experiments should also represent a decisive step forward to complete actual knowledge mostly based on incubation experiments. But we are convinced that the PE will not respond directly, but likely through plants, which are the main producers of fresh organic matter generating soil PE. Therefore, the concomitant elevated CO_2, and plant nutrient requirements, also have to be taken into account.

655

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1196

1197 SUPPORTING INFORMATION

1198 Appendix S1: History of priming effect exploration.

1199 Table S1: List of PE publications from 1973 to April 2021

1200 Figure S1: Network of the international scientific community working on the soil priming effect

1201 Figure S2: Main topics tackled in articles focused on soil priming effect, as deduced from network
1202   analysis

1203

1204 TABLES AND FIGURES

1205 TABLE 1   List of the main PE generation mechanisms proposed by the literature, with their given
1206 name, definition, drivers and reference articles. SOM: Soil Organic Matter; FOM: Fresh Organic Matter.
1207 For each mechanism, the last two columns indicate whether microbial populations assimilating the
1208 primed carbon are characterized by high or low carbon use efficiency (CUE), assuming that r-strategists
1209 (or copiotrophs) have high CUE while K-strategists (or oligotrophs) have low CUE, and whether the
1210 primed SOM pool has a fast, medium or low turnover rate. Generation mechanisms have been colored
1211 to follow a proposed gradient of implication in the fate of soil C: from C gain (brown) to C loss (blue).
1212

1213 TABLE 2   List of bacterial and fungal taxonomic groups identified in the literature as associated
1214 to FOM mineralization and/or to one of the main PE generation mechanisms: negative PE or positive
1215 PE (supposed to involve high CUE or low CUE actors). PE generation mechanisms have been colored to
1216 follow a proposed gradient of implication in the fate of soil C: from C gain (brown) to C loss (blue). FOM
1217 mineralization was not considered in this gradient and was colored in green. In the colored cells, names
1218 correspond to the specific FOM that generated PE in the different publications.

1254   release of soluble nutrients by the low-CUE microbial community that is stimulated by plant carbon

1255   inputs (plant-induced priming effect, PIPE). Plant C inputs also supply the high-CUE microbial

1256   community that builds SOM (storing nutrients and carbon) when plant nutrient demand is low or

1257   inexistent. The relative abundance and activity of low-CUE and high-CUE communities is driven by

1258   plant uptake of soluble nutrients and carbon inputs (quantity & quality). The numbers indicate the

1259   chronology of events.

1260

Box 1: Shared terminology

FOM: Fresh Organic Matter relates to organic matter entering the soil and which will generate a PE on organic matter already present in soil (SOM). It can derive from plants like rhizodeposits (exudates or mucus or dead apical cells) or dead roots, leaves, crop residues or dead animals.

SOM: Soil Organic Matter gathers all type of dead organic matter present in the soil before the input of new FOM, from small molecular dissolved compounds to plants residues.

MAOM: Mineral Associated Organic Matter designates organic compounds associated to minerals, from small molecules to microbial or plant cell tissues. MAOM is a specific fraction of SOM.

PE: In the present review, the Priming Effect is a change in the rate SOM-C mineralization induced by an input of FOM.

RPE: The Rhizosphere Priming Effect is the PE generated by live roots and associated rhizosphere organisms when compared to SOM-C mineralisation from rootless soils under the same environmental conditions.

PIPE: The plant Induced Priming Effect is the PE induced by whole plants on SOM-C, including RPE but also PE induced by dead roots and aerial parts of plants like litter or crop residues.

Table 1

PE generation mechanisms			definition	Drivers	References	Microbial CUE	SOM pool turnover rate
SOM decomposition deceleration (negative PE)	real	preferential substrate utilisation	Microbial populations decomposing SOM switch to FOM decomposition.	Labile part of added substrate > 2x microbial biomass Nutrient concentration (soil + substrate) should meet microbial demand	Cheng 1999; Kuzyakov 2000	High and low	?
		Necromass recycling	Microbial populations decomposing SOM switch to the decomposition of the necromass of FOM-decomposers.	Microbial necromass can also be sorbed on minerals. Clays and metals in the soil can compete with microbes for available necromass (Creamer et al. 2019).	Cui et al. 2020		
SOM decomposition acceleration (positive PE)	apparent	triggering effect	Acceleration of internal microbial metabolism by trace amounts of substrate with an immediate (several minutes to several hours) increase in the respiratory activity	labile part of added substrate < microbial biomass	Blagodatskaya & Kuzyakov 2008	High and low	Fast
		microbial activation	The N content of the soluble part of the amended substrate will be used to produce first extracellular enzymes. It can also correspond to N-containing root exudates in the rhizosphere.	labile part of added substrate < microbial biomass N rich FOM	Kuzyakov 2002		
		pool substitution	acceleration of the turnover of the whole microbial biomass (several days or weeks)	labile part of added substrate = microbial biomass	Blagodatskaya & Kuzyakov 2008		
	real	stoichiometric decomposition	FOM increases biomass and general enzymatic activity, accompanied by domination of r-strategists and increase in SOM decomposition.	FOM complexity with nutrient availability (Plant residues)	Chen et al. 2014	High	Fast to medium
		cometabolism	Assimilation of FOM and SOM by similar populations. FOM brings the energy to decompose less available SOM compounds.	labile part of added substrate superior to microbial biomass; Nutrient availability (substrate and dissolved) slighly inferior to microbial demand	Kuzyakov 2000; Blagodatskaya & Kuzyakov 2008; Yu et al. 2018	High or low	
		mechanism 1	Enzymes produced by r-strategists in order to decompose FOM may be partially efficient for SOM degradation	FOM complexity with nutrient availability (Plant residues)	Fontaine et al. 2003	Low	
		mechanism 2	K-strategists mineralizing SOM, with slow growth rates, may benefit from FOM polymerised substrates which have a long residence time in soil. Thus K-strategist populations and associated enzyme production increase, causing PE.	labile energy rich substrate with nutrient depletion	Fontaine et al. 2003	Low	Medium
		N-mining	FOM supply increases the biomass and enzymatic activity of K-strategists and increases SOM decomposition.	labile energy rich substrate with N depletion	Chen et al. 2014		
		Abiotically mediated	Plant roots exudate organic acids, which desorb labile MOS associated to minerals (MAOM). Liberated compounds will be accessible to microbial decomposition. Fungi and bacteria are also able to produce organic acids or chelatants and could generate such PE	N-limitation	Keiluweight et al. 2015 ; Fang et al. 2018	High and or low ?	Slow

TABLE 2

phylogenetic affiliation						FOM mineralization	Negative PE	High-CUE PE	Low-CUE PE	References
Kingdom	Gram	Phylum	Order	Family	Genus					
Bacteria	Gramm-						Grass residue	Urea, cellulose		Hamer et al. 2009; Potthast et al. 2010; Blagodatskaya et al. 2014
		G- (16:1w7c)						Glucose		Derrien et al. 2014
		G- (18:1w7c)							Glucose	Derrien et al. 2014
		Proteobacteria				Alfalfa residue				Pascault et al. 2013
		Proteobacteria			Unclassified			Wheat residue		Razanamalala et al. 2018a
						Wheat residue		Glucose, wheat residue		Bernard et al. 2012, Razanamalala et al. 2018b; Fang et al. 2018
		beta-Proteobacteria	Burkholderiales	Comamonadaceae	Variovorax	Wheat residue			Wheat residue	Bernard et al. 2007; 2009
					Ramlibacter			Wheat residues		Bernard et al. 2007
					Acidovorax				Wheat residue	Bernard et al. 2009
					Aquabacterium			Wheat residues	Wheat residue	Bernard et al. 2007; 2009
				Oxalobacreaceae	Massilia	Wheat residue				Bernard et al. 2007; Liu et al. 2021
					Janthinobacterium				Wheat residue	Bernard et al. 2009
					Noviherbaspirillum				Wheat residue	Liu et al. 2021
				Burkholderiaceae	Burkholderia	Glucose; wheat residue				Arcand et al. 2017; Liu et al. 2021
					Ralstonia				Wheat residue	Liu et al. 2021
					Methylibium				Wheat residue	Bernard et al. 2009
		gamma-Proteobacteria				Wheat residue		Wheat residues, glucose		Razanamalala et al. 2018a, 2018b; Maron et al. 2018
			Pseudomonadales	Pseudomonadaceae	Pseudomonas	Wheat residue				Bernard et al. 2007
			Xantomonadales	Rhodanobacteraceae	Dyella				Wheat residue	Liu et al. 2021
		alpha-Proteobacteria				Wheat residue		Wheat residues		Razanamalala et al. 2018a, Maron et al. 2018
						Wheat residue				Bernard et al. 2009
			Sphingomonadales	Sphingomonadaceae	sphingosinicella				Wheat residue	Bernard et al. 2007
					Kaistobacter				Wheat residue	Bernard et al. 2007
					sphingomonas	Glucose				Arcand et al. 2017
			Rhizobiales	Methylobacteriaceae		Wheat residue				Bernard et al. 2009
				Bradyrhizobiaceae	Bradyrhizobium				Wheat residue	Liu et al. 2021
			Hyphomicrobiales	Nitrobacteraceae	Rhodoplanes	Glucose				Arcand et al. 2017
			Micropepsales	Micropepsaceae	Rhizomicrobium				Wheat residue	Liu et al. 2021
		Bacteroidetes					Biochar	Wheat, alfalfa residues, glucose		Bernard et al. 2012; Pascault et al. 2013; Arcand et al. 2017
			Chitinophagales	Chitinophagaceae	Flavisolibacter	Wheat residue				liu et al. 2021
		Acidobacteria					Tree leaves		rice & wheat residue; Glucose; tree leaves	Pascault et al. 2013; Razanamalala et al. 2018a,b; Maron et al. 2018; Fang et al. 2018; Yu et al. 2020
			GP2					Wheat residues		Razanamalala et al. 2018a
			GP6					Wheat residues		Razanamalala et al. 2018a
			GP4						Wheat residue	Razanamalala et al. 2018a
			GP1						Wheat residue	Liu et al. 2021
		Verrucomicrobia				Glucose	Tree leaves	Rice and wheat residue	Wheat residue, tree leaves	Pascault et al. 2013; Morrissey et al. 2017; Razanamalala et al. 2018b; Yu et al. 2020
		Gemmatimonadetes					Biochar		Wheat residue	Pascault et al. 2013; Razanamalala et al. 2018b; Maron et al. 2018; Yu et al. 2020
			Gemmatimonales	Gemmatimonaceae	Gemmatimonas				Wheat residue	Liu et al. 2021
		Planktomycetes						Wheat residues	Wheat residue, Glucose	Razanamalala et al. 2018a,b; Maron et al. 2018
			Planctomycetales	Isosphaeraceae	Aquisphaera				Wheat residue	Liu et al. 2021
		Chloroflexi						Rice and wheat residue	Wheat residue	Razanamalala et al. 2018a,b; Maron et al. 2018
	Gramm+						Bracken		Cellulose, litter, rice straw and roots	Potthast et al. 2010; Blagodatskaya et al. 2014; Wang et al. 2014, 2020
						Wheat residue, Glucose		Glucose	Wheat residue	Razanamalala et al. 2018a,b; Maron et al. 2018; Arcand et al. 2017; Liu et al. 2021; Bernard et al. 2009
						Glucose			Glucose, wheat residue	Garcia-Pausas et al. 2011; Bernard et al. 2012; Morissey et al. 2017
		Actinobacteria	Actinomycetales	micrococcaceae	Arthrobacter	Glucose				Mau et al. 2015
					Sinomonas	Wheat residue				Liu et al. 2021
				Streptomycetaceae		Glucose				Morrissey et al. 2017
						Glucose				Arcand et al. 2017
					Streptomyces	Wheat residue				Liu et al. 2021
			Micromonosporales	micromonosporaceae		Glucose				Morrissey et al. 2017
			Propionibacteriales	Nocardioidaceae		Glucose				Morrissey et al. 2017
			Hyphomicrobiales	Hyphomicrobiaceae		Wheat residue				Bernard et al. 2009
			Acidimicrobideae						Wheat residue	Bernard et al. 2007
			Gaiellales	Gaiellaceae	Gaiella				Wheat residue	liu et al. 2021
		Firmicutes				Wheat residue		Glucose, Alfalfa, rice and wheat residues		Pascault et al. 2013; Razanamalala et al. 2018a,b; Maron et al. 2018; Fang et al. 2018; Yu et al. 2020
			Bacillales	Bacillaceae	Bacillus	Wheat residue			Glucose	Mau et al. 2015; Arcand et al. 2017; Liu et al. 2021
				Paenibacillaceae	Paenibacillus				Glucose	Arcand et al. 2017
				Alicyclobacillaceae	Alicyclobacillus				Wheat residue	Liu et al. 2021
		Unknown							Wheat residue	Razanamalala et al. 2018a
fungi	saprotrophic						Tree leaves	Urea, cellulose, bracken	Glucose, cellulose, litter, tree leaves, Rhizosphere PE	Hamer et al. 2009; Potthast et al. 2010; Garcia-Pausas et al. 2011; Blagodatskaya et al. 2014; Wang et al. 2014; Shahzad et al. 2018; Tian et al. 2019; Fang et al. 2018; Yu et al. 2020
	saprotrophic	Leothiomycota						Cellulose		Blagodatskaya et al. 2014
		Ascomycota							Wheat residue	Maron et al. 2018
			Pezizomycotina	Sordariomycetes					Biochar	Yu et al. 2018
		Basidiomycota				Wheat residue				Maron et al. 2018
			Agaricomycotina	Tremellomycetes					Biochar	Yu et al. 2018
		Zygomycota					Biochar			Yu et al. 2018

FIGURE 2

CO$_2$

Root / microorganism

N

Microorganism

Inputs of root and microbial exudates

H$^+$
e$^-$

MAOM : mineral associated organic matter
low accessibility to microorganisms' enzymes

ES

IS

Adsorbed organic molecules
IS : internal sphere
ES : external sphere

Coprecipitated molecules

Complexed or bridged molecules

OM dissociated from minerals
high accessibility to microorganism enzymes

① ② ③

① **Mineral dissolution**
② **Replacement**
③ **Enzymatic hydrolysis without desorption**

Low-molecular-weight organic acids

SOM molecule (N-rich)

Extracellular enzyme

SOM monomer < 600 Da

N Soluble N

Mineral

Short range order mineral

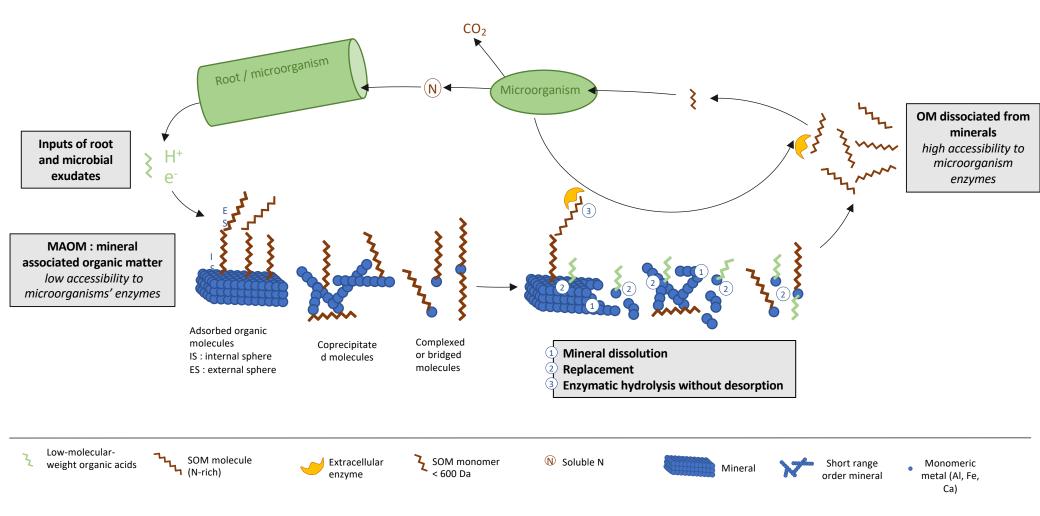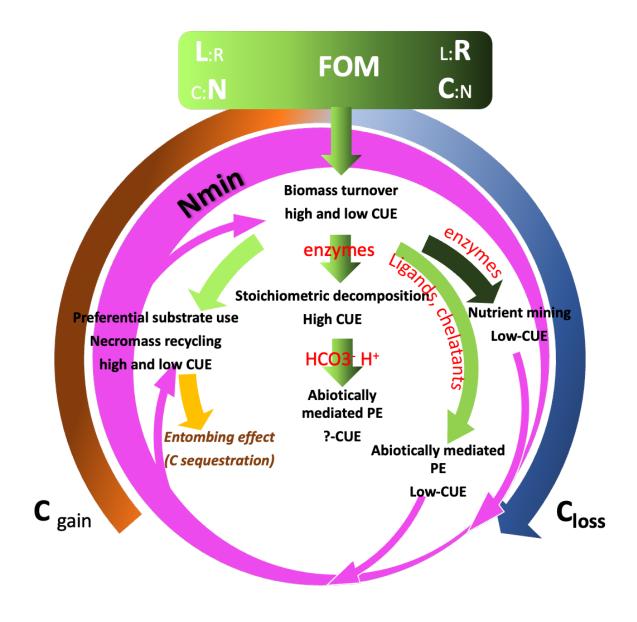Monomeric metal (Al, Fe, Ca)

FIGURE 1

FIGURE 3

Aggregates — **Spatial Scale** — Globe

Days — **Time Scale** — Centuries

Models with no name
(Neill and Gignoux, 2006;
Blagodatsky et al., 2010)

SYMPHONY
(Perveen et al., 2014)
SOMKO
(Gignoux et al., 2001)

ORCHIDEE-PRIM
(Guenet et al., 2018)
MIMICS
(Wieder et al., 2013)

Why to develop a priming model?

Predict the impact on C balance

Increase our
Understanding / test assumptions

What are the constraints to develop a priming model?

Parameterization difficulties / Data to evaluate
impacted by other factors

Maximum model complexity /
detailed data needed for evaluation

FIGURE 4