Higher Long-Term Soil Moisture Increases Organic Carbon Accrual Through Microbial Conversion of Organic Inputs

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Abstract

High long-term soil moisture may either stimulate or inhibit soil organic carbon (SOC) losses through changes to mineral and chemical composition, and resultant organo-mineral interactions. Yet, the trade-off between mineralization and accrual of SOC under long-term variation in unsaturated soil moisture remains an uncertainty. In this study, we tested the underexplored relationships between long-term soil moisture and organo-mineral chemical composition, and its implications for SOC persistence. The results provide new insights into SOC accrual mechanisms under different long-term moisture levels commonly observed in well-drained soils. Differences in long-term mean volumetric water content ranging from 0.4 - 0.63 (v/v) on fallow plots in an experimental field in New York, USA, were positively correlated with SOC contents ($R^2 = 0.228; P = 0.019; n = 20$), mineral-associated organic matter (MAOM) ($R^2 = 0.442; P = 0.001; n = 20$) and occluded particulate organic matter (oPOM) contents ($R^2 = 0.178; P = 0.033; n = 20$). Higher long-term soil moisture decreased the relative content of sodium pyrophosphate extractable Fe ($R^2 = 0.33; P < 0.005; n = 20$), increased that of sodium dithionite extractable Fe ($R^2 = 0.443; P < 0.001; n = 20$), and increased the overall importance of non-crystalline Al pools (extracted with sodium pyrophosphate and hydroxylamine extractable) for SOC retention. Higher long-term soil moisture supported up to a four-fold increase in microbial biomass (per unit SOC), and lower C:N ratios in MAOM fractions of high-moisture soils (from C:N 9.5 to 9, $R^2 = 0.267, P = 0.011, n = 20$). This was reflected by a 15% and 10% greater proportion of oxidized carboxylic-C to aromatic-C and O-alkyl C, respectively, as measured with $^{13}$C-NMR, and a more pronounced FTIR signature of N-containing proteinaceous compounds in high-moisture MAOM fractions, reflective of microbial metabolites. SOC accrual increased with increasing soil moisture ($P = 0.019$), exchangeable Ca$^{2+}$ ($P = 0.013$), and pyrophosphate-extractable Al content ($P = 0.0001$) and Al/Fe ratio ($P = 0.017$). Taken together, our results show that high long-term soil moisture resulted in SOC accrual by enhancing microbial conversion of plant inputs to metabolites that interact with reactive minerals.
1. Introduction

Soils comprise the largest terrestrial store of organic carbon (OC) (Friedlingstein et al., 2019) and play a substantial role in the global C cycle (Scharlemann et al., 2014). As a result, increasing soil organic carbon (SOC) storage by decreasing losses to CO₂ may be a feasible strategy to withdraw atmospheric CO₂ and partially offset anthropogenic emissions driving climate change (Minasny et al., 2017). As climate change is expected to drastically alter soil moisture conditions globally (Seneviratne et al., 2010; Grillakis, 2019), a better understanding of the trade-off between SOC mineralization and stabilization under long-term changes to soil moisture is needed to manage SOC stocks (Falloon et al., 2011). Importantly, unlike soil temperature which is expected to rise and increase SOC mineralization (Soong et al., 2021), soil moisture is a parameter that can be managed - through irrigation, controlled drainage, and wetland restoration - and can indirectly curb soil warming through evapotranspiration (Seneviratne et al., 2010).

Soil moisture controls SOC turnover and storage by regulating fundamental processes such as soil biotic activity (Moyano et al., 2013), solute transport, gaseous exchange, and mineral weathering (Schjønning et al., 2003; Moyano et al., 2013; Kramer and Chadwick, 2018). Short-term increases in soil moisture are known to stimulate microbial activity (Ghezzehei et al., 2019), yet long-term increases may either stimulate or decrease SOC losses by changes to vegetation (Chen et al., 2020), mineral composition (Doetterl et al., 2015; Khomo et al., 2017), SOC chemical composition (Hall et al., 2018) and resultant organo-mineral interactions (Mikutta et al., 2009; Kramer et al., 2012). Most of these studies evaluated broad gradients in precipitation, soil mineral properties, or soil age, which inevitably interact in complex ways. Yet, the direct effects of long-term differences in soil moisture on SOC turnover has received little attention (Moyano et al., 2013), largely due to the challenge of isolating measurable variation in long-term soil moisture at field scale. This has limited our ability to predict and manage SOC stocks in the face of changing soil moisture conditions (Todd-Brown et al., 2014; Berg and Sheffield, 2018).

SOC is widely considered to consist of microbial residues and necromass (Schurig et al., 2013; Zhu et al., 2020) which have been protected from further decomposition through interactions with the mineral
matrix (Kleber et al., 2021). Therefore, SOC turnover is largely dependent on the efficiency of microbial conversion of organic inputs into microbial biomass (i.e., substrate use efficiency), with higher efficiency indicative of greater C storage in the soil as microbial biomass (Manzoni et al., 2012) and mineral-associated SOC (Zhu et al., 2020). Long-term soil moisture affects this efficiency by regulating diffusion rates of substrates used by microbial communities (Butcher et al., 2020). Low water content inhibits diffusion of dissolved compounds in the soil solution due to thinning of water films in soil pores, increased flow path length and solution viscosity (Butcher et al., 2020), thus limiting the probability of encounters between microbes, extracellular enzymes, and substrates. Recent work has shown that limited substrate diffusion at low soil water contents can decrease microbial carbon use efficiency (Butcher et al., 2020).

Additionally, increased microbial activity at high moisture conditions can enhance the conversion of high molecular weight, hydrophobic plant inputs into soluble, charge-containing, low molecular weight compounds which tend to interact with mineral surface (Lehmann and Kleber, 2015). Thus, SOC cycled under higher long-term soil moisture may have greater persistence and result in net SOC accrual, yet experimental evidence is lacking.

The amount and nature of soil minerals often control the content, composition, and persistence of adsorbed or occluded SOC (Sanderman et al., 2014; Rasmussen et al., 2018; Kleber et al., 2021). SOC may predominantly interact with non-crystalline or crystalline oxide phases, or with phyllosilicate clay minerals, depending on their type and relative content in the soil (Mikutta et al., 2006; Khomo et al., 2017; Hall et al., 2018; Rowley et al., 2018). Furthermore, long-term differences in soil moisture can result in different rates of chemical weathering of minerals which can in turn alter their reactivity and capacity to retain SOC, with implications for SOC stabilization (Doetterl et al., 2015). However, the effects of mineral composition on organo-mineral interactions may be confounded by other factors such as climate and vegetation.

Therefore, we aimed to investigate the effects of increasing long-term soil moisture under unsaturated conditions on microbial turnover of SOC, SOC storage in soil density fractions, and the chemical composition of organo-mineral associations. In addition, we probed the effect of these variables on the
persistence of SOC by determining mineralizability in laboratory incubations. We hypothesized that (1) soils experiencing higher long-term moisture that is more conducive to microbial activity will have more oxidized SOC functional groups and higher contents of non-crystalline oxide phases due to higher decomposition and weathering, respectively, and (2) that organo-mineral interactions in high-moisture soils will support SOC accrual.

2. Materials and Methods

2.1 Field site

The field site located near Ithaca, New York, USA (42N28.20',76W25.94'), included three primary soil series: well drained Canaseraga (coarse-silty, mixed, active, mesic typic Fragiudept), slightly poorly drained Dalton (Coarse-silty, mixed, active, mesic aeric Fragiaquept) and poorly drained Madalin (fine, illitic, mesic mollic Endoaqualf) (Das et al., 2018). The epipedon texture is primarily silt loam, characterized by dense subsoil fragipans and recurring perched water tables resulting from the restrictive subsoil layers. The field topography is undulating, with slopes varying from 0 to 8% (and a small area with short slopes of up to 15% on the eastern edge). The mean annual temperature and precipitation at the site are 10 °C and 940 mm, respectively.

Four fallow plots were laid out to capture the range of soil moisture regimes, varying from moderately well-drained to poorly drained (Das et al., 2019). The characteristic soil moisture regime along five sampling subplots in each of the four plots was previously determined following a soil moisture measurement campaign carried out from 2011 to 2015 (Das et al., 2018) (See Figure S1 for 2011-2018 data). The volumetric water content of each sampling site’s surface layer was measured via replicated time-domain reflectometry (TDR) and the relative soil water content was determined by dividing the sample location volumetric water content by the mean field value for that sampling event. For each of over 40 sampling events, a normalized moisture value was calculated for each of 80 sampling sites (including the 20 fallow sites and 60 other non-fallow sites not reported here) based on the multi-year mean of its relative
water content. Thus, the multi-year mean cumulatively represents several thousand readings at the field (Das et al., 2018). The normalized moisture values were binned into quantiles (Das et al., 2018) (Table S1).

The driest subplots (denoted as Q1) had the lowest water contents relative to the field mean, and the wettest subplots (denoted as Q5) had the highest water contents relative to the field mean.

The plant cover at the time of sampling consisted of fallow (unmowed for circa 10 years) grasses dominated by legacy reed canarygrass (*Phalaris arundinacea* L.) interspersed with numerous other grasses and broadleaf forbs. The plots were undisturbed aside from small (1 m²) hand harvests at subplots after dormancy to characterize yields for comparison with other cropping treatments.

### 2.2 Sampling method

Soils were sampled from the surface Ap layer (0 - 0.15 m depth) in October 2018 at two locations equidistant (1.2 m) from the center of each of 20 sampling points (marked by a permanent flag). Approximately two kg of soil were dug from each of the two locations, composited in a bucket, mixed, and transported to the lab in a cooler. Half of each sample was air dried and passed through a 2 mm sieve, and half was passed through a 4 mm sieve and stored at -20 °C for microbial biomass and soil respiration analyses. Visible rocks and plant material were removed from each sample. Composited samples, prepared by combining equal amounts of the soils from quintiles Q1, Q3, and Q5 (Table S1), were used for ^13^C NMR and C NEXAFS analyses. These composited samples are referred to as Low moisture, Mid moisture, and High moisture, respectively. Collection of above- and below-ground biomass was described previously (Das et al., 2019).

### 2.3 Soil analyses

#### 2.3.1 Soil characterization

Gravimetric and hygroscopic water contents were determined on field moist and air-dry subsamples, respectively, after drying each to constant weight at 105 °C. Gravimetric soil water content at saturation, used to calculate the water filled pore space, was determined as previously described (DeCiucies et al., 2018). Soil pH was measured on 1:2.5 soil:deionized water extractions. Exchangeable cations were
extracted using 1 N ammonium acetate at pH 7 and measured on an inductively coupled plasma spectrometer (Thermo iCAP 6000 series).

### 2.3.2 Soil fractionation

We used a combination of size and density fractionation to isolate operationally defined - but ecologically relevant - fractions. Soil samples (10 g) were gently agitated for 10 s with 35 mL of sodium polytungstate (SPT) adjusted to a density of 1.65 Mg m$^{-3}$. The samples were left to settle overnight, centrifuged (3000 RCF, 30 min), filtered (GF/F, 0.7 µm glass fiber filter), and washed with 500 mL of deionized water. The obtained material is referred to as the free particulate organic matter (fPOM). Fresh SPT solution (35 mL, 1.65 Mg m$^{-3}$) was added to the samples and a vortex was used to re-disperse the soil. The samples were sonicated (XL 2020, QSonica, Newtown, CT, USA) at 350 J mL$^{-1}$ of energy (operated at 75 J s$^{-1}$), left overnight to allow the particles to settle, and centrifuged (3000 RCF, 45 min). The floating material, referred to as the occluded particulate organic matter fraction (oPOM), was isolated as described above. The remaining pellet was washed with deionized water and centrifuged (3000 RCF, 30 min) three times to remove the SPT (supernatant density was 1±0.02 Mg m$^{-3}$). Next, the soils were shaken end-to-end with sodium hexametaphosphate (35 mL, 0.5% w/v) for 16 hours and wet sieved (53 µm) to separate the sand sized fraction (material remaining on the sieve) from the silt and clay size fractions (material passing the sieve). The material passing the sieve was referred to as the mineral associated organic matter (MAOM) fraction. The MAOM and sand fractions were transferred to a pre-tared aluminum tin and dried at 60 °C. The four obtained fractions – fPOM, oPOM, MAOM, and sand - were weighed, ball milled (except the sand), and stored. Total C, N, and isotope ratios of the fractions (except the sand) and bulk soil samples were measured using a Delta V Isotope Ratio Mass Spectrometer (Thermo Scientific, Germany) coupled to a Carlo Erba NC2500 Elemental Analyzer (Italy). We assumed total C equaled organic C since these soil samples did not contain carbonates (Das et al., 2018) and soil pH was always < 7. The mean mass recovery was 102%. Recovery of C ranged from 96 - 109% and recovery of N ranged from 98 -
106%. Sand was assumed to contain zero C and N. The average C concentrations in fPOM and oPOM were 35% and 25%, respectively, indicating that some minerals were present in these isolated fractions.

2.3.3 Oxide extraction

We performed a sequential oxide extraction following a modified protocol (Heckman et al., 2018) using sodium pyrophosphate (PY), hydroxylamine hydrochloride (HH), and sodium dithionite (DITH) to sequentially release SOC retained by different forms of soil Al and Fe. Based on published extraction protocols (Ross et al., 1985; Parfitt and Childs, 1988; Wagai and Mayer, 2007; Courchesne and Turmel, 2007; Heckman et al., 2018) we assume that PY extracts organically complexed Al and Fe, HH extracts non-crystalline inorganic Fe including ferrihydrite, and DITH extracts crystalline Fe phases like goethite, hematite, and lepidocrocite. We acknowledge that Al and Fe released by PY, HH, and DITH extractions only approximately correspond to specific phases present in the soil as these extractions are not perfectly selective (Parfitt and Childs, 1988; Kaiser and Zech, 1996). Yet, this remains an important approach to estimate Al and Fe phases and their impact on SOC storage and turnover times (Masiello et al., 2004; Porras et al., 2017; Wagai et al., 2020).

Briefly, sodium pyrophosphate (25 mL, 0.1 M), hydroxylamine hydrochloride (25 mL, 0.1 M prepared in 0.25 M HCl) and sodium dithionite (3 mL, 0.5 M) were sequentially added to soil samples (0.1 g), shaken for 16 hours, centrifuged (3000 RPM, 1 hour), and the clear supernatant was passed through glass fiber filter papers (0.45 µm, Advantec). Following extraction with sodium dithionite, soils were shaken for one hour with an HCl solution (1 mL, 0.05 M) which was combined with the sodium dithionite extract (Heckman et al., 2018). The concentration of dissolved Fe and Al was measured by on an inductively coupled plasma spectrometer (Thermo iCAP 6000 series). The concentration of OC in each extract, considered the C associated with Al and Fe, was measured by combustion catalytic oxidation (TOC-VCPN TOC analyzer Shimadzu, Japan).
2.3.4 Microbial biomass carbon

Microbial biomass C was measured following the chloroform-fumigation-extraction method (Witt et al., 2000). Briefly, ethanol-free chloroform (4 mL) was added to 10 g of field moist soil (<4 mm) in stoppered 250 mL Erlenmeyer flasks. The samples were incubated for 24 hours, after which the flasks were vented in a fume hood until the chloroform had fully evaporated. Fumigated and non-fumigated soils were agitated for 1 hour with K$_2$SO$_4$ (0.05 M, 50 mL) and filtered through No. 1 Whatman paper. The total extractable organic C was measured by combustion catalytic oxidation (TOC-VCPI TOC analyzer Shimadzu, Japan). Microbial biomass C was calculated as the difference between paired chloroform-fumigated and non-fumigated subsamples.

2.4 Spectroscopic studies

2.4.1 Fourier Transform Infrared (FTIR) spectroscopy

FTIR spectra were collected on a Vertex 70 spectrometer (Bruker Corp., Billerica, MA) equipped with an ATR sampling accessory (GladiATR, Pike Technologies, Madison, WI). Spectra were collected on dry (60 °C) ball-milled samples ($n = 20$) of soils, MAOM, fPOM, and oPOM fractions, and on ashed (500 °C for 3 h) MAOM samples. Spectra representing an average of 64 scans were collected from 450 to 4500 cm$^{-1}$ at a resolution of 4 cm$^{-1}$. Replicate spectra for each sample ($n = 3$) were baseline corrected, normalized, and averaged. To identify features originating from organic constituents, spectra of ashed MAOM were subtracted from non-ashed MAOM samples, and smoothed (17-point Savitsky-Golay smoothing). Spectral processing was performed using the OPUS 7.2 software (Bruker Corp., Billerica, MA). ATR-FTIR vibrational assignments of organic and inorganic constituents were based on previous publications (Lehmann et al., 2007; Parikh et al., 2014; Fine et al., 2018).

2.4.2 $^{13}$C Nuclear Magnetic Resonance (NMR)

The molecular structure of organic matter in the bulk soil and soil fractions was analyzed using $^{13}$C NMR spectroscopy (Biospin DSX 200 NMR spectrometer, Bruker, Rheinstetten, Germany) with a contact
time of 0.001 sec with a pulse delay of 0.4 sec for bulk soil and silt and clay fractions and 1 sec for fPOM and oPOM fractions. At least 100,000 accumulated scans were performed. The spectra were integrated using four major chemical shift regions: 0 to 45 ppm (alkyl-C), 45 to 110 ppm (O/N-alkyl-C), 110 to 160 (aryl-C), and 160 to 220 ppm (carboxyl-C) (Knicker and Lüdemann, 1995). Sample pre-treatment with hydrofluoric acid was not necessary to obtain a well-resolved spectrum. We applied a molecular mixing model (Nelson and Baldock, 2005) to estimate the proportions of biochemical component equivalents from $^{13}$C NMR spectral regions and N:C ratios. The biochemical components of the model are carbohydrate, protein, lignin, lipid, char, and carbonyl.

2.4.3 Carbon K-edge Near Edge X-Ray Absorption Fine Structure (NEXAFS)

SOC chemical composition in bulk soil and soil fractions was analyzed with C K-edge NEXAFS collected at the spherical grating monochromator (SGM) beamline of the Canadian Light Source (Saskatoon, SK). The beamline is equipped with a silicon drift detector (SDD) for partial fluorescence yield (PFY). Samples were deposited on gold (Au)-coated silicon wafers, and air dried. Partial fluorescence yield (PFY) scans (60s) were collected with an ~50 x 50 μm probe in slew mode. The PFY with maximum C signal (detector 90° to incident beam) was normalized to the PFY scattering signal from a clean Au-coated Si wafer to account for in-line C contamination. The spectra were edge-normalized (edge step=1) and flattened in Athena (Demeter 0.9.25) (Ravel and Newville, 2005). SOC composition was assessed by comparing peak height ratios of the 3 primary spectral features: aromatic (C=C) at ~285.6 eV, substituted aromatic (e.g., C=C-OH) at ~286.7 eV, and carboxylic (C=O-OH) at ~288.7 eV (Solomon et al., 2009; Heymann et al., 2011). Peak heights were determined by the Gaussian function fitting algorithm in Fityk v. 1.3.1 (Wojdyr, 2010), identifying the maximum of a Gaussian function of full-width half maximum (FWHM) = 0.6 eV (Possinger et al., 2020).

2.5 Heterotrophic respiration incubations

For the incubation experiment, soil samples were thawed and air dried. 5 g from each soil sample were weighed in triplicate into 60 mL Qorpak vials, which were placed in 473 mL Mason jars along with a
20 mL glass vial containing a CO₂ trap (15 mL KOH, 0.18 M) made with CO₂-free deionized water. CO₂-free deionized water (5 mL) was added to the bottom of the jar to maintain a humid atmosphere. To account for the small amount of CO₂ present in the jar, measurements from blank jars with no soil were used. Samples were hydrated to a moisture level equivalent to 50% of water filled pore space with deionized water and incubated for 53 days at 20 °C in the dark. On days 2, 7, 18, 33, and 53 the jars were opened, and the electrical conductivity of the KOH solutions was determined. After each measurement, the CO₂ traps were replaced with fresh KOH solutions in new vials, and fresh deionized water was added to the bottom of the jar. At each sampling event, the average (n = 3) electrical conductivity value of the KOH solution from the blank jars was subtracted from each jar’s KOH electrical conductivity value. This corrected value was then converted into volume of CO₂ released by the sample using a standard curve, and further converted to mass C-CO₂ by applying the ideal gas law. The standard calibration curve was made by injecting known of volumes of 99.99% CO₂ (Airgas, Inc, Elmira, NY) into septa-lidded Mason jars containing empty Qorpak vials, CO₂ traps, and 5 mL CO₂-free water on the bottom. The electrical conductivity of the KOH solution was measured 24 hours after injection. Cumulative respiration was reported per unit soil (termed mineralization; mg CO₂-C/g soil), and per unit SOC (termed mineralizability; mg CO₂-C/g SOC).

2.6 Statistical methods

All statistical analyses were done in R (Version 4.04). Linear regression of response variables to normalized moisture was performed using the lm function. The effects of normalized moisture values and soil fraction on response variables were analyzed using ANOVA, followed by Tukey's HSD post hoc test to determine significant differences between treatments. Pearson’s correlation matrix was plotted using the corrplot package. Partial least squares (PLS) regression models were constructed using the mdatools package, to identify the variables that best explain SOC content, mineralization, and mineralizability. PLS regression models reduce dimensionality and create components that maximize the covariance between predictor and response variables. As such, PLS models are especially useful in datasets that contain a
similar number of predictor variables and observations, and/or when predictors are highly correlated. Selectivity Ratio and Variable Importance in Projection tests were used to inform variable selection. Variable Importance for Projection provided the best goodness of fit for all models, which was evaluated based on the highest cross-validated R$^2$ ($R^2_{cv}$), and lowest root mean of square error (RMSE).

3. Results

3.1 Soil organic matter fractions

SOC and TN content increased with increasing normalized moisture value ($R^2 = 0.228$, $P = 0.019$, and $R^2 = 0.249$, $P = 0.014$, respectively) (Figure 1). The amount of oPOM and MAOM per unit mass of soil increased with increasing normalized moisture ($R^2 = 0.178$, $P = 0.033$, and $R^2 = 0.442$, $P < 0.001$, respectively), but fPOM content did not ($P = 0.9$) (Figure 1). However, the proportion of C in each fraction per unit SOC did not differ across normalized moisture values (Figure S2).

C:N ratios decreased in the order of roots > shoots > fPOM > oPOM > MAOM = bulk soil but were not significantly different across normalized soil moisture values (Figure S3 and Table S2), apart from the C:N ratio of MAOM, which significantly decreased with increasing soil moisture level ($R^2 = 0.267$, $P = 0.011$) (Figure 2A). The $\delta^{13}$C values (mean±sd) in the bulk soil (-27.53±0.3 ‰) and MAOM (-27.35±0.3 ‰) fractions were significantly higher than those of the fPOM (-28.37±0.5 ‰) and oPOM (-28.80±0.4 ‰) fractions and plant biomass ($P < 0.0001$), though the interaction of normalized moisture and fraction on $\delta^{13}$C was not significant (Figure S4, Table S2). Additionally, $\delta^{13}$C values of fPOM, oPOM, MAOM fractions and bulk soils decreased with increasing normalized moisture values (Figure S4). Normalized soil moisture or its interaction with soil fraction did not significantly affect the $\delta^{15}$N values (Figure S4 and Table S2).
Figure 1. Contents of soil organic carbon (SOC) (A), total nitrogen (TN) (B), mass percentage of free and occluded particulate organic matter fractions (fPOM and oPOM) (C, D), and mineral associated organic matter (MAOM) fraction (E), as a function of normalized soil moisture values. Lines indicate significant linear regressions (P < 0.05).

3.2 Microbial biomass

Microbial biomass C increased with increasing soil moisture, even when normalized to unit mass SOC (P < 0.0001, $R^2 = 0.564$), while dissolved organic C (DOC) as a proportion of SOC decreased with increasing soil moisture ($R^2 = 0.23$, $P = 0.018$) (Figure 2). Overall, DOC and microbial biomass C constituted up to 0.1% and 1% of the SOC, respectively.
Figure 2. C:N ratio of the mineral associated organic matter (MAOM) (A), microbial biomass C (B) and dissolved organic C (C) (both normalized to SOC content), as a function of normalized soil moisture. Lines indicate significant linear regressions (P < 0.05).

3.3 Oxide extraction

The relative concentration of Al in each extract was similar across normalized soil moisture values (Figure 3A). However, the relative concentration of Fe in organic-metal complexes (Fe_{PY}) decreased with increasing normalized soil moisture (R^2 = 0.33, P = 0.005) and the relative Fe concentration in crystalline phases (Fe_{DITH}) increased with increasing normalized soil moisture (R^2 = 0.443, P = 0.001) (Figure 3B). In total, the ratio of Al to Fe in the PY and HH extracts increased with increasing normalized moisture (R^2 = 0.287, P = 0.009 for PY; R^2 = 0.363, P = 0.003 for HH), but remained similar in the DITH extract (Figure S6). In addition, the concentration of extracted DOC increased as the ratio of Al/Fe in the PY and HH extracts increased, (R^2 = 0.26, P = 0.013 for PY; R^2 = 0.18, P = 0.035 for HH) (Figure 3C), indicating that the increase in Al/Fe ratio in these extracts with increasing moisture was consistent with increasing DOC.

While the distribution of C across oxide pools varied with normalized moisture, total oxide-associated C normalized to SOC content (which constituted on average 44% of the SOC) did not differ across moisture values (P = 0.79, Figure S7). Linear regression analyses show that extractable Al was better correlated with DOC_{PY} and DOC_{HH} than Fe. Al_{PY} and Al_{HH} were significantly and positively correlated to DOC in the
respective extract ($R^2 = 0.513; P = 0.0002$, and $R^2 = 0.411; P = 0.0014$ respectively) (Figure 3E) while Fe\textsubscript{PY} and Fe\textsubscript{HH} were not significantly correlated to extracted DOC (Figure 3D). Neither Al\textsubscript{DITH} nor Fe\textsubscript{DITH} were significantly correlated to DOC\textsubscript{DITH}.

![Graphs](image-url)

**Figure 3.** Proportion of total Al (A), and total Fe (B), in sodium pyrophosphate, hydroxylamine hydrochloride, and sodium dithionite extracts, as function of normalized soil moisture value. Extractable dissolved organic carbon (DOC) as a function of Al/Fe ratio (C), extractable Fe (D) and Al (E) in sodium pyrophosphate, hydroxylamine hydrochloride, and sodium dithionite extracts. Lines indicate significant linear regressions ($P < 0.05$).
3.3 ATR-FTIR analysis

Bulk soil and MAOM spectra feature prominent bands assigned to Si-O stretching vibrations in quartz and clay minerals at 775, 795, and 1000 cm\(^{-1}\), and bands assigned to O-H and N-H stretching, and clay mineral structural hydroxyl groups at 3390, and 3622 cm\(^{-1}\), respectively (Figure S8). fPOM and oPOM spectra featured O-H and N-H stretching around the broad band centered at 3350 cm\(^{-1}\) and bands assigned to asymmetric and symmetric stretching of aliphatic CH\(_2\) at 2930 and 2855 cm\(^{-1}\). Features originating from organic components (1250-1800 cm\(^{-1}\)) were less apparent in MAOM and bulk soil spectra (Figure S7). To further investigate how differences in C:N ratios in the MAOM fractions were reflected in SOC composition, spectra of ashed MAOM samples were collected and subtracted from non-ashed MAOM spectra to reveal details of the organic constituents in the MAOM fraction. (Figure 4). The spectra between 1250-1800 cm\(^{-1}\) consisted of bands tentatively assigned to stretching of C=O at 1645 cm\(^{-1}\), COO\(^-\) and C=C at 1600 cm\(^{-1}\), C=N and bending N-H vibrations at 1545 cm\(^{-1}\), and COO\(^-\) at 1420 cm\(^{-1}\). We found higher absorbance in bands assigned to asymmetric and symmetric carboxylate COO\(^-\) stretch and amide/ketone C=O stretch (1420, 1600, and 1645 cm\(^{-1}\), respectively), symmetric O-H stretching (3340 cm\(^{-1}\)) and asymmetric and symmetric CH\(_2\) stretching (2930 and 2855 cm\(^{-1}\)) with increasing moisture level. Furthermore, there was a relative increase in the absorbance at 1545 cm\(^{-1}\), assigned to stretching of aromatic C=N and bending of N-H.

3.4 \(^{13}\)C NMR analysis

In the MAOM fraction, the ratio of carboxyl-C/aromatic-C increased with increasing moisture level \((R^2 = 0.731; P = 0.239, \text{Table S3})\), as measured by \(^{13}\)C NMR (Figure 4 and Figure S9). In addition, the ratio of carboxyl-C/O-alkyl-C and the ratio of O-alkyl-C/alkyl-C in the MAOM fraction increased with increasing moisture level (Table S3). Comparing fractions, carboxyl-C and alkyl-C forms increased, and aromatic-C and O/N-alkyl-C decreased in the MAOM fraction compared to the fPOM and oPOM fractions. The composition of bulk soil SOC resembled that of the MAOM fraction, since MAOM contained most of the organic matter. According to the molecular mixing model, this shift echoes higher relative contents of
carbohydrates and char in the fPOM and oPOM fractions in comparison with higher relative contents of proteinaceous and aliphatic compounds in the MAOM fraction and bulk soil (Figure S10).

Figure 4. Spectroscopic investigation of the chemical composition of mineral associated organic matter (MAOM) fractions from low-, mid-, and high-moisture levels. ATR-FTIR spectra of ashed MAOM fractions \((n = 20)\) subtracted from non-ashed MAOM samples \((n = 20)\), highlighting the spectral features of SOC and differentiating them from mineral bands. Spectra were averaged across the moisture levels. The information-poor spectral region of 2000 – 2800 \(\text{cm}^{-1}\) is omitted (A and B). \(^{13}\)C NMR spectra of MAOM fractions of composited samples from each moisture level (C). Carbon K-edge near-edge X-ray absorption fine structure (XANES) spectra of MAOM fractions of composited samples from each moisture level (D).
3.5 Carbon K-edge NEXAFS analysis

Confirming the $^{13}$C-NMR results, we found that the ratio carboxyl C / (aromatic + substituted aromatic C), measured with C NEXAFS (Figure 4D), increased with increasing long term moisture level. We did not evaluate changes in alkyl-C and O-alkyl-C forms since NEXAFS spectra do not have a strong and well-defined feature corresponding to these bonding environments (Heckman et al., 2017). Overall, carboxyl-C forms relatively increased, while aromatic-C forms relatively decreased in the MAOM fraction relative to the fPOM and oPOM fractions ($R^2 = 0.866; P = 0.166$, Table S3).

3.6 SOC mineralization and mineralizability

Cumulative SOC mineralization and mineralizability after 53 days of incubation are plotted as means and standard errors for each moisture level (Figure 5, Table S1). Mineralization was positively and significantly correlated with normalized moisture values ($R^2 = 0.346$, $P = 0.0037$), but no significant relationship was found between SOC mineralizability (mineralization normalized to SOC) and soil moisture value ($P = 0.68$).

Figure 5. Cumulative SOC mineralization (A) and mineralizability (B) over time in soils varying in moisture levels ($n = 20$), plotted as mean and standard error for each moisture level.
Pearson’s correlation coefficients and partial least squares (PLS) regression coefficients for SOC content, mineralization, and mineralizability are presented in Figure 6 and Table 1, respectively. SOC contents were highly positively correlated with $\text{Ca}_{\text{EX}}$ ($r = 0.67$), $\text{Al}_{\text{PY}}$ ($r = 0.65$), $\text{Al}_{\text{PY}}/\text{Fe}_{\text{PY}}$ ($r = 0.62$), and normalized moisture ($r = 0.52$), and negatively correlated with DOC ($r = -0.71$). Similarly, the PLS regression model ($R^2_{cv} = 0.449$; RMSE$_{cv} = 0.439$) identified $\text{Al}_{\text{PY}}, \text{Ca}_{\text{EX}}, \text{Al}_{\text{PY}}/\text{Fe}_{\text{PY}},$ and normalized moisture as positively associated with SOC contents. SOC mineralization was correlated to SOC ($r = 0.81$) and $\text{Ca}_{\text{EX}}$ ($r = 0.7$), and negatively correlated with DOC ($r = -0.41$). The PLS model identified SOC and fPOM as positively associated with SOC mineralization ($R^2_{cv} = 0.695$; RMSE$_{cv} = 0.109$). SOC mineralizability was weakly positively correlated with microbial biomass C ($r = 0.39$), and negatively correlated with C:N ($r = -0.49$), $\text{Al}_{\text{PY}}$ ($r = -0.43$), and $\text{Fe}_{\text{PY}}$ ($r = -0.42$). Similarly, the PLS model identified $\text{Al}_{\text{py}}, \text{Fe}_{\text{py}},$ and $\text{Fe}_{\text{DITH}}$ as negatively associated with SOC mineralizability ($R^2_{cv} = 0.033; \text{RMSE}_{cv} = 1.864$).

**Table 1.** Results of partial least squares regression models. Coefficients are listed if they are statistically significant at $P < 0.05$, italicized at $P < 0.01$, and bolded at $P < 0.001$. The coefficient of determination of the calibration ($R^2_{cal}$) and cross-validation sets ($R^2_{cv}$), and the root mean of square error (RMSE$_{CV}$) are provided.

|                  | Moisture | SOC  | $\text{Ca}_{\text{EX}}$ | $\text{Al}_{\text{PY}}$ | $\text{Fe}_{\text{PY}}$ | $\text{Al}_{\text{PY}}/\text{Fe}_{\text{PY}}$ | $\text{Fe}_{\text{DITH}}$ | fPOM | $R^2_{cal}$ | $R^2_{cv}$ | RMSE$_{CV}$ |
|------------------|----------|------|--------------------------|---------------------------|---------------------------|-----------------------------------------------|----------------------------|------|-------------|-------------|-------------|
| SOC              | 0.204    | --   | 0.266                    | **0.255**                 | --                        | 0.244                                        | --                         | --   | 0.601       | 0.449       | 0.439       |
| Mineralization   | --       | **0.416**| --                        | --                        | --                        | 0.272                                        | 0.836                      | 0.695 | 0.109       |             |             |
| Mineralizability | --       | --   | --                       | $-0.183$                  | $-0.147$                  | $-0.139$                                      | 0.304                      | 0.033 | 1.864       |             |             |
**Figure 6.** Heatmap showing Pearson correlations ($r$) for selected variables with SOC content, mineralization, and mineralizability. Only significant ($P < 0.05$) correlations are plotted in the heatmap. The correlations between all variables are reported in Figure S12.

4. Discussion

4.1 Soil organic carbon accrual is unrelated to low microbial activity at high moisture

We found that SOC and TN increased with increasing long-term soil moisture in the fallow plots investigated in this study (Figure 1). This is consistent with previous work on this experimental site, which reported a similar trend, averaged over several cropping systems (Das et al., 2018). Here, we focused on the fallow plots to eliminate potential effects of the cropping system to allow a direct investigation of the effects of long-term soil moisture on SOC dynamics. Enhanced SOC accumulation can result from higher inputs, lower outputs, or both. Since aboveground biomass, belowground biomass (Das et al., 2018), and the amount of initially decomposing plant litter (Figure 1C) were similar across the moisture gradient, we argue that SOC accumulation resulted from reduced SOC output per unit C input.

While accumulation of SOC on seasonally saturated soils has been explained in terms of limited mineralization (O’Brien et al., 2010), our evidence suggests that this was not the case here. Measurements indicate that the soil moisture in the wettest sites was lower than 80% water filled pore space, i.e. less than saturated, in 82% of the measurements taken over a period of four years, from 2011 to 2014 (Das et al., 2019). A moisture level corresponding to 80% water filled pore space can be expected to decrease heterotrophic respiration by approximately 10% from the maximum level (Moyano et al., 2012). Although
we do not have detailed water content data for November through March, the existing data (and our field observations) indicate that soil moisture is the greatest in these winter months. Since SOC mineralization is expected to be the lowest during the winter regardless of water content, saturation-driven limitation on mineralization were likely not important. Furthermore, if limited mineralization at high moisture had been a substantial driver, one might expect an impact on the decomposition trajectory of plant inputs to POM and MAOM fractions, resulting in a greater proportion of SOC remaining in the less processed POM fractions. The increase in oPOM and MAOM amounts with increasing moisture can be attributed to higher contents of silt- and clay-sized particles, and their tendency to aggregate (Fig 1). However, our data show that the distribution of C (per unit SOC) among these fractions is similar with regards to long-term moisture (Figure S2), indicating that organic matter mineralization followed a similar (and expected) decomposition trajectory in all the soils studied. We therefore suggest that the principal driver of SOC accumulation were differences in SOC transformation and cycling, which was mediated through microbial metabolism and organo-mineral interactions, and not because of oxygen limitation in transient saturated conditions that coincide with low mineralization during the winter months.

4.2 High long-term moisture increased microbial conversion of organic inputs into MAOM

Our results show a strong response of microbial biomass and DOC to differences in long-term soil moisture (Figure 2). Decreasing DOC concentrations and increasing microbial biomass C (per unit SOC) with increasing long-term moisture may have been the result of greater microbial oxidation of DOC, leading to enhanced interactions with minerals surfaces (Lehmann and Kleber, 2015), and ultimately to a decrease in DOC concentration. This hypothesis is supported by decreasing MAOM C:N values with increasing moisture, indicating that they consisted of more processed and oxidized microbial products. Although microbial biomass was measured only once, we assume that this finding reflects a more general trend, as reflected by the different C:N ratios of the MAOM fractions.
Further insights into the chemical composition of MAOM fractions across the moisture gradient was provided by spectroscopic techniques. MAOM fractions from high-moisture soils were enriched in proteinaceous compounds, oxidized carboxylic-C and O-alkyl-C functional groups, and depleted in alkyl-C and aromatic-C functional groups (Figure 4, Table S3, Figure S9, Figure S10). These results indicated that MAOM of high-moisture soils contained more products of microbial decomposition of aromatic and aliphatic plant residues, with concomitant enrichment in microbial proteinaceous components (Figure S10). The possible role of mineral composition in shaping SOC composition is discussed below. Taken together, our results clearly show that high long-term moisture led to greater oxidation of SOC that explains its accumulation.

4.3 Long-term moisture shaped mineral composition and C interactions with minerals

Our results reveal substantial differences in Fe phases across a narrow range of moisture conditions in which reducing conditions are a transient occurrence, and unsaturated conditions predominate. Higher long-term moisture was associated with a 7% relative decrease in FePY, which comprises organic-Fe complexes and nanocrystalline Fe components (Thompson et al., 2011), and a 10% relative increase in crystalline FeDITH (Figure 3), resulting in increasing crystallinity of Fe phases. Previous work has shown that wet-dry cycles, which can induce reducing and oxidizing conditions, may cause reduction of Fe in non-crystalline phases and subsequent increase in Fe crystallinity (Thompson et al., 2006). Similar observations have been reported for an upland soil experiencing substantial precipitation events (Hodges et al., 2019). Al phases were not affected by moisture (Figure 3A) because they are pH-dependent but not redox-dependent (Bertsch and Bloom, 2018).

Although SOC content has previously been shown to correlate with the amounts of non-crystalline Fe oxides (Mikutta et al., 2006), surprisingly, DOC extracted from different oxide pools did not significantly correlate with Fe content in each pool (Figure 3D). This suggests that OC contents associated with oxide minerals did not consistently follow the shift in Fe phases along the moisture gradient. However, our results clearly show the increasing importance of AlPY and AlHH phases for SOC retention with increasing
long-term moisture (Figure 3C, 3D, and 3E). Several recent studies have similarly observed a greater
contribution of non-crystalline Al pools for SOC retention, albeit with greater saturation extent and
frequency (Possinger et al., 2020) or annual higher precipitation (Inagaki et al., 2020). Solubilization of
Fe\textsubscript{PY}-C complexes in high moisture soils and subsequent interaction of C with Al\textsubscript{PY} and Al\textsubscript{HH}, as already
appreciated in soils experiencing reducing conditions (Huang and Hall, 2017; Possinger et al., 2020), was
likely the reason for the increasing importance of these Al phases in storing SOC. Indeed, a recent analysis
of data from 2574 mineral horizons from National Ecological Observatory Network sites across North
America showed that oxalate extractable non-crystalline Al was a better predictor of SOC storage than
oxalate extractable Fe (Yu et al., 2021), suggesting that this effect was more common than previously
thought. Taken together, our results indicate that SOC stabilization by Al\textsubscript{PY} and Al\textsubscript{HH} was directly
influenced by long-term moisture levels that are not commonly associated with a loss of Fe\textsubscript{PY}.

There is also a possibility (which we did not test) that the chemical composition and mode of
interaction of SOC associated with Al differs from that of Fe, and that it may vary across soil moisture.
Considering the mounting evidence of the critical role of Al in SOC stabilization at a wide range of
moisture conditions, we suggest that future research should seek to understand the composition of SOC
associated with non-crystalline Al, and the processes that influence the persistence of this important SOC
pool.

Higher long-term moisture was also positively correlated with Ca\textsubscript{EX} and pH (Figure S12), and in turn
Ca\textsubscript{EX} was highly correlated with SOC ($r = 0.67$), suggesting that increased Ca\textsubscript{EX} in high-moisture soils
contributed to SOC accrual. The positive correlations between moisture, Ca\textsubscript{EX}, and pH can be explained by:
(1) dissolution of subsoil carbonates and upward migration of Ca\textsuperscript{2+} ions facilitated by capillary rise in the
high moisture soils (Ap horizon soils in this study did not have measurable calcium carbonate contents)
(Das et al., 2019), and/or (2) higher contents of clay minerals containing Ca\textsubscript{EX} due to lateral flow and
deposition of fine-grained particles in shallower and wetter locations in the field (Das et al., 2018). This
resulted in greater SOC retention in high Ca\textsubscript{EX} soils. Thus, moisture affected SOC accrual through an increase in Ca-bearing fine particles.

### 4.4 Long-term moisture regulates SOC accrual through microbial activity and mineral composition

High long-term soil moisture enhanced microbial oxidation of organic inputs and concomitantly influenced the distribution of oxide mineral phases and content of Ca\textsubscript{EX}-bearing minerals, which in turn interact with the oxidized organic compounds (Figure 6 and Table 1). This complex interaction of soil moisture with biotic and abiotic soil processes was pivotal for SOC accrual in the high moisture soils (Figure 1). These conclusions are consistent with and extend the MEMS framework (Cotrufo et al., 2013) which explains SOC dynamics as an interaction between microbial efficiency for processing organic inputs, and the mineral component capacity to stabilize these processed inputs.

Microbial efficiency can potentially explain the association between long-term soil moisture, microbial biomass, and SOC accrual. High moisture reduces diffusional constraints which could have increased microbial carbon use efficiency (Moyano et al., 2013; Butcher et al., 2020), and resulted in more C taken up by microbes per unit C input into the soil. Such diffusional constraints may have indeed caused the inverse relationship between long-term moisture and DOC concentration (Figure 2). Our observations suggest that the dependence of carbon use efficiency on soil moisture can result in SOC accrual at field scale. However, more work is needed to understand how to manage soil moisture for optimized carbon use efficiency and increased SOC stocks.

The superposition of microbial activity and an abundance of Ca\textsubscript{EX} acted to enrich high moisture MAOM fractions in carboxylate-rich compounds. The contribution of Ca\textsubscript{EX} for SOC content and the prevalence of carboxylic C in MAOM fractions (Figure 4, Table S3, Figure S10) indicate that clay-Ca-COO\textsuperscript{-} interactions were an important mechanism for SOC protection in the high moisture soils (Mouvenchery et al., 2012; Chen et al., 2014; Rowley et al., 2018; Wan et al., 2021). These interactions are expected to prevail in the higher pH values found in the high-moisture soils which, since high pH promotes electrostatic interactions between Ca\textsubscript{EX} and carboxylates (Rowley et al., 2020). The insignificant
correlation between MAOM and SOC contents indicate that specific effects of Ca\textsubscript{EX}, rather than greater contents of silt- and clay-sized particles, played a role (Table 1, Figure 6, and Figure S12). As highlighted recently, mineralogical composition is a better predictor than particle size distribution for SOC content (Khomo et al., 2017; Rasmussen et al., 2018; Singh et al., 2018). While it has been suggested that Ca and non-crystalline Fe can form ternary structures with carboxylate groups (Sowers et al., 2018), our soils showed a negative correlation between Fe\textsubscript{PY} and Ca\textsubscript{EX} (Figure S12), implying that ternary Ca-Fe-C structures were less important for SOC accrual in our study. Long-term moisture also increased the role of Al in SOC storage (Figure 6, Table 1). As non-crystalline Fe was reduced under transient saturation events and leached out of the surface soils, the importance of non-crystalline Al phases for storing C increased (Figure 3C), even at a lower moisture range in comparison to other studies (Inagaki et al., 2020; Possinger et al., 2020). The composition of compounds that preferentially adsorb to non-crystalline Al and Al-organic complexes is still unclear, but the available information on podzolic soils suggests that Al\textsubscript{PY} may interact with carboxylate-rich compounds (Parfitt et al., 1999; Possinger et al., 2020).

5. Conclusions

In the trade-off between mineralization and accrual, high long-term moisture increased SOC accrual, even at a range of soil moisture not expected to constrain decomposition or alter soil oxide mineral composition, indicating that such changes are likely to occur even along moderate soil moisture gradients. The direction and magnitude of changes to SOC stocks under changing long-term soil moisture are likely to be dependent on additional soil properties that affect microbial conversion of organic inputs and mineral interaction of these products. For example, soils which do not have the mineral capacity to interact with oxidized organic compounds may not benefit from increases to moisture. Climate change is not only changing soil moisture, but critically, also altering the magnitude and temporal patterns of moisture variability through extreme floods and drought. The relative roles of mean long-term moisture vs. moisture temporal variability on stabilization mechanisms are still not clear. Currently, conventional C models consider soil moisture as a physical variable only, and models that account for microbial traits do not
include the potential effect of moisture on carbon use efficiency. These processes will have to be clarified to maintain SOC stocks by driving forward stabilization of organic inputs.

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Supplementary Material for

Higher Long-Term Soil Moisture Increases Organic Carbon Accrual
Through Microbial Conversion of Organic Inputs

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| Plot | Subplot | Moisture level [Q1=driest] | Normalized moisture value |
|------|---------|---------------------------|---------------------------|
| A    | 1       | Q3                        | 1.037                     |
|      | 2       | Q5                        | 1.387                     |
|      | 3       | Q2                        | 1.034                     |
|      | 4       | Q4                        | 1.196                     |
|      | 5       | Q1                        | 0.936                     |
| F    | 1       | Q1                        | 0.869                     |
|      | 2       | Q3                        | 1.072                     |
|      | 3       | Q5                        | 1.206                     |
|      | 4       | Q2                        | 0.999                     |
|      | 5       | Q4                        | 1.106                     |
| I    | 1       | Q2                        | 0.850                     |
|      | 2       | Q3                        | 1.040                     |
|      | 3       | Q4                        | 1.062                     |
|      | 4       | Q5                        | 1.200                     |
|      | 5       | Q1                        | 0.727                     |
| P    | 1       | Q3                        | 0.990                     |
|      | 2       | Q4                        | 0.990                     |
|      | 3       | Q5                        | 1.224                     |
|      | 4       | Q2                        | 0.930                     |
|      | 5       | Q1                        | 0.805                     |
Table S2. Results from elemental and isotope characterization of plant biomass and soil density fractions. The average value is followed by the standard deviation (± S.D.) in parentheses. The level of significant from the 2-way ANOVA model of fraction and moisture is reported as *** P < 0.001, or non-significant (ns).

| Soil fraction/biomass | C:N       | $\delta^{13}$C (%) | $\delta^{15}$N (%) |
|-----------------------|-----------|--------------------|--------------------|
| Roots                 | 50.21 (7.5) | -28.62 (0.5)       | 3.50 (1.2)         |
| Shoots                | 38.86 (6.1) | -28.11 (0.3)       | 1.97 (1.2)         |
| fPOM                  | 29.97 (3.0) | -28.37 (0.5)       | 4.39 (1.9)         |
| oPOM                  | 22.88 (2.2) | -28.80 (0.4)       | 3.81 (1.4)         |
| MAOM                  | 9.20 (0.19) | -27.35 (0.3)       | 5.29 (0.7)         |
| Bulk soil             | 10.31 (0.4) | -27.53 (0.3)       | 5.82 (3.2)         |

Source of variance

|                | Fraction | Moisture | Fraction*Moisture |
|----------------|----------|----------|------------------|
|                | ***      | ***      | ns               |
| Moisture       | ns       | ***      | ns               |
| Fraction*Moisture | ns     | ns       | ns               |

Table S3. $^{13}$C-NMR integral ratios and C-NEXAFS peak height ratios of MAOM fractions from composited samples of low, mid, and high moisture levels.

| Moisture level | carboxyl-C/aromatic-C | $^{13}$C-NMR | carboxyl/O-alkyl-C | O-alkyl-C/alkyl-C |
|----------------|-----------------------|--------------|--------------------|------------------|
|                | C-NEXAFS              | $^{13}$C-NMR | $^{13}$C-NMR       | $^{13}$C-NMR     |
| Low            | 1.349                 | 0.533        | 0.262              | 1.454            |
| Mid            | 1.363                 | 0.594        | 0.286              | 1.928            |
| High           | 1.429                 | 0.615        | 0.290              | 2.063            |
Figure S1. Soil volumetric water content from 2011 to 2018. Each data point represents a mean of time domain reflectometry (TDR) measurements ($n = 3$). Data from Das et al. (2019)
Figure S2. Organic carbon content in free particulate, occluded particulate, and mineral associated organic matter fractions ([fPOM, oPOM, and MAOM, respectively], relative to total SOC, as a function of normalized soil moisture values. Regression models were non-significant.
**Figure S3.** C:N ratio of plant biomass, and free particulate, occluded particulate, and mineral associated organic matter fractions ((fPOM, oPOM, and MAOM, respectively) as function of normalized soil moisture values. Significant regression models are shown with regression lines, regression coefficients, and P values.
Figure S4. $\delta^{13}C$ values of plant biomass, and free particulate, occluded particulate, and mineral associated organic matter fractions (fPOM, oPOM, and MAOM, respectively) as function of normalized soil moisture values. Significant regression models are shown with regression lines, regression coefficients, and P values.
**Figure S5.** δ¹⁵N values of plant biomass, and free particulate, occluded particulate, and mineral associated organic matter fractions (fPOM, oPOM, and MAOM, respectively) as function of normalized soil moisture values. Significant regression models are shown with regression lines, regression coefficients, and P values.
Figure S6. Al/Fe ratio in different oxide extractions as a function of normalized soil moisture values. Significant regression models are shown with regression lines, regression coefficients, and P values.
Figure S7. Total dissolved organic carbon (DOC) normalized to mass of soil organic carbon (SOC) as a function of normalized soil moisture values. Regression model was non-significant.
Figure S8. ATR-FTIR spectra in the range of 2500 – 4000 cm$^{-1}$ (A) and in the range of 550 – 1840 cm$^{-1}$ (B) of bulk soils and MAOM fractions ($n = 20$) and fPOM and oPOM fractions ($n = 3$) averaged across all long-term normalized moisture values.
Figure S9. $^{13}$C NMR spectra of fPOM, oPOM, MAOM, and bulk soils from composited sampled from low, mid, and high moisture levels.
Figure S10. Molecular mixing model results for $^{13}$C-NMR analysis. Relative amount of C in different molecular components for each soil fraction and normalized moisture level.
Figure S11. C K-edge NEXAFS spectra of fPOM, oPOM, and MAOM fractions from low, mid, and high moisture levels.
Figure S12 Heatmap of Pearson correlations (r).