



Changes in the Size of the Active Microbial Pool Explain Short-Term Soil Respiratory Responses to Temperature and Moisture

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Salazar-Villegas A, Blagodatskaya E and Dukes JS (2016) Changes in the Size of the Active Microbial Pool Explain Short-Term Soil Respiratory Responses to Temperature and Moisture. Front. Microbiol. 7:524. doi: 10.3389/fmicb.2016.00524 Heterotrophic respiration contributes a substantial fraction of the carbon flux from soil to atmosphere, and responds strongly to environmental conditions. However, the mechanisms through which short-term changes in environmental conditions affect microbial respiration still remain unclear. Microorganisms cope with adverse environmental conditions by transitioning into and out of dormancy, a state in which they minimize rates of metabolism and respiration. These transitions are poorly characterized in soil and are generally omitted from decomposition models. Most current approaches to model microbial control over soil CO2 production relate responses to total microbial biomass (TMB) and do not differentiate between microorganisms in active and dormant physiological states. Indeed, few data for active microbial biomass (AMB) exist with which to compare model output. Here, we tested the hypothesis that differences in soil microbial respiration rates across various environmental conditions are more closely related to differences in AMB (e.g., due to activation of dormant microorganisms) than in TMB. We measured basal respiration (SBR) of soil incubated for a week at two temperatures (24 and 33°C) and two moisture levels (10 and 20% soil dry weight [SDW]), and then determined TMB, AMB, microbial specific growth rate, and the lag time before microbial growth (t_{lag}) using the Substrate-Induced Growth Response (SIGR) method. As expected, SBR was more strongly correlated with AMB than with TMB. This relationship indicated that each g active biomass C contributed ~0.04 g CO2-C h^{-1} of SBR. TMB responded very little to short-term changes in temperature and soil moisture and did not explain differences in SBR among the treatments. Maximum specific growth rate did not respond to environmental conditions, suggesting that the dominant microbial populations remained similar. However, warmer temperatures and increased soil moisture both reduced t_{lag} , indicating that favorable abiotic conditions activated soil microorganisms. We conclude that soil respiratory responses to short-term changes in environmental conditions are better explained by changes in AMB than in TMB. These results suggest that decomposition models that explicitly represent microbial carbon

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pools should take into account the active microbial pool, and researchers should be cautious in comparing modeled microbial pool sizes with measurements of TMB.

Keywords: soil respiration, microbial dormancy, microbial biomass, substrate-induced growth response, carbon pool

INTRODUCTION

Microbial respiration responds rapidly to changing environmental conditions, strongly influencing soil carbon cycling, and its feedbacks to climate change (Allison et al., 2010; Frey et al., 2013; Wieder et al., 2013; Sulman et al., 2014). However, soil biogeochemical processes are primarily driven by only a small fraction of soil microbes-those that are physiologically active (Blagodatskaya and Kuzyakov, 2013). In general, more than 80-90% of soil microorganisms are in a dormant or inactive physiological state in which they have minimal respiratory activity (Anderson and Domsch, 1985; Lennon and Jones, 2011). These active and dormant fractions of soil microbial biomass can change in response to environmental and nutritional conditions (Van de Werf and Verstraete, 1987) but typically are not considered when analyzing microbial control over soil CO2 production. If they have represented microbes at all, decomposition models have most commonly represented microbial biomass as a single pool (Zhang et al., 2014; Wieder et al., 2015) without differentiating between its active and dormant fractions. This approach cannot sufficiently represent physiological processes that are important to explain soil respiratory responses to environmental conditions (Wang G. et al., 2014; He et al., 2015). Since active microbes overwhelmingly drive soil carbon processes, we investigated whether active microbial biomass (AMB) is a more accurate predictor of soil CO₂ fluxes than total microbial biomass (TMB).

Dormancy is a common strategy in nature, used by a variety of organisms to cope with adverse environmental conditions (Dworkin and Shah, 2010; Jones and Lennon, 2010; Lennon and Jones, 2011). Although, there are different ways by which soil microorganisms become dormant (e.g., spore formation by Scutellospora castanea or thick-walled structure formation in filaments of Cylindrospermum sp.) (Jones and Lennon, 2010), in all cases there is a strong reduction of physiological activity (Lennon and Jones, 2011). In this state of reduced metabolic activity, microorganisms have almost no influence on biogeochemical processes such as soil CO₂ production. However, dormant microorganisms can be activated when adverse environmental and nutritional conditions become favorable (Jones and Lennon, 2010; Placella et al., 2012; Aanderud et al., 2015). These transitions between active and dormant physiological states may play an important role in large-scale processes such as global carbon cycling (He et al., 2015).

Differentiation between the pools of active and dormant microbial biomass could provide important opportunities to better understand responses of soil CO₂ efflux to environmental factors, e.g., temperature and soil moisture. In regions such

as the Mediterranean, rainfall-induced activation of dormant microorganisms generates soil CO2 pulses that approach the annual net carbon exchange of other terrestrial ecosystems (Xu et al., 2004; Placella et al., 2012). Conversely, when the frequency of rainfall events in a region causes drying-rewetting stress on soil microbial communities, SBR can decrease even if TMB increases (Fierer and Schimel, 2002). This could be a consequence of smaller fractions of physiologically active microorganisms in environmentally stressed soils (Fierer and Schimel, 2002) or of shifts toward microbial communities with higher carbon use efficiency. In addition to soil moisture conditions, microbial respiration, growth, and activity can respond strongly to temperature (Pietikåinen et al., 2005; Steinweg et al., 2012; Suseela et al., 2012; Hagerty et al., 2014). Like soil moisture, warming can increase SBR without affecting TMB (Hagerty et al., 2014), potentially via activation of dormant microorganisms. Thus, TMB does not always respond to environmental changes (Holmes and Zak, 1994; Blagodatskaya et al., 2010) and it remains unclear whether its responses (e.g., TMB decrease under water or nutrient limitation) are proportional to those of AMB.

Current approaches to modeling microbial control over soil CO₂ production mainly consider changes in TMB (Wieder et al., 2015) and essentially ignore changes in the pools of active and dormant microbial biomass (e.g., due to activation of dormant microorganisms; Wang G. et al., 2014). This may be because the current microbial databases available to modelers only represent TMB (Serna-Chavez et al., 2013; Xu et al., 2013) and do not distinguish its active and dormant fractions.

To quantify the importance of active and dormant microbial pools in explaining soil respiratory responses to abiotic factors, we incubated soil at different temperature and moisture levels for a week and subsequently analyzed the correlation of SBR with TMB and AMB. Since, (1) only active microorganisms are able to drive soil biogeochemical processes and (2) abiotic factors can cause microorganism to be activated or inactivated without necessarily altering TMB, we hypothesized that soil respiratory responses to changing environmental conditions would be better explained by changes in AMB than in TMB. We used a kinetic approach based on the Substrate Induced Growth Response (SIGR) technique (Panikov and Sizova, 1996; Blagodatsky et al., 2000; Wutzler et al., 2012) to measure TMB and AMB, as well as other microbial parameters (i.e., microbial specific growth rate and the length of the lag-time before exponential growth starts in response to substrate inputs t_{lag}) that help shed light on the mechanisms by which microbes influence soil CO₂ production.

MATERIALS AND METHODS

Soil Sampling and Preparation

We collected three soil cores separated by 10 m (linear transect) from the top 0-15 cm layer (using a soil core sampler and

Abbreviations: AMB, active microbial biomass; SBR, soil basal respiration; SDW, soil dry weight; SIGR, Substrate-induced growth response; t_{lag} , lag time before microbial growth; TMB, total microbial biomass; μ , microbial specific growth rate.

slide hammer; AMS, Inc.) in a deciduous forest at Purdue University's Ross Biological Reserve, Indiana, USA ($40^{\circ} 24'46''$ N, $87^{\circ} 03'48''$ W), in May 2014. The soil is classified as (2–6% slope) Russell (Alfisol) silt loam (USDA, 2014) and has a pH of 6.97. The mean annual temperature and mean annual precipitation at this site are 11.4° C and 953 mm, respectively (USDA, 2014).

Immediately after sampling, we transported the soil cores to the laboratory (<1 h, at ambient temperatures), prepared a composite sample by sieving the soil through a 2 mm mesh, and adjusted the soil moisture content to 10% (moisture deficit) and 20% (optimal soil moisture) of soil dry weight (SDW). Finally, we placed 25 g of soil (dry weight) in 0.26 L septum-capped glass jars (microcosms) and stored them at 24 and 33°C for 1 week (see Section Experimental Setup below).

Experimental Setup

After soil samples had been incubated for 1 week at 24 and 33°C, (hereafter unheated and heated) and 10 and 20% SDW soil moisture (hereafter dry and wet) conditions, we indirectly measured TMB and AMB (and other microbial parameters) using soil respiratory responses (SIGR) to glucose and mineral nutrient inputs (Panikov and Sizova, 1996). To induce these responses, we homogeneously spread a solution (1 mL per jar) containing 10 mg glucose, 1.90 mg (NH₄) SO₄, 2.25 mg K₂HPO₄, and 3.62 mg MgSO₄ per gram of soil (Blagodatskaya et al., 2014) onto the soil in each jar. The addition of 1 mL solution per sample increased the soil moisture in dry and wet soils from 10 to 20%, during the incubation period, to 14 (still below optimum) and 24% (still within optimal moisture range) SDW during the substrate-induced growth period. We defined the temperature and soil moisture treatments based on ranges of environmental conditions commonly experienced by soil microbial communities in the study area (Goldberg, 2015; ICLIMATE, 2015). Within these ranges, warming and increased soil moisture generally increase soil respiration rates (Li et al., 2008; Yu et al., 2011; Suseela et al., 2012). To create the warming treatment and maintain constant soil moisture we stored the jars in an Environmental Growth Chamber (M18SI) during the incubation period, kept them closed, and confirmed that there were no changes in soil weight (i.e., due to water losses). We measured the CO₂ concentration in each microcosm's headspace every 0.5 for 4 h before adding the solution, and every 0.5 for 19 h (exponential growth phase) afterwards. We measured CO_2 concentration by withdrawing 5 mL of gas from the microcosm headspace using a syringe and injecting it into an infrared gas analyzer (EGM-4, PP Systems, Amesbury, Massachusetts, USA). To avoid negative pressure, we opened the jars at the end of each measurement (i.e., after measuring C_1), aerated the headspace until the CO₂ concentration in the microcosms equilibrated with ambient air, closed the jars, and withdrew the initial gas sample (i.e., C_0) of the following measurement.

We calculated soil respiration rates as in Speratti and Whalen (2008):

$$R_s\left(t\right) = \frac{V(C_1 - C_0)}{Wt}$$

Where $R_s(t)$ is soil respiration rate at time t (in mg C g⁻¹ h⁻¹), V is the volume of the microcosm headspace, C_1 - C_0 is CO₂ concentration change in mg L⁻¹, W is the dry mass of the soil (i.e., 25 g), and t is the time between the first (C_0) and the second (C_1) CO₂ concentration measurements (i.e., 0.5 h).

Kinetic Respiration Analysis

To estimate TMB, AMB, microbial specific growth rate, and t_{lag} , we used the model proposed by Panikov and Sizova (1996) (see also Wutzler et al., 2012):

$$R_{s}(t) = R_{u} + R_{c} * \exp\left(\mu t\right)$$

Where R_u is initial respiration rate uncoupled from growth, R_c is initial respiration rate coupled with growth, and μ is microbial specific growth rate (an intrinsic feature of microbial species). μ is defined as the slope of the growth curve at its inflection point when nutritional and environmental conditions are optimum (Pirt, 1982; Zwietering et al., 1990). At non-optimum conditions, μ is not maximal and reflects environmental constraints of growth rather than any intrinsic feature of the dominant microbial population (Pirt, 1982). It is important to note that soil moisture in the dry treatment (10% SDW) was below optimum (generally 20-40% sdw; Ilstedt et al., 2000) during the period of substrate-induced growth, so μ reflects growth limited by moisture. However, since excessive amounts of substrate were added and homogeneously spread in the soil, the main assumption of exponential growth in an excess of substrate was valid in all treatments. Violating assumptions on optimal growth conditions related to temperature and moisture affect estimation of maximal specific growth rates (μ_{max}). To avoid violating this assumption, only specific growth rates (μ) were calculated for different treatments. AMB was estimated based on substrateinduced exponential growth curves. Differences in exponential curves reflect differences in AMB, caused by the treatments, at the time point immediately before substrate addition.

We fitted the model parameters to measured soil respiration rates (R_s) during the lag and exponential phases that followed substrate amendment. We omitted the first 3 h of measurements from the analysis to exclude the transient effects of mixing and preparing the soil on soil respiration rates (Wutzler et al., 2012).

We estimated TMB as:

$$TMB = \frac{R_c \lambda Y_{CO_2}}{r_0 \mu}$$

Where λ is a basic stoichiometric constant assumed to equal 0.9, which represents the ratio between productive (i.e., respiration that is coupled with ATP generation and cell growth) and total respiration under excess of substrate (Akimenko et al., 1983; Panikov and Sizova, 1996); Y_{CO2} is another constant, the biomass yield per unit of respired CO₂, assumed to equal 1.5 for soil heterotrophs (Payne, 1970; Blagodatsky et al., 2000); and r_0 is the *fraction of AMB*, given by:

$$r_0 = \frac{R_c(1-\lambda)}{R_u + R_c(1-\lambda)}$$



Where the numerator accounts for the maintenance respiration of growing (i.e., active) microbial biomass, and the denominator accounts for total maintenance respiration.

Finally, we estimated AMB as:

$$AMB = TMB * r_0$$

To better understand the mechanisms by which soil microbes control soil carbon dynamics, we also calculated t_{lag} as in Blagodatskaya et al. (2014).

$$t_{lag} = \frac{ln\left(\frac{R_u}{R_c}\right)}{\mu}$$

Note that these estimates of AMB, TMB, and r_0 are for the time period immediately preceding substrate amendment, despite being derived from measurements that follow substrate amendment.

Statistical Analysis and Curve Fitting

We estimated SBR as the mean of the eight soil respiration measurements taken in the 4 h prior to substrate amendment. For the CO_2 evolution curve fitting we used the non-linear least square (nls) function in R (version 3.1.1). We used R^2 as a measure of goodness of fit.

To estimate the significance of the differences in SBR and microbial parameters between treatments, we applied a twoway ANOVA using the aov() function in R (version 3.1.1). When significant differences were found, we conducted posthoc pairwise comparisons using the TukeyHSD function (these results are shown in the Supplementary Material section). To meet assumptions regarding the normality of residuals and homogeneity of variances we log₁₀-transformed AMB for statistical analyses, but for clarity we present the untransformed data in the text and figures. All values are means of three replicates per treatment.

We calculated the percentage of the variance explained by each independent factor by dividing the sum of squares of each factor

(from the ANOVAs mentioned above) by the total sum of squares (examples of calculations based on sums of squares and statistics are in Supplementary Table 1).

RESULTS

As we expected, warmer and wetter conditions both significantly increased SBR (**Figure 1**). The highest SBR occurred in heated, wet soils, whereas the unheated, dry soils had the lowest (**Figure 1**) Soil CO₂ production in heated, wet soils was 2.4 times higher than in unheated, dry soils.

Soil respiration curves showed clear responses to substrate addition, with particularly marked differences in response between heated and unheated treatments (**Figure 2**).

Responses of TMB to warming and soil moisture differed from those of SBR (compare **Figures 1**, **3**). The highest TMB occurred in heated, dry soils whereas the lowest occurred in heated, wet soils (i.e., 19% decrease in TMB due to a soil moisture increase in heated soils; **Figure 3**). Soil moisture level did not significantly affect TMB of unheated soils.

In contrast to TMB, the responses of AMB to temperature and soil moisture were similar to those of SBR (**Figure 4**). As with SBR, warming increased AMB and the greatest AMB occurred in heated, wet soils. Unheated soils had the least AMB. While moisture did not affect AMB in the unheated soils, it increased AMB by 250% in heated soils.

AMB responses to warming and soil moisture were strongly negatively correlated with t_{lag} [$R^2 = 0.936$, $F_{(1, 11)} = 145.51$, P < 0.001; **Figure 4**]. Warming reduced t_{lag} by 43% in dry treatments and by 79% in wet soils. In contrast, μ did not respond to temperature and soil moisture Supplementary Tables 9,10 and was less correlated with AMB [$R^2 = 0.14$, $F_{(1, 11)} = 1.6$, P = 0.235].

As we expected, soil respiration rates across the temperature and soil moisture treatments were strongly correlated with AMB, but not with TMB (**Figure 5**). The correlation between AMB and SBR was positive and indicated an approximate increase of 0.04 μg CO₂-C g^{-1} soil h^{-1} per μg active biomass C g^{-1} soil.



FIGURE 2 | Soil respiration rates after addition of a glucose and nutrient solution in wet (A) and dry (B) soils at different temperatures (heated soils, red solid circles/solid line; and unheated soils, open blue circles/dashed line). Symbols represent means \pm SE. Lines were obtained by fitting the model parameters to measured soil respiration rates (see Materials and Methods Section). Fitted parameters are in Supplementary Table 2. Fitted lines are based on mean parameters values for each treatment. R^2 -values were calculated based on linearized model (Supplementary Figure 1) but exponential curves are shown to illustrate the exponential nature of the SIGR. Temperature typically explained more of the variation in soil microbial parameters (especially AMB and t_{lag}) than did soil moisture (**Figure 6**). However, both environmental factors were important in explaining variation in SBR, and the two together accounted for 70% of this variation. This analysis accounts for the direct effects of temperature and moisture on SBR (e.g., 29% of variance explained by temperature) and microbial parameters (e.g., 59% of variance in AMB explained by temperature). In contrast, the relationships between SBR and microbial biomass (**Figure 5**) suggest that microbial activation was the biological mechanism through which these abiotic factors influenced soil respiration rates (e.g., 62% of the variance in SBR was explained by environmentally—mainly temperature—driven changes in the pool of AMB).

DISCUSSION

Researchers long ago established that soil microbial respiration responds to changes in environmental conditions (Birch, 1958), but only recently have soil microbial processes been explicitly incorporated into soil carbon models (Fujita et al., 2014; Zhang et al., 2014; Wieder et al., 2015). The most common approach has been to represent microbial biomass as a single pool, without differentiating between microorganisms in active and dormant physiological states. Here, the observed increases in soil CO₂ efflux in response to warming and increased soil moisture were explained by an increase in the active fraction of the soil microbial community rather than by any increase in the size of the total community (i.e., TMB). As we predicted, SBR was more strongly correlated with AMB than with TMB. An increase in AMB after warming and wetting with no significant change in TMB suggests a quick shift of dormant microorganisms to an active physiological state, which corresponded to the increase in respiration. Given that the specific growth rate is an intrinsic feature of the microbial population (Panikov, 1995; Rousk and Bååth, 2007), the insensitivity of μ -values to soil wetting and warming indicates that this microbial activation did not dramatically change the functional structure of the





microbial community (i.e., the dominant population growing on added substrate). Moreover, the stronger correlation of t_{lag} with AMB than with μ supports an observation that lag time is regulated primarily by activity state rather than by maximal specific growth rate, even though both parameters are used in calculation of t_{lag} (Blagodatskaya et al., 2014). Overall, these results are consistent with observations from other studies in which soil respiration responses to warming (Hagerty et al., 2014) and soil moisture (Fierer and Schimel, 2002; Butterly et al., 2009; Placella et al., 2012) were not explained by changes in TMB but by changes in the physiology of soil microbial communities, such as resuscitation of physiologically clustered microbial groups (Placella et al., 2012; Aanderud et al., 2015; Barnard et al., 2015).

Direct effects of temperature and moisture on SBR can differ in magnitude from those on specific respiration-related microbial processes (e.g., activation of dormant biomass). The combined analysis of the proportion of variances explained by abiotic factors and the relationships between SBR and total and active MB suggests that SBR responses to temperature are strongly linked to changes in AMB. The activation of dormant microorganisms under warmed conditions raises the question of how and why environmental factors trigger activation of different microbial taxa. While we cannot directly address these questions here, we speculate that the temperature of heated soils may have been closer to the optimum for microbial processes than that of unheated soils. On the other hand, moisture explained a larger portion of the variation in SBR than in AMB. This could be explained in part by the fact that AMB, but not SBR, was insensitive to moisture under unheated conditions. This suggests that the importance of microbial physiology for explaining soil carbon processes could vary across moisture (Manzoni et al., 2016) and temperature gradients. The proportions of the variances explained by temperature and moisture also differed between SBR and TMB. This further supports the lack of a



biomass [top; $F_{(1, 11)} = 15.97$, $p = 0.003 R^2 = 0.62$; statistics were calculated using log₁₀-transformed AMB] and total microbial biomass [bottom; $F_{(1, 11)} = 0.47$, p = 0.509, $R^2 = 0.04$].

relationship between TMB and SBR, at least at the temporal scales relevant for this study.

Different responses of TMB and AMB to environmental conditions may be explained by recycling of soil nutrients. The 1-week incubation at optimum warming and soil moisture conditions (i.e., the warmer and wetter soils) decreased TMB but increased AMB, suggesting that part of the dead biomass was re-utilized by the active microbial fraction. By recycling energy and nutrients from dead biomass, microorganisms in the more favorable conditions likely remained in an active state, which reduced total biomass, but maximized active biomass and led to the highest fraction of active biomass $(5.0 \pm 0.2\%$ of TMB) in the warm and wet treatment. Rapid declines in TMB after re-moistening of pre-conditioned soil have been detected under laboratory conditions without substrate addition (Butterly et al., 2009; Blagodatskaya et al., 2011; Tian et al., 2015). Our study revealed that such a decline in TMB is not necessarily accompanied by a decrease in AMB,



suggesting a possible mechanism for maintaining activity under unfavorable conditions (e.g., starvation). Similar results have been observed in soil from an old-field experiment in Wagga Wagga, NSW, Australia, where single drying-rewetting events strongly decreased microbial biomass C and P, but increased microbial activity (Butterly et al., 2009). Decreases in TMB accompanied by increases in AMB have also been observed at the seasonal scale: TMB and AMB decreased and increased, respectively, from January to June in fallow and winter wheat soils from an experimental field in Hélécine, Belgium (Van de Werf and Verstraete, 1987). Taken together, these observations suggest that, at both short-term (i.e., few days) and seasonal scales, TMB and AMB can have different and even opposite responses to environmental conditions.

Another explanation of the phenomena observed in this study could be that warm and moist conditions quickly activated the grazing of microbial biomass by protozoans (Pomeroy, 1974), rapidly decreasing microbial biomass and enabling rapid nutrient recycling by protozoan grazers (Bonkowski et al., 2000). Nutrients released by protozoans could have facilitated microbial turnover, increasing the active fraction of the microbial community (Blagodatskaya et al., 2014). Conversely, drier conditions may have prevented such fast self-digesting by reducing the activity of protozoa, which are very sensitive to drought (Geisen et al., 2014). We conclude that the mechanisms controlling soil nutrient recycling (e.g., reutilization of dead biomass or belowground grazing) drive the different responses of TMB and AMB to environmental conditions.

At the seasonal scale, variations in soil respiration rates are generally much larger than variations in TMB. Soil respiration rates are faster in the growing season than in winter in a variety of ecosystems (Grogan and Chapin, 1999; Suseela et al., 2012; Suseela and Dukes, 2013; Keidel et al., 2014; Reynolds et al., 2014; Wang Y. et al., 2014). Typically, these variations closely follow changes in temperature and soil moisture (Grogan and Chapin, 1999; Suseela et al., 2012; Suseela and Dukes, 2013). Across seasons, TMB generally varies less than soil heterotrophic respiration (Holmes and Zak, 1994; Gunapala and Scow, 1998; Blume et al., 2002; but see Devi and Yadava, 2006). Our results suggest that this difference could be explained in part by seasonal changes in the pools of active/dormant microbial biomass. However, this hypothesis remains untested.

At the global scale, net changes in the size of active and dormant microbial biomass pools (e.g., due to climate change) could strongly affect biogeochemical processes such as carbon cycling. Transitions between active and dormant physiological states have been incorporated into some dynamic microbial community models to simulate community responses to environmental changes, such as drying-rewetting cycles (Konopka, 1999; Bär et al., 2002; Stolpovsky et al., 2011). However, few attempts have been made to include these transitions in larger, ecosystem-scale models (Manzoni et al., 2014; Wang G. et al., 2014; He et al., 2015; Tang and Riley, 2015). Although other microbial-based models only consider TMB (Wieder et al., 2015), microbial biomass in these models is capable of growing and readily responding to substrate inputs. Thus, in practice, these models actually simulate the active fraction of TMB (although they ignore changes in the AMB pool due to active \leftrightarrow dormant biomass transitions). Our results suggest that it would be more appropriate to validate/calibrate these models with a microbial database that includes AMB or the active fraction of TMB than with databases that only consider TMB (Serna-Chavez et al., 2013; Xu et al., 2013). Currently, though, there are not enough empirical data to create such a database, and our understanding of spatial and temporal patterns of AMB remains rudimentary. Some models have been validated based on TMB data (Wang G. et al., 2014; Buchkowski et al., 2015; He et al., 2015). In agreement with our findings, TMB has been found to be far less responsive to external factors than predicted by models, and a poor predictor of soil respiration rates (Buchkowski et al., 2015). Attempts to validate models that explicitly represent active/dormant pools using TMB data have resulted in estimates of large fractions of active biomass (70-90%, Wang G. et al., 2014) that seem unrealistic when compared with our findings and with most current reports (generally <10-20%, Lennon and Jones, 2011), or that produce simulated relationships that are not supported by our findings (e.g., more variation of AMB explained by moisture than by temperature, He et al., 2015). Overall, incorporation of microbial dormancy in ecosystem models has facilitated model development and validation (Wang et al., 2015) produced more accurate predictions of soil heterotrophic respiration and microbial biomass (He et al., 2015), and led to predictions of weaker carbon-climate feedbacks than those given by microbial models that do not represent active/dormant biomass pools (Tang and Riley, 2015). Taken together, this suggests that incorporation of dormancy in ecosystem models influences predictions of future carbonclimate feedbacks and leads to a more realistic (and still mathematically synthesizable, computationally plausible, and experimentally testable) representation of microbial influences on soil carbon cycling. This also highlights the need for experimental work that tests these hypotheses at large spatiotemporal scales.

We know of only three previous studies that have investigated the relationship between AMB and microbial respiration in soils (Alvarez et al., 1998; Aanderud et al., 2015; Barnard et al., 2015), and only one of these studies (Barnard et al., 2015) was designed to examine relationships among environmental conditions, AMB, TMB, and soil respiration. The paucity of previous studies limits our ability to extrapolate from the AMBrespiration relationship that we observed to AMB-respiration relationships in other ecosystems and environmental conditions. In our study, each gram of C in AMB was associated with the emission of ~ 0.04 g CO₂-C h⁻¹. In a pasture topsoil (0–15 cm) from Pergamino, Argentina, this relationship varied from 0.01 to 0.18 g CO₂-C h⁻¹ per g active biomass C, depending on the availability of labile C (Alvarez et al., 1998). This variation was explained by changes in the composition of the active microbial pool, or by changes in the physiology of the extant soil microbial community (e.g., prevalence of aged cells with lower CO₂ production per unit of biomass in soils with scarce or inaccessible labile substrates). The CO₂ released from 15-cm deep cores of California grassland soils in the first 2 h after a rewetting event ranged from 8 to 33 mmol m^{-2} , depending on the precipitation pattern preceding the rewetting event. This peak in soil respiration was strongly linked (R = 0.83, p < 0.001) to increases in the abundance of potentially active bacteria (Barnard et al., 2015). Similarly, in agricultural, grassland, and forest soils from southwestern Michigan, USA, the resuscitation of "rare biosphere" (defined as soil bacterial taxa that were not detectable in dry soils but became detectable after a rewetting event) was associated with 5-20 fold increases in net production of soil CO₂ (Aanderud et al., 2015). Other studies have quantified fractions of AMB across a variety of systems. Estimates range from 4 to 49%, depending on season, land use, and soil depth (Van de Werf and Verstraete, 1987), 0.02 to 24.2% depending on soil age (Khomutova et al., 2004), and 0.24 to 0.32% depending on root presence (Blagodatskaya et al., 2014). However, these studies did not explicitly quantify the relationship between AMB and soil carbon flux, which limits our ability to conduct a broader analysis of the AMB-respiration relationship.

Another factor to consider when comparing observations of AMB from different studies is the diverse array of methods used to make these estimates (Blagodatskaya and Kuzyakov, 2013; Yakushev, 2015). In contrast, most TMB data have been collected using the same method: chloroform fumigation (Serna-Chavez et al., 2013; Xu et al., 2013). The different techniques used to estimate AMB have not been directly compared with one another, but typically produce AMB estimates of similar magnitudes (Blagodatskaya and Kuzyakov, 2013). The kinetic approach that we used in this study has important advantages, including that it can be used to quantify microbial responses to the environment at the community level, and that the outcomes are mathematically compatible with the type of data used to parameterize/validate models (e.g., size of carbon pools, units of carbon released per unit of AMB, etc). Other

methods can give information that improves our understanding of soil microbial ecology and biochemistry, but that currently cannot be incorporated into models due to mathematical and/or computational limitations (e.g., use of molecular techniques to study the responses of specific microbial phylogenetic groups to environmental conditions). One of the caveats of the SIGR method is that it relies on assumptions (see Section Kinetic Respiration Analysis) that, despite having been exhaustively tested in many systems (Payne, 1970; Akimenko et al., 1983; Panikov and Sizova, 1996; Blagodatsky et al., 2000; Wutzler et al., 2012), may not be true in all cases. Another characteristic of this method is that it simplifies the continuous gradient between active, potentially active, and dormant biomass (Blagodatskaya and Kuzyakov, 2013) into two discrete pools: active and dormant biomass. Given the marked differences between the metabolism of microorganisms in active and dormant physiological states, and the ways they interact with their environment (Lennon and Jones, 2011), this seems a fair simplification, and a useful one for modeling purposes.

While we observed a strong response of AMB to temperature (e.g., **Figure 6**), and a strong link between AMB and soil respiration (e.g., **Figure 5**), the overall importance (and dependence on other factors, e.g., water-resource availability interactions) of changes in active/dormant microbial pools for soil processes at large spatio-temporal scales still remains uncertain. It is not clear, for example, whether climatic changes predicted for this century (e.g., warming and increased precipitation variability) could cause a net increase in global AMB by (net) activating some fraction of dormant soil microorganisms. If this were to occur, predictions of future soil CO_2 production based on TMB only (i.e., without considering active/dormant transitions) could underestimate future soil CO_2 emissions.

We conclude that soil respiratory responses to short-term changes in environmental conditions are better explained by changes in the active fraction of the soil microbial pool than by changes in TMB. Based on these results, we suggest that decomposition models that explicitly represent microbial processes should take into account the active microbial pool, and recommend researchers be cautious when comparing modeled microbial pool sizes with measurements of microbial biomass.

AUTHOR CONTRIBUTIONS

AS substantially contributed to the conception, design, and critical review of the work, as well as to the acquisition, analysis, and interpretation of the data. EB substantially contributed to the analysis and interpretation of the data, as well to the critical review of the work. JD substantially contributed to the conception, design, and critical review of the work, as well as to the analysis, and interpretation of the data.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: http://journal.frontiersin.org/article/10.3389/fmicb. 2016.00524

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Conflict of Interest Statement: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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