Mapping Soil Organic Carbon in Wildfire-Affected Areas of the McKenzie River Basin, Oregon, USA

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June 16, 2023

Abstract

Large-scale wildfires are increasing in frequency and are likely to become more severe under future Pacific Northwest climate scenarios. The effects of wildfires on soil organic carbon (SOC) remain difficult to estimate because soil heterogeneity limits generalizations. We mapped fired severity within the footprint of the Holiday Farm Fire (McKenzie River, Oregon, 2020) and sampled a burn severity gradient (unburned, low, high) in a detailed scheme to account for inter- and intra-site variation (20 soil profiles/half-hectare for burned sites, 9/hectare for unburned) at three depths (0-2 cm, 2-20 cm, 20-40 cm). We measured total SOC, mineral-associated organic carbon (MAOC), particulate organic carbon (POC), and pyrogenic carbon (PyC). We found significant SOC differences in the high severity fire in most carbon pools and depths, with the largest total SOC decrease of 6.48% (56% change) in 0-2 cm. Compared to unburned, the low severity site had higher MAOC (0-2 cm: +0.48%, 22% change; 2-20 cm: +0.28%, 17% change) and significantly lower POC (0-2 cm: -5.12%, 54% change; 2-20 cm: -1.73%, 48% change). We found lower PyC in burned sites, indicating combustion of this pool. SOC stocks at 0-20 cm were higher in low severity (total SOC: +7.45 kg/m2, 71% change; MAOC: +4.81 kg/m2, 153% change) compared to unburned. There was remarkable variation within each site, but the consistent high levels of MAOC in low severity area support prescribed burning as a technique to mitigate wildfire risk while limiting losses or increasing SOC compared to high severity fires.

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10	Key Points						
11	1. Total soil organic carbon declined with increasing fire severity.						
12	2. Mineral associated organic carbon was higher in low severity areas compared to						
13	unburned or high severity areas.						
14	3. Particulate and pyrogenic carbon were highly variable in low severity areas and						
15	less variable under high severity or unburned conditions.						
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1.73%, 48% change). We found lower PyC in burned sites, indicating combustion of this
pool. SOC stocks at 0-20 cm were higher in low severity (total SOC: +7.45 kg/m², 71%
change; MAOC: +4.81 kg/m², 153% change) compared to unburned. There was
remarkable variation within each site, but the consistent high levels of MAOC in low
severity area support prescribed burning as a technique to mitigate wildfire risk while
limiting losses or increasing SOC compared to high severity fires.

36 Plain Language Summary

37 Forest fires are becoming more common in the Pacific Northwest of the United States 38 because of droughts and rising temperatures. To understand how fire influences forest 39 soil, we mapped fire severity within the footprint of the Holiday Farm Fire, which burned 40 in western Oregon in 2020. We then measured differences in soil organic carbon within 41 and across sites representing fire severity gradient. We sampled soils at unburned, low 42 severity, and high severity sites and found that the total soil carbon content was lowest in 43 highly burned areas. However, soil properties were highly variable on a small scale, 44 within sites, and some types of soil carbon had the highest amount in the low severity 45 sites. Mineral-associated carbon were higher in lightly burned than in highly burned and 46 unburned site. We suggest that low severity controlled burns may be a way to manage 47 future wildfires while increasing mineral-associated carbon pools in the region.

48 **1. Introduction**

49 The Pacific Northwest (PNW) of the United States is experiencing increases in 50 fire frequency due to changes in temperature and precipitation patterns (Halofsky et al., 51 2020). Wildfires are an integral part of forest ecosystem dynamics in the PNW, but 52 catastrophic fires have become increasingly frequent and severe, especially in working

53	landscapes where intensive forest management can amplify climatic stress (Zald & Dunn,
54	2018). In recent years, the record-breaking effects of climate change and fire disturbance
55	on PNW forests also impacted people's livelihoods near and far from fire-affected
56	landscapes (Higuera & Abatzoglou, 2021; Weisberg, 2009), in some cases, releasing
57	carbon to the atmosphere at rates that could offset the benefits of management for carbon
58	sequestration (Jerrett et al., 2022). As carbon sequestration projects continue to be
59	incentivized in landscapes, fire seasons are becoming longer and warmer, and rainy
60	winters are becoming shorter and drier; new research is needed to inform conservation
61	and management plans for PNW "forests of the future" (Case et al., 2021).
62	It is well known that PNW forests hold large quantities of soil organic carbon
63	(SOC), and the response of SOC to wildfire is expected to have major implications for
64	carbon sources and carbon sinks throughout the region (Nave et al., 2011; Pellegrini et
65	al., 2022). However, fire severity and carbon sequestration are still assessed primarily
66	from the perspective of impacts on vegetation. Soils and SOC are variable at the
67	landscape level even under the homogeneous vegetation cover; therefore, site-specific
68	and geographically-focused research is needed to understand how climate change and
69	wildfires affect ecosystem carbon balance across landscapes (Loescher et al., 2014). To
70	this end, we designed a spatially explicit approach to quantify fire effects on SOC at the
71	landscape level, which could be replicated across the PNW to determine climate change
72	effects on a regional scale to inform policy and management. The goal of this project is to
73	map the spatial distribution of SOC in unburned, low severity, and high severity sites
74	from a recent catastrophic wildfire in a typical PNW working landscape, in the McKenzie
75	River basin. By combining field sampling techniques with GIS mapping and

76	quantification of fire severity impacts on carbon pools across the soil profile, we infer the
77	transformation and movement of SOC across the landscape influenced by fire severity.
78	Our objectives for this post-fire McKenzie River landscape study can be summarized in
79	the following research questions:
80	(Q1): What are the percentages and stocks of different soil organic carbon pools
81	at the varying burn severities in one-year post-fire conditions?
82	(Q2): What is the spatial and depth variation of the effects of wildfire on soil
83	organic carbon?
84	To answer these questions, we partitioned SOC into three pools: mineral-
85	associated organic carbon (MAOC), particulate organic carbon (POC), and pyrogenic
86	carbon (PyC). MAOC is the organic matter that has been adsorbed onto mineral surfaces,
87	such as clays, or in physically protected structures, such as soil micro-aggregates (Kögel-
88	Knabner et al., 2008). It is characterized by its long-term persistence in the soil and heavy
89	fraction density (Lehmann & Kleber, 2015), compared to more labile forms of light
90	fraction POC (Lavallee et al., 2020). Mapping the distributions of unstable POC and
91	stable MAOC is crucial for understanding the persistence of carbon in the soil on both
92	short and long timescales. Recent studies estimated that the MAOC fraction of total SOC
93	has turnover times up to 1000 times longer than POC. Thus, enhancing MAOC formation
94	may be a key to lasting soil carbon sequestration (Georgiou et al., 2022.). However,
95	severe fires can cause major losses of MAOC and POC to the atmosphere, or transform
96	large fractions of those pools into PyC, which is the carbon that is found in charred
97	biomass, charcoal, and soot (Santín et al., 2016).

98	PyC does not have physical or chemical characteristics that can be grouped with
99	POC or MAOC (Lavallee et al., 2020) and is less predictable than the other pools of SOC
100	in terms of persistence and stability. In some cases, PyC formation can be derived mostly
101	from POC and occasionally from MAOC (Bowring et al., 2022). In all cases, regardless
102	of its source, PyC is expected to represent a significant fraction of the total SOC carbon
103	pool after wildfires, and therefore it is important to include PyC, as well as MAOC and
104	POC, in the analysis carbon losses following fire disturbance (Santín et al., 2016). This is
105	particularly important when SOC is considered as a form of climate change mitigation.
106	As a problem of pattern and scale, enhancing SOC as a mitigation tool requires more
107	research because the risk and severity of fire disturbance on SOC stocks are difficult to
108	generalize, even when the impacts on vegetation are obvious. For example, one current
109	hypothesis for the PNW is that increasing wildland fire would significantly decrease soil
110	carbon storage, but low severity burns could minimize the risk and severity of fire
111	impacts on SOC (Nave et al., 2022), with the production of PyC through pyrolysis, which
112	can increase relative to baseline levels (Pingree & DeLuca, 2018), or be consumed and
113	released back to the atmosphere after catastrophic fires (Miesel et al., 2018; Reisser et al.,
114	2016).

115 **2. Methods**

116 2.1 Study sites

117 Our study site is the McKenzie River basin of the Willamette Valley, Oregon, 118 USA. The study region encompasses typical working landscapes of the PNW, where 119 different types of conservation and management strategies co-exist across steep 120 topographic gradients. The McKenzie River basin is dominated by Douglas-fir

121	(Pseudotsuga menziesii) forests - with other common species including western hemlock
122	(Tsuga heterophylla), incense cedar (Calocedrus decurrens), and red alder (Alnus rubra)
123	- under a natural fire rotation that is estimated to range between ~ 160 years for the pre-
124	settlement period (1550–1849) to \sim 500 years for the recent fire-suppression period
125	(Weisberg, 2009). In September 2020, this region experienced the Holiday Farm Fire, a
126	catastrophic wildfire and at over 170,000 acres, one of the largest forest fires in Oregon's
127	history (InciWeb, 2020). The burn severities in terms of canopy cover mortality of the
128	sampling sites are shown in Figure 1. The site locations for our soil severity assessments
129	are also depicted on the map on the western side of the McKenzie River region. The
130	study sites occur at elevation between 300 and 400 meters, annual precipitation around
131	1700 mm, winter temperatures averaging 4 °C, and summer temperatures averaging 18
132	°C (Sproles et al., 2013).





Figure 1. Map of burn severities from the Holiday Farm Fire based on Basal Area Mortality from Rapid Assessment of Vegetation Condition after Wildfire. This method uses change detection analysis from Landsat and similar satellite imagery, wherein a Relative Differenced Normalized Burn Ratio (RdNBR) image is created by subtracting post-fire imagery from pre-fire imagery (USDA Forest Service & Geospatial Technology and Applications Center, 2020). Most of the wildfire was high severity, where more than 75 percent of biomass was burned.

142 2.2 Sampling Design

143

We collected soil samples in September 2021, one year after the fire occurred. We

- 144 estimated severity based on biomass loss (canopy cover mortality), wherein the low
- severity site had canopy cover loss of 0-25 percent and the high severity site had canopy
- 146 cover loss of 75-100 percent. We used a gridded approach to take soil samples from one
- 147 low severity site and one high severity site. The low severity site is located at the

McKenzie River Discovery Center (44.142562, -122.607505), and the high severity site
is near Finn Rock Landing (44.142265, -122.36458), owned by the McKenzie River
Trust. As a baseline control, we used samples collected in 2018 using similar methods
from an unburned old-growth forest site at the HJ Andrews Experimental Forest
(44.26706, -122.17198) (Farinacci, 2020).

At the three sampling locations, soils are categorized as Inceptisols with predominantly silty and sandy loam properties. The sampling landscapes have a gentle slope ranging from ~3 to ~5 percent inclination. Previous studies have estimated SOC stocks for the McKenzie River basin between 8 and 25 kg/m² (Nave et al., 2022;

157 Walkinshaw et al., 2021).

158 At the low and high severity sites, we sampled a total of 20 profiles per 0.5 159 hectare. After removing the litter layer, we sampled three soils per profile at depths 0-2 160 cm, 0-20 cm, and 20-40 cm depth. In addition to soil samples for carbon content analysis, we collected plant litter in 25 by 25 cm^2 around each of the soil profiles before soil 161 162 collection and determined bulk density at each soil depths using three profiles per site, 163 where the average was used to estimate SOC stocks. For bulk density, we collected soil samples by pressing 100 cm³ rings into each layer of a sequentially dug soil layer, 164 165 preventing soil compaction. At the unburned reference site, we took samples in a three-166 by-three grid, where each row spanned 100 meters, for a total of nine profiles per hectare. 167 We sampled soils at depths 0-20 centimeters and 20-40 cm and bulk density for each row, 168 using the same approach explained above for litter and soil collection from the 0-2 cm 169 depth (2022).

170 2.3 SOC Processing

171	In the laboratory, we dried soil samples at room temperature after which the fine					
172	roots were separated by hand and samples were ground and homogenized. We measured					
173	total SOC and SOC pools (MAOC, POC, and PyC) through combustion using a Thermo					
174	Scientific FlashSmart Elemental Analyzer (Waltham, MA, USA). In all cases, we					
175	determined percent concentration and estimated total stocks for each SOC fraction on a					
176	mass basis using standard methods and international units (see SI for full dataset).					
177	To separate the heavy density fraction (MAOC) and the light density fraction					
178	(POC), we used methods developed in previous density separation studies (Pierson et al.,					
179	2021; Sollins et al., 2006), where 10 g of each sample and 10 mL of sodium					
180	polytungstate (density of 1.85 g/cm ³) were shaken for 2 hours and centrifuged for 10					
181	minutes at 3000 revolutions per minute.					
182	We filtered each sample three times with sodium polytunstate and three times					
183	with distilled deionized water. We verified that the sum of the MAOC and POC values					
184	were close to the separately measured total SOC.					
185	We measured SOC concentrations (% mass) using the equation 1 for the MAOC					
186	and POC pools and equation 2 for the PyC pool:					
187	$C_{\text{fraction}} (\% \text{ C}) = \frac{\text{weight of fraction (g)}}{\text{weight of total sample (g)}} * \% \text{ C} $ (1)					
188	$C_{PyC} (\% C) = \frac{\text{post-digestion mass } (g)}{\text{pre-digestion mass } (g)} * \% C $ (2)					

189 where % C is the measured percent carbon from the elemental analyzer before 190 correction. We calculated stocks (kg/m²) using the equation 3 (Villarino et al., 2017): 191 C stocks (kg/m²) = [soil depth (m)] * [bulk density (g/cm³)] * [% C * 10 (g/kg)] (3)

192	We used a weak acid-peroxide digestion to separate pyrogenic carbon from the
193	total sample (Kurth et al., 2006). We added 5 mL 1 M HNO3 and 10 mL 30% $\rm H_2O_2$ to 0.4
194	g soil sample, which we then covered with aluminum foil, placed in 90° C heat bath for
195	16 hours, and filtered the remaining soil in each sample to isolate pyrogenic carbon.
196	We calculated stocks at the 0-20 cm and 20-40 cm depths because bulk density
197	samples were collected at these depths; we added stocks from the 0-2 cm and 2-20 cm
198	depths to yield the 0-20 cm depth. The >2 mm fraction of the soil sample was small
199	enough to ignore in the C stocks calculation. No carbonates are present in regional
200	bedrock; therefore, hydrolysis of carbonate minerals and inorganic carbon inputs are not
201	a plausible source of variation in total carbon across our sites, which we verified with soil
202	pH values (values for the three sites ranged between 5.08 and 6.4) (Swanson & James,
203	1975). For bulk density, we weighed the soil samples after drying them in an oven at 70
204	$^{\circ}$ C for 2 days and divided by the volume of the collection cylinder. Litter stocks were
205	calculated by dividing litter biomass (g/cm ²) by 2, under the assumption that biomass is
206	composed of 50 percent carbon, and converted to units (kg/m^2) matching the SOC stocks
207	(Houghton et al., 2009).

208 2.4 Spatial Data Analysis and Interpolation

Interpolated maps of the four pools of SOC for each site were created using ArcGIS Pro 2.7.0 by Esri (Redlands, CA, USA). For each SOC pool, site, and depth, the inverse distance weight spatial interpolation technique was used to visualize the variation of SOC data (Almasi et al., 2014; Robinson & Metternicht, 2006). Standard parameters were used, wherein power = 2, minimum neighbors = 10, and maximum neighbors = 15. *2.5 Statistical Analysis* Due to the non-normal distribution of the data, we log transformed SOC percents and stocks to meet the assumptions of statistical analysis. We used planned contrast ANOVA tests of means to determine significant differences between carbon percentage and stock of each burn severity class within the same SOC pool and depth (Huang et al., 2023). We used p-values of less than 0.05 to determine significance and noted values between 0.05 and 0.10.

221 2.6 Data Availability

Data archiving is underway at a permanent online repository. The data used for all tables and figures shown below will also be included as Supplementary Information (SI) upon publication of this manuscript.

225 **3. Results**

226 3.1 SOC Concentrations and Stocks

To demonstrate the importance of a gridded sampling scheme for sampling soil properties, where high variation is common, Figure 2 shows histograms of each SOC pool. In all the SOC pools, the high severity site has the highest frequency of lower values. The unburned and low severity sites show similar frequency distributions.



- Unubrned - Low Severity - High Severity



235 We found significant differences between total SOC concentrations and stocks in 236 response to fire intensity and soil depth. Figure 3 shows the median percents of SOC 237 pools along the fire severity and depth gradient, and values discussed below are in mean 238 differences. Total SOC percent decreased significantly along the burn severity gradient, 239 from $\sim 10\%$ to 5% on average, in the topsoil of unburned and high severity fire-affected 240 profiles, respectively. The percent SOC in the unburned and low severity sites were 241 statistically similar across more than half of the comparisons across carbon pools and 242 depths, and we found a similar pattern, although with lower SOC losses, up to 40 cm 243 depth. 244 The effects of burning on the individual pools were more variable than on SOC.

245 MAOC was significantly lower in the low severity site than the high severity site (-1.08%

246 MAOC at 0-2 cm, 40% change; -0.44% MAOC at 2-20 cm, 22% change) and increased in the top two depths between the unburned site and the low severity site (+0.48%)

248 MAOC at 0-2 cm, 22% change; +0.28% MAOC at 2-20 cm, 17% change). The unburned

249 MAOC and low severity MAOC were statistically similar at each depth. POC decreased

250 significantly in the 0-2 cm (-5.12% POC, 54% change) and 2-20 cm (-1.73% POC, 48%

- change) depths between unburned and low severity but was otherwise minimally changed
- between treatments.

PyC decreased along the burn gradient and by depth. The unburned and low severity sites showed similarity in the topsoil and 20-40 cm depth and were both significantly higher than the high severity site at every depth. In the 2-20 cm depth, each site was statistically different from one another, with decreases compared to the unburned site of 0.61% PyC (58% change) in the low severity fire and 0.91% PyC (85% change) in the high severity fire.





Figure 3. Boxplots of SOC percent. Letters above plots correspond to significant differences between sites (i.e. fire severity treatments) at any given depth. Note the differences in scale for each organic carbon pool.

264 265 266 N = 9 for each depth of unburned (27 total), n = 20 for each depth of low severity (60 total), and n = 20 for each depth of high severity (60 total). See Table 1 in SI for corresponding values.

266 267	Figure 4 shows the breakdown of total SOC as proportions of MAOC and POC.
268	In the unburned site, POC holds a larger percentage than MAOC in all depths, though the
269	differences become smaller by the 20-40 cm depth. The low severity and high severity
270	sites show smaller differences in MAOC and POC proportions and converging at
271	shallower depths than the unburned site. In the high severity 20-40 cm depth, MAOC
272	holds a larger percentage of total SOC than POC.
273	In the topsoil, POC proportion is significantly higher in the unburned and high
274	severity sites than the low severity site, while MAOC proportion is significantly higher in
275	the burned sites than the unburned site. PyC is significantly lower in the high severity site
276	than the unburned and low severity sites. For the 2-20 cm depth, POC proportion is
277	significantly highest in the unburned site and significantly lowest in the low severity site,
278	both of which are statistically significant from the high severity site. Proportion of
279	MAOC is significantly lowest in the unburned site than the low and high severity sites,
280	and PyC proportion decreases significantly along the burn severity gradient. In the lowest
281	depth, proportions of POC decreased with increasing burn severity. MAOC proportions
282	were similar between unburned and low severity and was significantly higher in the high
283	severity site. PyC proportions were similar between unburned and low severity and was
284	significantly lower in the high severity site.



285 Depth (cm) 286 Figure 4. Proportion of total SOC that is composed of each fraction (MAOC, POC, and PyC). Note that the 287 proportions of the three fractions added together for each site and depth exceed 100 percent due to the 288 overlap between PyC and its foundational components of mainly POC and some MAOC. N = 9 for each 289 depth of unburned (27 total), n = 20 for each depth of low severity (60 total), and n = 20 for each depth of 290 high severity (60 total). See Table 1 in SI for corresponding values.

292 Median SOC stocks by burn treatment and depth are shown in Figure 5, and the 293 values discussed are mean differences. Litter was significantly lower after the high severity burn (-0.03 kg/m², 70% change) and remained similar after the low severity burn 294 295 $(-0.01 \text{ kg/m}^2, 22\% \text{ change})$. At the 0-20 cm depth, total SOC stocks were significantly higher in the low severity fire $(+7.45 \text{ kg/m}^2, 71\% \text{ change})$ and higher in the high severity 296 fire $(+3.25 \text{ kg/m}^2, 31\% \text{ change})$ compared to the unburned site. We found that stocks 297 were higher in the low severity fire in the 20-40 cm depth ($+0.87 \text{ kg/m}^2$, 8.4% change) 298 and lower in the high severity fire (-1.79 kg/m², 17% change). The difference in stocks 299 300 between the low severity site and the high severity site (-2.66 kg/m², 23% change) at the 301 20-40 cm depth is statistically significant.

302	POC stock was higher in the 0-20 cm depth and lower in the 20-40 cm depth
303	along the burn severity gradient, with significantly lower POC stock in the high severity
304	site compared to the unburned in the deeper soils (-1.03 kg/m ² , 21% change).
305	There was statistically higher MAOC stock in the low severity site in the 0-20 cm
306	depth than in the unburned site (+4.81 kg/m ² , 153% change) and the high severity site
307	(+2.52 kg/m ² , 80% change). The 20-40 cm depth had similar MAOC stock. Compared to
308	the unburned site, PyC was significantly lower in the high severity burn at both the 0-20
309	cm depth (-1.37 kg/m ² , 68% change) and the 20-40 cm depth (-0.90 kg/m ² , 66% change)
310	while remaining similar between the unburned and low severity sites.





312 Figure 5. Boxplots of SOC stock along the burn severity gradient. Stocks used the 0-20 cm and 20-40 cm 313 depths because we collected bulk density samples at these depths; we calculated stocks for the 0-20 cm 314 depth by adding the stocks from the 0-2 cm and 2-20 cm depths. Note the differences in scale for each 315 organic carbon pool. Letters indicate significant differences at any given depth between sites along the fire 316 severity gradient. Bulk density values used for total SOC and PyC stock calculations are as follows: 317 unburned (0-20 cm: 0.91±0.36 g/cm³; 20-40 cm: 1.51±0.08 g/cm³), low severity (0-20 cm: 1.95±0.24 318 g/cm³; 20-40 cm: 1.75±0.31 g/cm³), high severity (0-20 cm: 1.85±0.16 g/cm³; 20-40 cm: 1.69±0.07 g/cm³). 319 For bulk density samples, n = 3 for each site. For soil samples, n = 9 for each depth of unburned (27 total), 320 n = 20 for each depth of low severity (60 total), and n = 20 for each depth of high severity (60 total). See 321 Table 2 in SI for corresponding values. 322

323 *3.2 SOC Interpolations*

324	Interpolation maps for each site of the burn severity gradient and SOC pool are
325	shown in Figure 6 When observing the hotspots of SOC in these sites, note that total
326	SOC is composed of the other three carbon fractions (PyC, MAOC, and POC); thus, the

total SOC interpolation maps incorporate variation and accumulation in hotspots for theMAOC and POC carbon fractions combined.

329 The unburned site shows hotspots of POC and PyC in the west section of the site, 330 with minimal MAOC spatial variability. In the topsoil of the low severity region, there 331 were hotspots of MAOC in the north-central and north-east parts of the site, as well as 332 higher percentages of POC in the north-central and central sections. These hotspots 333 persisted throughout the soil profile even as overall SOC percentages decreased. The 334 areas of high accumulation for POC, MAOC, and PyC generally matched in the low 335 severity site. In the high severity site, there was minimal spatial variability in MAOC. 336 There were small hotspots of POC in the eastern parts of the site in the 0-2 cm and 2-20 337 cm depths, that line up with areas of accumulation of PyC in the topsoil, and in the south-338 central section of the 20-40 cm depth.



339 340

Figure 6. Spatially interpolated SOC types for the unburned, low severity, and high severity sites. Note the same color ramp is used across all plots. Number of soil cores in each plot upon which the interpolation is based varied by burn severity (9 in unburned, 20 in low severity, and 20 in high severity). See Table 1 in SI for corresponding values.

346 **4. Discussion**

347	Percent SOC	decreased after	[•] burning in	the maj	ority of	pools in mos	t depths
			0	J	2	1	1

348 (Figure 3). The high severity site showed the lowest SOC in all pools for most depths,

349 with the exception of POC in the 2-20 cm depth. This is consistent with other findings,

- 350 where wildfire causes a decrease in SOC quantity in Pacific Northwest and temperate
- 351 forests (Bormann et al., 2008; Homann et al., 2015; Nave et al., 2011). The low severity
- 352 fire was similar to the unburned site in most carbon pools and depths, pointing to the

resiliency of the soil after low severity burns, which were once common in thisecosystem.

355 As an aboveground process, fire rarely directly affects soil C at depths lower than 356 20 cm as there is little heating of mineral soils deeper than 20 cm (Brady et al., 2022; 357 Heckman et al., 2013). Indeed, we did not find major changes in the 20-40 cm depth. We 358 also note that % C at 20-40 cm depth in the unburned site was comparable to that of the 359 burned plots at 2-20 cm depth, which is consistent with loss of most of the surface-soil 360 POC and subsequent soil compaction. The effect of burn severity on soil C will likely 361 manifest over longer timescales due to processes such as plant succession, pedogenesis, 362 microbial activity, and soil leaching (Nave et al., 2011; Pierson et al., 2021). Future 363 studies that focus on the effects of fire severity on soil carbon pools over longer periods 364 of time than one-year post-fire in the McKenzie River basin could attempt to answer 365 these phenomena.

366 While the total amount of MAOC was similar across unburned and burned at all 367 depths (Fig. 3), the proportion of SOC occurring as MAOC was significantly higher in 368 the burned plots than the unburned control plots (Fig. 4), suggesting that mineral-369 association can protect the soil carbon pool through fire. As partially decomposed plant 370 material, POC is more susceptible to burning than MAOC, especially during high severity fires. In PNW ecosystems where the O horizon is thick and active, forest fires 371 372 cause greater losses to POC than MAOC, leading to shifts in the pool makeup of SOC 373 (Pierson et al., 2021).

Bulk density was higher in the low severity site than the unburned and high
severity sites at each corresponding depth, as is found in other systems (Agbeshie et al.,

376 2022). The shift from higher proportions of POC (light fraction) to equal proportions or 377 higher MAOC (heavy fraction) can explain the higher bulk density in the low severity 378 site (Figure 4). Total SOC stock and MAOC stock in the 0-20 cm depth increased after a 379 low severity burn, driven by bulk density differences rather than concentration. With soil 380 compaction occurring, this may point to evidence of soil organic carbon sequestration 381 through low severity fires in the McKenzie River landscape, which supports other 382 findings where in some cases low severity or prescribed burns mitigates the loss of SOC 383 as compared to a high severity fire (Homann et al., 2011; Pellegrini et al., 2021). 384 The finding of low PyC in the high severity site is somewhat consistent with other 385 findings, where PyC either remained consistent across burn severities or can be 386 consumed by fires (Doerr et al., 2018; Miesel et al., 2018; Reisser et al., 2016). The high 387 severity fire more thoroughly consumed the woody debris and mineral-associated carbon 388 that may have been converted to PyC. In the unburned site, the central-western location 389 of high accumulation of PyC could be understood through the methods. Samples that 390 included high levels of organic material and POC also may contain high levels of 391 recalcitrant carbon that is not pyrogenic in origin, and thus survive the acid-peroxide treatment (Schmidt & Noack, 2000). Additionally, historical fires that occurred in the 20th 392 393 century in nearby areas, while not directly impacting the unburned site, may have dispersed PyC onto the unburned study site (as was common after the 2020 fires), thus 394 395 affecting PyC at that site. 396 The spatial variation of SOC is unique to each site (Figure 6). There are locations

397 of high accumulation of SOC at each of the variably burned sites, especially in the

398 unburned site. These are potentially due to minute topographic variations in the landscape

that could be both relatively long-term, such as minor slope changes, or short-term, such as woody debris or fallen snags in the burned areas. Using a gridded sampling scheme of 20 soil cores in the burned areas and 9 in the unburned site helped to account for the large variation in soil carbon within one site, and these concepts should be considered when working towards generalizing the effects of phenomena, such as wildfire, on a landscape's variable soil properties (Pellegrini et al., 2022).

405 **5.** Conclusion

Wildfire, at both the low and high severity levels, affects soil organic carbon in
the McKenzie River landscape. In the case of this landscape, wildfire decreased SOC in
most SOC pools and depths. Because of the variability of soils even within a single
landscape, generalizations about the effects of wildfire on soil are difficult to make.
Rather, site-specific and localized research is needed to understand connections between
soil and disturbance.

412 We found evidence that low severity fire has the potential to mitigate the loss of 413 SOC as compared to a high severity fire and, in some cases, as compared with 414 undisturbed landscapes. Specifically, we found that mineral-associated C persisted or 415 even increased through low-severity fire, indicating that MAOC can protect soil C stocks 416 from fire. This may point to the soil advantages of prescribed burning in the McKenzie 417 River Basin and other similar montane forest landscapes, where fire suppression and 418 intensive management can lead to catastrophic disturbance and major carbon losses that 419 can offset carbon gains from biomass accumulation. This fire management strategy could 420 prevent the buildup of fuels, reduce the loss of SOC that would occur in a stand-replacing 421 fire, and increase soil carbon stock in some SOC pools. With wildfires increasing in

422	frequency under climate change in the PNW, land managers and policy makers would
423	benefit from considering how fires affect belowground carbon and what conservation and
424	management strategies can preserve or increase SOC storage going forward.
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SUPPLEMENTARY TABLES

SOC Pool	n	0-2 cm	2-20 cm	20-40 cm
SOC				
Unburned	20	11.63 ± 6.74 ^a	$5.14\pm1.07~^{a}$	$3.45\pm0.96^{\text{ a}}$
Low Severity	20	7.32 ± 3.19 ^b	$4.29\pm1.33~^a$	3.22 ± 1.06^{a}
High Severity	9	5.15 ± 1.78 ^c	3.56 ± 1.47 ^b	2.55 ± 1.32 ^b
POC				
Unburned	20	$9.44\pm6.88~^a$	3.61 ± 1.31 ^a	$1.67 \pm 0.64^{\ a}$
Low Severity	20	4.32 ± 2.17 ^b	$1.88 \pm 1.06^{\ b}$	$1.42\pm0.70~^{ab}$
High Severity	9	3.72 ± 1.72 ^b	2.09 ± 1.22 ^b	1.17 ± 1.14 ^b
MAOC				
Unburned	20	$2.21\pm0.65~^a$	1.68 ± 0.23 ^{ab}	1.48 ± 0.37 ^a
Low Severity	20	$2.69\pm1.47~^{a}$	$1.96\pm0.38~^a$	$1.43\pm0.44~^a$
High Severity	9	1.61 ± 0.36 ^b	$1.52\pm0.37~^{b}$	$1.34\pm0.44~^a$
РуС				
Unburned	20	$1.65\pm1.07~^{a}$	$1.06\pm0.44~^a$	$0.45\pm0.15~^a$
Low Severity	20	$0.87\pm0.44~^{a}$	$0.45\pm0.16\ ^{b}$	$0.40\pm0.17~^a$
High Severity	9	0.39 ± 0.32 ^b	$0.15\pm0.07~^{c}$	0.14 ± 0.13 ^b

447 Table 1. Summary of Percent Carbon Values for Each Burn Severity Site and Depth

448 Values are means $\pm 1\sigma$. Letters indicate significant differences at any given depth

449 between sites along the fire severity gradient.

459 Table 2. Summary of Carbon Stocks \pm standard deviation (kg/m²) for Each Burn Severity

460 Site and Depth

SOC Pool	n	0-20 cm	20-40 cm	
Total SOC				
Unburned	20	10.50 ± 2.75 ^a	$10.40\pm2.90~^{ab}$	
Low Severity	20	17.95 ± 2.58 ^b	11.27 ± 3.72 ^a	
High Severity	9	$13.75 \pm 5.30^{\ a}$	8.61 ± 4.44 ^b	
POC				
Unburned	20	7.61 ± 2.98 ^a	$5.04 \pm 1.75^{\text{ ab}}$	
Low Severity	20	$8.32\pm4.02~^{a}$	$4.96\pm2.47~^{\rm a}$	
High Severity	9	$8.35\pm4.33~^{a}$	3.93 ± 3.84 ^b	
MAOC				
Unburned	20	3.15 ± 0.39^{a}	4.47 ± 1.17^{a}	
Low Severity	20	7.96 ± 1.87 ^b	4.98 ± 1.49 ^a	
High Severity	9	5.67 ± 1.28 ^c	$4.52\pm1.47~^{\rm a}$	
РуС				
Unburned	20	2.03 ± 0.73 ^a	1.36 ± 0.46^{a}	
Low Severity	20	$1.91\pm0.68^{\text{ a}}$	1.39 ± 0.58 ^a	
High Severity	9	0.66 ± 0.28 ^b	0.46 ± 0.43 ^b	
Litter				
Unburned	20	0.036 ± 0.015 ^a		
Low Severity	20	$0.028 \pm$	0.015 ^a	
High Severity	9	0.011 ± 0.003 ^b		

462 Values are means $\pm 1\sigma$. Letters indicate significant differences at any given depth 463 between sites along the fire severity gradient.

472 Table 3. Summary of bulk density along the burn severity at each depth, in units g/cm³
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Site	n	0-20 cm	20-40 cm
Unburned	3	0.91 ± 0.36 ^a	1.51 ± 0.08 ^a
Low Severity	3	$1.95\pm0.24~^{\mathrm{b}}$	1.75 ± 0.31 ^{ab}
High Severity	3	1.85 ± 0.16 ^b	1.69 ± 0.07 ^b
Values are means ± 1	σ . Letters indicate	significant differences at	any given depth
between sites along th	ne fire severity grad	lient.	50 1
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512 Table 4. Summary of ANOVA planned contrast tests for SOC percent across the burn

- 513 severity gradient by soil profile depth

			Total SOC	POC	MAOC	PyC
Depth	Comparison	df	р	р	р	р
0-2 cm	Unburned - High Severity	46	< 0.001	< 0.001	0.025	< 0.001
	Low Severity - High Severity	46	0.012	0.352	< 0.001	< 0.001
	Unburned - Low Severity	46	0.021	0.002	0.265	0.142
2-20 cm	Unburned - High Severity	46	0.002	0.002	0.184	< 0.001
	Low Severity - High Severity	46	0.037	0.525	0.001	< 0.001
	Unburned - Low Severity	46	0.111	0.001	0.128	< 0.001
	Unburned - High Severity	46	0.018	0.021	0.324	< 0.001
20-40 cm	Low Severity - High Severity	46	0.021	0.054	0.511	< 0.001
	Unburned - Low Severity	46	0.574	0.415	0.637	0.520

515 *Bold value indicates significant difference of p < 0.05, italicized value indicates

516 significant difference of p < 0.1

- 534 Table 5. Summary of ANOVA planned contrast tests for ratio of SOC fraction to total
- 535 SOC across the burn severity gradient by soil profile depth

			POC	MAOC	PyC
Depth	Comparison	df	р	р	р
0.2	Unburned - High Severity	46	0.179	0.003	0.008
0-2 0m	Low Severity - High Severity	46	0.015	0.300	0.031
CIII	Unburned - Low Severity	46	0.002	< 0.001	0.303
2 20	Unburned - High Severity	46	0.012	0.001	< 0.001
2-20	Low Severity - High Severity	46	< 0.001	0.864	< 0.001
CIII	Unburned - Low Severity	46	< 0.001	< 0.001	< 0.001
20.40	Unburned - High Severity	46	0.090	0.003	< 0.001
20-40	Low Severity - High Severity	46	0.522	0.001	< 0.001
CIII	Unburned - Low Severity	46	0.228	0.783	0.425

537 *Bold value indicates significant difference of p < 0.05, italicized value indicates

538 significant difference of p < 0.1

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565 Table 6. Summary of ANOVA planned contrast tests for SOC stock across the burn

566 severity gradient by soil profile depth

			Total SOC	POC	MAOC	РуС	
Depth	Comparison	df	р	р	р	р	
	Unburned - High Severity	46		< 0.0	01		
Litter	Low Severity - High Severity	46		< 0.0	01		
	Unburned - Low Severity	46	0.147				
0-20 cm	Unburned - High Severity	46	0.064	0.803	< 0.001	< 0.001	
	Low Severity - High Severity	46	0.004	0.934	< 0.001	< 0.001	
	Unburned - Low Severity	46	< 0.001	0.753	< 0.001	0.810	
20-40 cm	Unburned - High Severity	46	0.099	0.065	0.948	< 0.001	
	Low Severity - High Severity	46	0.010	0.035	0.314	< 0.001	
	Unburned - Low Severity	46	0.666	0.854	0.465	0.964	

568 *Bold value indicates significant difference of p < 0.05, italicized value indicates

569 significant difference of p < 0.1

Table 7. Summary of ANOVA planned contrast tests for bulk density across the burn severity gradient by soil profile depth

				Bulk Density
	Depth	Comparison	df	р
		Unburned - High Severity	6	0.014
	0-20 cm	Low Severity - High Severity	6	0.5688
		Unburned - Low Severity	6	0.014
		Unburned - High Severity	6	0.0472
	20-40 cm	Low Severity - High Severity	6	0.7646
		Unburned - Low Severity	6	0.2626
582 583 584	*Bold value indica significant differer	ates significant difference of $p < 0.05$, ince of $p < 0.1$	talicized valu	ie indicates
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603 61. Map showing the soil burn severities of the Holiday Farm Fire based on Sentinel 2 satellite imagery
604 data, where pre- and post-fire images were compared to create a differenced Normalized Burn Ratio
605 (dNMR) dataset (USDA Forest Service et al., 2020). The majority of the wildfire was of moderate soil burn
606 severity, in which moderate but not significant effects of the fire were detected in the soil.
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High severity

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S2. Percent silt and clay compared to percent MAOC for a subset of samples across a range of depths and burn severities.







S3. Distribution of the frequency of percent SOC data as stacked histograms. Values are logged to show

633 634 normal distributions of each SOC fraction. N = 27 for unburned, n = 60 or low severity, and n = 60 for high severity. Corresponds to the lined histogram shown in Figure 2.





S4. Proportion of total SOC that is composed of each fraction (MAOC and POC), organized by SOC pool. The proportions of the two fractions were divided by the measured total SOC. Note that the proportions of the three fractions added together for each site and depth may exceed 100 percent due to the overlap between PyC and its foundational components of mainly POC and some MAOC elements. N = 9 for each depth of unburned (27 total), n = 20 for each depth of low severity (60 total), and n = 20 for each depth of high severity (60 total). See Table 1 in SI for corresponding values. Corresponds to Figure 4, which is organized by site.

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