Elemental composition and optical properties reveal changes in dissolved organic matter along a permafrost thaw chronosequence in a subarctic peatland

Suzanne B. Hodgkins¹,*, Malak M. Tfaily¹, David C. Podgorski³, Carmody K. McCalley⁴, Scott R. Saleska⁵, Patrick M. Crill⁶, Virginia I. Rich⁷, Jeffrey P. Chanton¹, William T. Cooper⁸

¹ Department of Earth, Ocean, and Atmospheric Science, Florida State University, Tallahassee, FL 32306, United States
² Environmental Molecular Sciences Laboratory and Biological Sciences Division, Pacific Northwest National Laboratory, Richland, WA 99354, United States
³ Future Fuels Institute, National High Magnetic Field Laboratory, Florida State University, Tallahassee, FL 32310, United States
⁴ Thomas H. Gosnell School of Life Sciences, Rochester Institute of Technology, Rochester, NY 14623, United States
⁵ Department of Ecology and Evolutionary Biology, University of Arizona, Tucson, AZ 85716, United States
⁶ Department of Geological Sciences, Stockholm University, 10691 Stockholm, Sweden
⁷ Department of Microbiology, The Ohio State University, Columbus, OH 43210, United States
⁸ Department of Chemistry and Biochemistry, Florida State University, Tallahassee, FL 32306, United States

Received 1 December 2015; accepted in revised form 7 May 2016; Available online 14 May 2016

Abstract

The fate of carbon stored in permafrost-zone peatlands represents a significant uncertainty in global climate modeling. Given that the breakdown of dissolved organic matter (DOM) is often a major pathway for decomposition in peatlands, knowledge of DOM reactivity under different permafrost regimes is critical for determining future climate feedbacks. To explore the effects of permafrost thaw and resultant plant succession on DOM reactivity, we used a combination of Fourier transform ion cyclotron resonance mass spectrometry (FT-ICR MS), UV/Vis absorbance, and excitation-emission matrix spectroscopy (EEMS) to examine the DOM elemental composition and optical properties of 27 pore water samples gathered from various sites along a permafrost thaw sequence in Stordalen Mire, a thawing subarctic peatland in northern Sweden. The presence of dense Sphagnum moss, a feature that is dominant in the intermediate thaw stages, appeared to be the main driver of variation in DOM elemental composition and optical properties at Stordalen. Specifically, DOM from sites with Sphagnum had greater aromaticity, higher average molecular weights, and greater O/C, consistent with a higher abundance of phenolic compounds that likely inhibit decomposition. In contrast, sites without Sphagnum, specifically fully-thawed rich fens, had more saturated, more reduced compounds, which were high in N and S. Optical properties at rich fens indicated the presence of microbially-derived DOM, consistent with the higher decomposition rates previously measured at these sites. These results indicate that Sphagnum acts as an inhibitor of rapid decomposition and CH₄ release in thawing subarctic peatlands, consistent with lower rates of CO₂ and CH₄ production previously observed at these sites. However, this inhibitory effect may disappear if Sphagnum-dominated bogs transition to more waterlogged rich fens that contain very little to no living

* Corresponding author. Tel.: +1 850 320 5006; fax: +1 850 644 2581.
E-mail address: sbh10c@fsu.edu (S.B. Hodgkins).

http://dx.doi.org/10.1016/j.gca.2016.05.015
2016 Elsevier Ltd. All rights reserved.
Sphagnum. Release of this inhibition allows for higher levels of microbial activity and potentially greater CH$_4$ release, as has been observed in these fen sites.

© 2016 Elsevier Ltd. All rights reserved.

Keywords: Dissolved organic matter; FT-ICR MS; UV/Vis; EEMS; Peatlands; Permafrost

1. INTRODUCTION

1.1. Decomposition in northern peatlands

Boreal and subarctic peatlands store between 270 and 455 Pg of carbon (Gorham, 1991; Turunen et al., 2002), of which ~277 Pg is in the permafrost zone (Schuur et al., 2008). This carbon has an uncertain fate as permafrost thaws due to continuing warming (Schuur et al., 2008). Although the Arctic is currently a net sink for carbon dioxide (CO$_2$), sequestering up to 0.8 Pg C/year mostly due to boreal forest growth, it is a net source of methane (CH$_4$) with annual emissions of 32–112 Tg CH$_4$/year, mostly from wetlands and lakes (Walter et al., 2007; McGuire et al., 2009; Thornton et al., 2015). Since CH$_4$ results in 25–33 times the radiative forcing of CO$_2$ (per kg of gas) on a 100-year timescale (Forster et al., 2007; Shindell et al., 2009), the response of CH$_4$ production to warming and permafrost thaw represents a major potential climate feedback. Net CH$_4$ and CO$_2$ balances in thawing permafrost peatlands depend on plant production, which determines net autotrophic CO$_2$ uptake, and belowground decomposition, which determines heterotrophic CH$_4$ and CO$_2$ release. Belowground decomposition rates and pathways depend on inputs and reactivity of fresh plant litter and thawed permafrost carbon, soil water saturation (which controls belowground oxygen availability and aboveground plant species), and microbial metabolic pathways. The responses of these factors, and thus CH$_4$ and CO$_2$ emissions, to a warming climate remain uncertain. In this study, we focus on the effects of permafrost thaw and associated vegetation change on peatland organic matter chemistry.

1.2. Controls on DOM composition and lability

Anaerobic decomposition in peatlands is typically fueled by dissolved organic matter (DOM) rather than solid peat (Chanton et al., 2008). According to the size-reactivity model (Alperin et al., 1994; Amon and Benner, 1996; Burdige and Gardner, 1998), the first step in DOM decomposition is the degradation of high molecular weight macromolecules (HMW-DOM), defined as either >1000 or >3000 Da, into either monomeric or polymeric low molecular weight DOM (mLMW-DOM or pLMW-DOM). mLMW-DOM is more labile and can be rapidly degraded into CO$_2$ or used in methanogenesis (Conrad, 1999), while pLMW-DOM is more refractory and often accumulates. The reactivity of peatland DOM is thus an important control on both overall decomposition rates and the proportion of carbon released as CH$_4$.

Inputs from aboveground plants and belowground thawing permafrost (Ward and Cory, 2015) determine the initial composition and lability (defined here as the potential for rapid decomposition) of raw DOM that enters peat pore water. The main compounds that plants introduce to the DOM pool—either via direct input or via release from solid-phase peat—including carbohydrates, proteins, lipids, lignin-derived compounds, tannins and other phenolics, and other miscellaneous compounds such as chlorophyll (Kögel-Knabner, 2002; Kalbitz et al., 2003b). Microbial activity can then change DOM composition by several processes including preferential mineralization of more labile DOM compounds, alteration of existing compounds, or assimilation into microbial biomass. Compounds that are usually enriched as others are degraded include lignin, humic acids, other aromatics, lipids, and alkyls with double bonds (Harvey et al., 1995; Glatzel et al., 2003; Kalbitz et al., 2003a, 2003b; Lorenz et al., 2007). Lignin in particular can only be degraded aerobically (Kirk and Farrell, 1987), and thus should accumulate in anaerobic peatlands.

DOM structure can influence decomposition not only directly via its reactivity, but also indirectly via reactions of specific DOM molecules. For example, humic substances and quinones can act as electron acceptors in anaerobic respiration, which due to more favorable energy yields outcompetes methanogenesis (Lovley et al., 1996; Cervantes et al., 2000; Heitmann et al., 2007; Keller and Bridgham, 2007; Blodau and Deppe, 2012; Bridgham et al., 2013). Phenolic compounds can also inhibit hydrolase enzymes and thereby suppress decomposition (Freeman et al., 2001, 2004).

Compared to other plant groups, litter from Sphagnum mosses decomposes unusually slowly. This slow decomposition is caused both directly by Sphagnum litter’s recalcitrance, and indirectly by the inhibitory activity of specific compounds that also slow the decomposition of organic matter from other sources (Verhoeven and Toth, 1995). The main component of Sphagnum that contributes to its tanning properties (i.e., preservation of animal tissues) is the polysaccharide sphagnan (Painter, 1991), which is unusual for its reactive carboxylic acid groups contained within uronic acid monomers. These acids can form complexes with proteins, including those found in Sphagnum litter, to form recalcitrant humic substances via abiotic Maillard reactions (Painter, 1983, 1991). Enzyme activity necessary for biodegradation is also suppressed, both by complexation of enzymes with sphagnan (Painter, 1991), and by enzyme-inhibiting phenolics such as tannin-like compounds and sphagnum acid (van Breen, 1995; Verhoeven and Toth, 1995; Verhoeven and Liefveld, 1997; Freeman et al., 2004). Experiments and field observations indicate that humic substances from Sphagnum inhibit methanogenesis, not only by competition as terminal electron acceptors, but also by possible direct toxicity to methanogens.
other factors, which may contribute to Sphagnum peat’s direct recalcitrance, include acidic pH and low nutrient availability, the latter caused in part by nitrogen entrapment within microbial biomass (Damman, 1988) and within sphagnum-derived humic acids (Painter, 1991).

In addition to external inputs, microbial activity also contributes compounds to DOM, resulting in a feedback mechanism in which DOM molecular structure both affects, and is affected by, microbial activity (Glaz et al., 2003; Hodgkins et al., 2014). Compounds that can be directly produced by microbes, either as biomass or via extracellular reactions, include carbohydrates, proteins, aminosugars, and lipids (Ogawa et al., 2001; Kögel-Knabner, 2002; Kalbitz et al., 2003b; Kindler et al., 2009; Miltner et al., 2009), as well as uncharacterized molecules of unknown structure (Ogawa et al., 2001; Marschner and Kalbitz, 2003). Although microbially-derived material is commonly more labile than plant-derived material, a portion of it is refractory (Ogawa et al., 2001; Gruber et al., 2006) and becomes increasingly resistant to further degradation as it is recycled through the microbial food web (Kindler et al., 2006, 2009; Miltner et al., 2009). For example, although carbohydrates and peptides are usually considered more labile (Marschner and Kalbitz, 2003), they can sometimes accumulate either in DOM or in the solid phase due to microbial synthesis of large volumes of these compounds (Kögel-Knabner, 2002; Kalbitz et al., 2003b; Miltner et al., 2009), for instance in exopolymERIC substances used for protection and adhesion (Wingender et al., 1999).

### 1.3. Analytical approach

Fourier transform ion cyclotron resonance mass spectrometry (FT-ICR MS) is an extremely powerful, ultrahigh-resolution method for obtaining detailed information on DOM elemental composition, and can thus reveal important differences in chemical composition between DOM from different sources and give insights on decomposition processes. Due to the very specialized instrumentation required and lengthy analysis process, this method is typically appropriate for targeted characterization of a limited number of samples. To complement FT-ICR MS, the analysis of DOM optical properties, including both absorption and fluorescence, allows for higher sample throughput and identifies many of the same overall characteristics—including molecular weight (MW), aromaticity, and microbial vs. terrestrial origins (Tfaily et al., 2015).

Identification of the exact chemical structures responsible for different DOM optical properties, and thus robustly interrelating optical and FT-ICR MS data, presents significant challenges. Parallel factor analysis (PARAFAC) can aid in this task by deconvoluting DOM fluorescence spectra into the individual fluorophores that contribute to a complete spectrum (Stedmon and Bro, 2008). However, the properties associated with each fluorophore identified in DOM have not yet been unequivocally assigned (Ishii and Boyer, 2012). Recently, identification of the DOM molecules associated with different PARAFAC components has been accomplished by statistical correlation with compounds found by FT-ICR MS (Stubbins et al., 2014; Kellerman et al., 2015). However, even these types of analyses do not necessarily indicate that the molecular formulas associated with each component are directly contributing to its fluorescence. Factors that may lead to indirect correlations between molecular formulas and PARAFAC components include solution matrix effects on fluorescence properties, bio- and photodegradation reactions of fluorescing compounds into non- or differently-fluorescing compounds, and common sources and sinks of multiple compounds that fluoresce differently (Sharpless and Blough, 2014; Stubbins et al., 2014). Nonetheless, some of these factors may prove advantageous as fluorescing DOM molecules can provide valuable information on non-fluorescing molecules with which they commonly co-occur (Stubbins et al., 2014). More studies are clearly needed to strengthen the methods for predicting bulk DOM elemental composition based on the optical properties of chromophoric DOM.

To help address this knowledge gap, while also improving our understanding of warming-related changes in peatland DOM, we performed a full comparison of FT-ICR MS and optical properties for 27 DOM samples gathered across a permafrost thaw succession with different peatland vegetation types in Stordalen Mire, a thawing peatland complex in subarctic Sweden. Specifically, we obtained UV/Vis absorbance spectra, and measured fluorescence by excitation-emission matrix spectroscopy (EEMS), to aid in the interpretation of pore water DOM elemental composition data obtained by FT-ICR MS. Methods used to analyze fluorescence spectra included common indices for inferring aromaticity and MW, as well as relative fluorophore intensities based on a PARAFAC model developed by Tfaily et al. (2015). This represents the largest intercomparison of FT-ICR MS and optical data from wetland samples yet published, and focuses on DOM from anaerobic peatland pore waters from across a permafrost thaw chronosequence. This study will not only significantly improve our ability to interrelate DOM optical properties with DOM elemental composition, it will also provide a higher-resolution understanding of previously observed changes in organic matter lability resulting from permafrost thaw.

### 2. METHODS

#### 2.1. Study site

Stordalen Mire (68.35°N, 19.05°E) is a peat plateau in subarctic Sweden underlain by discontinuous permafrost, which is thawing as the region warms due to global change. Sites with intact permafrost (palsas) are dry and ombrotrophic with an aerobic active layer, and are vegetated by a combination of lichens, Eriophorum vaginatum, and ericaceous and woody plants. In areas where the permafrost is degraded, land subsidence leads to varying degrees of water inundation. In this study, we focused on these inundated sites because the intact palsas are dry and therefore contain little to no DOM. The inundated sites are categorized based on the classification system used by Hodgkins et al. (2014),...
McCall et al. (2014), and Mondav et al. (2014), whose site classifications are based on those of Johansson et al. (2006). In approximate order of increasing time since inundation, these habitat types are as follows:

- Recently thawed and collapsed thermokarst sinkholes (“col. palsa”) surrounded by palsa; these holes contain no permafrost, can have a water table either above or below the peat surface, and may be vegetated by a combination of species depending on water level. In the collapsed palsa site measured in this study (site PHS; Table 1), the dominant plants are *E. vaginatum* and *Sphagnum* spp.

- Thawing *Sphagnum*-dominated bogs with a water table below the peat surface, which is often perched above a deeper permafrost layer.

- Poor fens containing a combination of *Sphagnum* spp. and tall sedges (e.g., *Eriophorum angustifolium*), which typically either have no permafrost or nearly thawed permafrost.

- Rich fens dominated by tall sedges, such as *E. angustifolium* and/or *Carex rostrata*, and which have no permafrost.

In terms of greenhouse gas balance, intact palsas are a net CO₂ source to the atmosphere and approximately CH₄ neutral due to their low plant productivity combined with relatively fast decomposition in their aerobic active layers (Christensen et al., 2004; Bäckstrand et al., 2010; McCall et al., 2014). Little is known about the net carbon and greenhouse gas balances of collapsed palsa sinkholes, but incubation experiments suggest that decomposition within collapsed palsas releases very little CH₄ and CO₂ compared to the other sites (Hodgkins et al., 2014). Bogs and fens are net CO₂ sinks due to their higher primary productivity, but their anaerobic subsurface conditions make them sources of atmospheric CH₄ (Bäckstrand et al., 2010; McCall et al., 2014). In the case of fens, this CH₄ production is so high that their net radiative forcing impact per unit area over a 100-year timescale is approximately 7 times that of intact palsas and 28 times that of bogs (Bäckstrand et al., 2010). Due to fens’ high CH₄ emissions and the expansion of fens and other wet sites with permafrost thaw (Malmer et al., 2005; Johansson et al., 2006; Bäckstrand et al., 2010), the net annual greenhouse gas balance of the entire mire (in terms of radiative forcing)

<table>
<thead>
<tr>
<th>Sample name</th>
<th>Habitat type</th>
<th>Site</th>
<th>Depth (cm)</th>
<th>Sampling date</th>
<th>pH</th>
<th>DOC (mM)</th>
</tr>
</thead>
<tbody>
<tr>
<td>PHS.1.10</td>
<td>Col. palsa</td>
<td>PHS</td>
<td>10</td>
<td>13 Jun 2011</td>
<td>4.0</td>
<td>4.60</td>
</tr>
<tr>
<td>PHS.1.30</td>
<td>Col. palsa</td>
<td>PHS</td>
<td>30</td>
<td>13 Jun 2011</td>
<td>4.1</td>
<td>5.35</td>
</tr>
<tr>
<td>PHS.1.50</td>
<td>Col. palsa</td>
<td>PHS</td>
<td>50</td>
<td>13 Jun 2011</td>
<td>4.3</td>
<td>8.62</td>
</tr>
<tr>
<td>Bog1.0.40</td>
<td>Bog</td>
<td>Bog1</td>
<td>40</td>
<td>31 Aug 2010</td>
<td>3.6</td>
<td>10.58</td>
</tr>
<tr>
<td>SOS.0.38</td>
<td>Bog</td>
<td>SOS</td>
<td>37.5</td>
<td>31 Aug 2010</td>
<td>3.7</td>
<td>4.00</td>
</tr>
<tr>
<td>SOS.1.10</td>
<td>Bog</td>
<td>SOS</td>
<td>10</td>
<td>13 Jun 2011</td>
<td>4.0</td>
<td>6.10</td>
</tr>
<tr>
<td>SOS.1.31</td>
<td>Bog</td>
<td>SOS</td>
<td>31</td>
<td>13 Jun 2011</td>
<td>4.0</td>
<td>5.13</td>
</tr>
<tr>
<td>S.0.38a</td>
<td>Bog</td>
<td>S</td>
<td>37.5</td>
<td>1 Sep 2010</td>
<td>3.9</td>
<td>3.77</td>
</tr>
<tr>
<td>S.0.38b</td>
<td>Bog</td>
<td>S</td>
<td>37.5</td>
<td>1 Sep 2010</td>
<td>3.9</td>
<td>3.72</td>
</tr>
<tr>
<td>S.2.12</td>
<td>Bog</td>
<td>S</td>
<td>12</td>
<td>21 Aug 2012</td>
<td>4.2</td>
<td>5.08</td>
</tr>
<tr>
<td>S.2.70</td>
<td>Bog</td>
<td>S</td>
<td>70</td>
<td>21 Aug 2012</td>
<td>5.0</td>
<td>6.42</td>
</tr>
<tr>
<td>EOS.0.38</td>
<td>Poor fen</td>
<td>EOS</td>
<td>37.5</td>
<td>31 Aug 2010</td>
<td>3.7</td>
<td>6.35</td>
</tr>
<tr>
<td>EOS.1.10</td>
<td>Poor fen</td>
<td>EOS</td>
<td>10</td>
<td>13 Jun 2011</td>
<td>4.9</td>
<td>2.55</td>
</tr>
<tr>
<td>EOS.1.25</td>
<td>Poor fen</td>
<td>EOS</td>
<td>25</td>
<td>13 Jun 2011</td>
<td>4.6</td>
<td>3.46</td>
</tr>
<tr>
<td>EOS.1.55</td>
<td>Poor fen</td>
<td>EOS</td>
<td>55</td>
<td>13 Jun 2011</td>
<td>4.9</td>
<td>3.17</td>
</tr>
<tr>
<td>E.0.38a</td>
<td>Rich fen</td>
<td>E</td>
<td>37.5</td>
<td>1 Sep 2010</td>
<td>5.4</td>
<td>0.72</td>
</tr>
<tr>
<td>E.0.38b</td>
<td>Rich fen</td>
<td>E</td>
<td>37.5</td>
<td>1 Sep 2010</td>
<td>5.5</td>
<td>0.90</td>
</tr>
<tr>
<td>E.1.3</td>
<td>Rich fen</td>
<td>E</td>
<td>2.5</td>
<td>15 Jun 2011</td>
<td>6.0</td>
<td>0.69</td>
</tr>
<tr>
<td>E.1.7</td>
<td>Rich fen</td>
<td>E</td>
<td>6.5</td>
<td>15 Jun 2011</td>
<td>6.0</td>
<td>0.74</td>
</tr>
<tr>
<td>E.1.26</td>
<td>Rich fen</td>
<td>E</td>
<td>25.5</td>
<td>15 Jun 2011</td>
<td>6.0</td>
<td>0.78</td>
</tr>
<tr>
<td>E.1.50</td>
<td>Rich fen</td>
<td>E</td>
<td>50</td>
<td>15 Jun 2011</td>
<td>5.7</td>
<td>0.80</td>
</tr>
<tr>
<td>E.2.7</td>
<td>Rich fen</td>
<td>E</td>
<td>7</td>
<td>21 Aug 2012</td>
<td>5.9</td>
<td>0.61</td>
</tr>
<tr>
<td>E.2.70</td>
<td>Rich fen</td>
<td>E</td>
<td>70</td>
<td>21 Aug 2012</td>
<td>6.2</td>
<td>0.61</td>
</tr>
<tr>
<td>Fen2.1.15</td>
<td>Rich fen</td>
<td>Fen2</td>
<td>15</td>
<td>14 Jun 2011</td>
<td>5.8</td>
<td>0.53</td>
</tr>
<tr>
<td>Fen2.1.30</td>
<td>Rich fen</td>
<td>Fen2</td>
<td>30</td>
<td>14 Jun 2011</td>
<td>5.5</td>
<td>1.19</td>
</tr>
<tr>
<td>Fen2.1.60</td>
<td>Rich fen</td>
<td>Fen2</td>
<td>60</td>
<td>14 Jun 2011</td>
<td>5.4</td>
<td>2.42</td>
</tr>
<tr>
<td>Fen2.1.85</td>
<td>Rich fen</td>
<td>Fen2</td>
<td>85</td>
<td>14 Jun 2011</td>
<td>5.7</td>
<td>1.09</td>
</tr>
</tbody>
</table>
ions were generated with a home-built electrospray ionizer (EHMFL) (Tallahassee, FL). Negatively-charged molecules were characterized using a Fourier transform ion cyclotron resonance mass spectrometer (FT-ICR MS) with a 9.4 Tesla superconducting magnet (Dittmar et al., 2008), the final DOC concentrations in each sample (Table 1). These samples were used to characterize the changes in DOM elemental composition and optical properties along the gradient of permafrost thaw.

Pore water was gathered by syringe suction through a 1-m long, 0.5-cm-diameter stainless steel tube with holes drilled along the bottom 3 cm, then filtered through 0.7-μm Whatman GF/F glass microfiber filters into 120-mL brown borosilicate bottles. The pH was measured on-site with an Oakton Waterproof pHTestr 10. The bottles were frozen within 8 h of collection and stored frozen until analysis.

**2.3. DOC concentrations**

Dissolved organic carbon (DOC) concentrations were measured by high-temperature catalytic oxidation on a Shimadzu Total Organic Carbon analyzer with a non-dispersive infrared detector. Each sample was analyzed with triplicate measurements, which always had a coefficient of variance of <2%.

**2.4. FT-ICR MS**

Although the samples were all freshwater, even small amounts of salts can affect FT-ICR MS ionization efficiencies. The samples were therefore desalted by solid phase extraction (SPE) with PPL cartridges (Dittmar et al., 2008). Thirty mL of each sample (acidified to pH = 2 with HCl) was extracted, and the DOM was eluted in methanol (see Tfailey et al. (2012) for detailed methods). Based on an extraction efficiency of 62% for freshwater samples (Dittmar et al., 2008), the final DOC concentrations in the extracts ranged from 0.2 to 0.3 mg/mL for collapsed palsas; 0.1 to 0.4 mg/mL for bogs; 0.1 to 0.2 mg/mL for poor fens; and 0.02 to 0.1 mg/mL for rich fens.

The extracted samples were analyzed on a custom-built Fourier transform ion cyclotron resonance mass spectrometer (FT-ICR MS) with a 9.4 Tesla superconducting magnet at the National High Magnetic Field Laboratory (NHMFL) (Tallahassee, FL). Negatively-charged molecular ions were generated with a home-built electrospray ionization (ESI) source. The samples were injected into the ESI source at 0.5 μL/min with a syringe pump connected to a 50-μm-internal-diameter fused silica tube. The experimental parameters were chosen based on previous fulvic acid characterizations for optimization of DOM MS methods (Stenson et al., 2003), and were as follows: needle voltage, −2.5 kV; tube lens, −340 V; and heated metal capillary, 8 W. The ion accumulation time per scan was adjusted to account for the differences in SPE extract concentrations, with longer accumulation times used for more dilute samples. This results in approximately the same number of ions reaching the ICR analyzer cell, regardless of sample concentration. Each spectrum was produced as a sum of 200 individual scans, and the average resolving power (m/Δm(m/z)) at 451 Da was >700,000. Spectra were not blank-corrected because it is not possible to know the ionization efficiency of compounds that produce background signals in the presence of the overwhelming concentration of extracted DOM compounds. Our interpretations of mass spectra rely on observed differences in samples, and this “differencing” removes any contributions from background peaks.

**2.5. UV/Vis absorption spectroscopy**

The UV/Vis absorbance of each pore water sample was measured between 250 and 400 nm on a Cary Varian 100 dual beam UV/Vis spectrometer using a 10 mm Suprasil cuvette, with blank-correction using Milli-Q water in an identical cuvette measured alongside the sample. To prevent inner filter effects during absorbance measurements, samples with absorbances >0.8 at a wavelength of 350 nm or >2.2 at 254 nm (absorbance units; AU) were diluted with Milli-Q water until they had absorbances below these values. To prevent inner-filter effects during subsequent fluorescence measurements, samples were further diluted with Milli-Q water until they had an absorbance of ≤0.02 AU at 350 nm (Kowalczuk et al., 2003); after adjusting for dilution, the absorbance values of these highly diluted samples (≤0.02 AU at 350 nm) were within 5% of the absorbance values of the same samples measured at higher concentration (≤0.8 AU at 350 nm, used in calculations with UV/Vis data). To make these measurements reflective of the in situ absorption and fluorescence of the samples, no pH adjustments were carried out. While we recognize that metals can influence absorption and fluorescence through quenching, metal concentrations in peatland pore water are in the ng/L range (Rausch et al., 2006), several orders of magnitude lower than the DOC concentrations in our samples (Table 1). Each sample was measured in duplicate.

**2.6. Fluorescence (EEMS)**

Fluorescence spectra (after samples were diluted to absorbance ≤0.02 AU at 350 nm) and a blank (Milli-Q water used for dilution) were measured in a 10-mm path length quartz cuvette using a Jobin Yvon SPEX Fluoromax-4 spectrometer with a Xenon lamp source. Excitation wavelengths were scanned from 240 to 500 nm in 5 nm increments, and emission was measured from 290 to 600 nm in 2 nm increments. Data were gathered in signal/reference mode, normalizing the fluorescence emission signal with the excitation intensity. The band pass for both excitation and emission monochromators was 5 nm, and the integration time was 0.1 s.
2.7. Calculations

2.7.1. Elemental composition

MIDAS Predator Analysis and Molecular Formula Calculator software from the NHMFL were used for internal calibration and molecular formula assignment, respectively. FT-ICR MS spectra were calibrated with two internal homologous series of formulas separated by 14 Da (representing -CH2 groups), and the mass accuracy was calculated as <1 ppm for singly charged ions across the mass distribution \((m/z = 170–800)\). Molecular formulas were calculated for signals >6\(\sigma\) RMS baseline noise (Tfaily et al., 2011, 2013) based on the presence of C, H, O, N, and S (Stenson et al., 2003), and only ion masses with a mass accuracy of ±1 ppm compared to the IUPAC exact mass were included in the data set. The abundance threshold of >6\(\sigma\) above the RMS baseline noise was chosen because it results in a less than 1% probability of picking peaks attributed to random noise (this threshold is twice the 95% confidence detection limit of 3\(\sigma\) RMS noise that is generally accepted for analytical measurements), while maximizing the number of formulas that can be assigned (small peaks below this threshold contribute <1% to the overall sum of peak intensities). Spacing of 1.0034 Da between 13C isotopologues of the same molecule \((12C_n, 13C_{n-1})\) confirmed that all ions with assigned formulas were singly charged (Brown and Rice, 2000; Kujawinski et al., 2002). In each sample, the relative abundances of all assigned formulas were normalized to have a sum of 100, and the formulas containing 13C were then removed. Reproducibility of assigned formulas in replicate samples was good (Sleighter et al., 2012): a pair of replicates from a bog (S.0.38a and S.0.38b) shared 86% of formulas in common while a pair of replicates from a rich fen (E.0.38a and E.0.38b) shared 87% of formulas, and there were no systematic differences in the elemental compositions of the remaining non-matching formulas in either pair of replicates. In contrast, pairs of unique (non-replicate) samples shared, on average, 64% of formulas in common (range 33–87%), and there were often systematic differences in elemental composition (see Section 3.2).

For initial analysis of differences between bogs and rich fens, data from six bog samples and six rich fen samples from similar sampling dates and depths (specifically, bog samples S.0.38a, S.0.38b, SOS.1.10, SOS.1.31, S.2.12, and S.2.70, and fen samples E.0.38a, E.0.38b, Fen.2.1.15, Fen.2.1.30, E.2.7, and E.2.70) were selected. The normalized relative abundances in each sample (normalized to a sum of 100, as described above) were then used to calculate average relative abundances for each compound across all 6 samples in each habitat type (bog and rich fen), using an abundance of 0 for formulas not observed in a given sample. The average abundance of each formula in rich fens was then subtracted from its average abundance in bogs, giving a measure of the formula’s “more bog-like” (positive differences) or “more fen-like” (negative differences) prevalence. The formulas were then visualized as 3-D van Krevelen (Fig. 1a) and molecular size (Fig. 1b) plots, with formulas more abundant in bogs or fens indicated by color.

To compare changes in elemental composition across all 27 samples, abundance-weighted average MW, O/C, H/C, N/C, and S/C were computed for each sample. For example, the abundance-weighted average O/C of each sample was calculated as:

\[
{\text{O/C}}_{\text{total}} = \frac{\sum_{i=1}^{p} (a_i \times nO_i)}{\sum_{i=1}^{p} (a_i \times nC_i)}
\]

where \(a_i\) is the relative abundance of formula \(i\), \(nO_i\) is the number of oxygen atoms in formula \(i\), and \(nC_i\) is the number of carbon atoms in formula \(i\). The other average elemental ratios were calculated in a similar manner. The abundance-weighted average MW in each sample was calculated as:

\[
{\text{MW}}_{\text{avg}} = \frac{\sum_{i=1}^{p} (a_i \times {\text{MW}}_i)}{\sum_{i=1}^{p} a_i}
\]

The formulas were also categorized into classes according to their H/C and O/C (Kim et al., 2003; Sleighter and Hatcher, 2007; Hodgkins et al., 2014). These classes included lipid-like, protein-like, aminosugar and carbohydrate-like (AS.carb), unsaturated hydrocarbon (UH), condensed aromatic (CA), lignin-like, and tannin-like compounds (Table 2, Fig. 1a). It is important to note that due to the different ionization efficiencies of different compounds, the relative abundances of different compound classes do not directly correspond to actual concentrations, but should rather be interpreted as relative amounts for different samples. For example, although the apparent relative abundance of the AS.carb class in all samples is nearly zero due to the very low ionization efficiencies of carbohydrates, the actual fraction of aminosugars and carbohydrates in DOM is likely much higher.

2.7.2. Bulk optical properties

The UV/Vis absorbance spectra were averaged across duplicates, then multiplied by the dilution factor (undiluted/diluted concentration ratio) and divided by the path length of the cuvette to obtain absorbance per meter (m\(^{-1}\)) in the undiluted samples. Each EEMS fluorescence spectrum was corrected for Raman scattering, Rayleigh scattering, and dilution using FL Toolbox 1.91 software in MATLAB according to the methods of Gonsior et al. (2008). Raman and Rayleigh scattering peaks were eliminated by removing portions (±10–15 nm FW) of the fluorescence signal centered on the respective scatter peaks (Zepp et al., 2004). As suggested by Lawaetz and Stedmon (2009), the data were then normalized to the Raman intensity of the blank (350ex/397em, 5 nm band pass; Tfaily et al., 2013) and converted to Raman-normalized quinine sulfate equivalents (QSE) in ppb through an external calibration (Coble et al., 1998). To correct the spectra for dilution, the scatter-corrected fluorescence of the blank was subtracted, and the resulting values were then multiplied by the dilution factor to obtain the fluorescence of the undiluted sample.

Based on the DOC concentrations, UV/Vis absorbance spectra, and EEMS spectra for each sample, we determined...
the following bulk optical properties that relate to molecular structure:

- Speciﬁc UV absorbance (SUVA), which is deﬁned as the ratio of the absorbance at 254 nm (mL/mg/L) to the DOC concentration (mg/L), and increases with aromaticity (Weishaar et al., 2003).
- Biological index (BIX), deﬁned as the (380 nm)/(430 nm) emission intensity ratio at an excitation wavelength of 310 nm. In a study of estuarine samples, Huguet et al. (2009) found that BIX increases with the contribution of microbial exudates to the DOM pool, with values <0.6 indicating little microbial activity and values >1 indicating DOM of primarily microbial origin.
- Humiﬁcation index (HIX), deﬁned as H/L, where H and L are the integrated emission intensities from 434 to 480 nm and 300 to 344 nm (respectively) measured at an excitation wavelength of 255 nm (Zsolnay et al., 1999). Since this index is positively correlated with aromaticity, it often increases with decomposition as organic matter becomes more humiﬁed (Kalbitz et al., 2003b; Hur, 2011). However, the accumulation of microbial exudates during decomposition can also lower HIX (Kalbitz et al., 2003b; Birdwell and Engel, 2010; Tfaily et al., 2015).
- Emission wavelength of maximum ﬂuorescence of peak C as deﬁned by Coble (1996) (Emax), which increases with unsaturation, with longer wavelengths indicating more conjugated molecules and condensed aromatic structures (Senesi, 1990).
- Ratio of peak C intensity (QSE) (Fmax) to the absorption coeﬃcient at 340 nm (A340, deﬁned as the absorbance in mL/mg/L times 2.303). This ratio, Fmax/A340 decreases with increasing MW (Stewart and Wetzel, 1981; Baker et al., 2008).

2.7.3. PARAFAC model

Due to the relatively small number of samples used in this study, and the consequently poor statistical reliability of a PARAFAC model built solely from these EEMS data, a recently developed peatland PARAFAC model based on pore water gathered from the Glacial Lake Agassiz peatlands (GLAP) in northern Minnesota (Tfaily et al., 2015) was used to analyze the EEMS spectra. This peatland had similar vegetation characteristics and DOM concentrations as those present at Stordalen. This model includes ﬁve components (Table A.1), with four (C1–C4) categorized as humic-like and one (C5) as protein-like. C1 likely represents plant-derived humic-like substances that can be biodegraded.
C2 and C3 have shorter emission wavelengths than the other humic-like components, and thus appear to be microbially-derived, with C3 a possible microbial metabolite of C1. C4 appears refractory and likely represents either ubiquitous terrestrial high-molecular-weight humic-like aromatics, or terrestrial reduced quinones identified by Cory and McKnight (2005). The protein-like component, C5, corresponds with peak T as defined by Coble (1996) and is a marker for fresh (undecomposed) tryptophan-like structures.

The relative intensity of each component in the Stordalen samples was calculated by fitting the EEMS spectra to this PARAFAC model, with the loadings of each component interpreted as relative intensities in each sample. Detailed comparisons of PARAFAC component intensities between Stordalen and the GLAP, as well as additional comparisons of elemental composition and optical properties, are given in the Supporting information.

2.7.4. Statistical analysis of trends between samples

For each FT-ICR MS-derived value and optical property, as well as pH and DOC concentrations, one-way analyses of variance (ANOVA) were used to compare samples from different sites and habitat types. Since sampling only captured a few depths at each site (on each sampling date), depth trends were assessed with paired t-tests comparing the shallowest vs. the deepest samples in each depth profile that had at least two distinct depths. The results of these statistical tests are shown in Table 3.

2.7.5. Comparison of FT-ICR MS and optical measurements

To reveal general correlations between the FT-ICR MS and optical data, as well as the general trends in both datasets that separate different DOM samples, we performed two principal components analyses (PCA) with different combinations of FT-ICR MS and optical variables (Sleighter et al., 2010; Tfaily et al., 2015; Wagner et al., 2015). The first PCA examined general trends in elemental composition and optical properties without reference to compound class divisions or PARAFAC components, and included FT-ICR MS abundance-weighted average molecular characteristics (specifically, MW and element ratios) and non-PARAFAC optical properties (SUVA, BIX, HIX, Em\text{max}, and F\text{max}/A\text{abs}). The second PCA examined interrelations between compound classes and PARAFAC components, and included FT-ICR MS compound class abundances and PARAFAC component relative intensities. Two separate PCAs were performed rather than combining all of the variables into one PCA in order to minimize auto-correlation between related sets of variables that were based on the same measurements (e.g., average O/C and tannin-like compound abundance). Both PCAs were performed with all variables scaled to unit variance to account for different measurement units and ranges of values.

Given that PCA may show misleading results due to dimensionality reduction, individual comparisons of each FT-ICR MS-derived variable (including compound classes as well as average molecular characteristics) with each optical property across all samples represent a more robust method for correlating molecular characteristics with optical properties. To this end, the trends revealed by the PCAs were further explored and validated based on linear regressions, with the bulk optical properties and PARAFAC component intensities as the independent variables, and the FT-ICR MS results—including average molecular characteristics, total relative abundance of each compound class, and total relative abundance of various combinations of up to four compound classes—as the dependent variables. Combinations of compound classes were chosen so that all classes occupied adjacent regions of the van Krevelen diagram and were not surrounded on more than two sides by other compound classes (Fig. 1a).

3. RESULTS

3.1. DOC concentrations and pH

One-way ANOVAs (Table 3) revealed significant differences in both pH and DOC concentrations between habitat types and between specific sites, with no significant differences with depth (based on paired t-tests of shallow vs. deep samples; Table 3) or between sampling dates (data not shown). Both pH and DOC mainly differed between rich fens and other sites (collapsed palsas, bogs, and poor fens, which all had Sphagnum), with higher pH and much lower DOC concentrations in rich fens (Table 1).

3.2. FT-ICR MS

Based on van Krevelen and molecular size plots of all formulas, the overall elemental compositions of bog and rich fen DOM were distinctly different, with lower O/C and higher H/C (Fig. 1a) and lower molecular sizes (Fig. 1b) in rich fen DOM compared to bog DOM.

Across the entire thaw gradient, rich fens had a higher abundance of lipid-like, protein-like, AS.carb, UH, and lignin-like compounds, as well as higher average H/C, N/C, and S/C, while the sites with Sphagnum had more tannin-like compounds, higher O/C, and higher MW (Fig. 2; exact values are given in Table A.2, and compound class abbreviations and ranges are defined in Table 2). One-way ANOVAs (Table 3) revealed that these trends were significant for the protein, AS.carb, lignin, and tannin compound classes. MW and O/C, H/C, and S/C also varied significantly between habitat types, and O/C, H/C, N/C, and S/C varied significantly between sites. Far fewer trends were observed with depth as compared with the trends between sites and habitat types. Specifically, H/C and the AS.carb compound class both decreased with depth in rich fens, while UH decreased with depth across all sites (Fig. 2, Table 3).

3.3. Optical properties

As with DOM elemental composition, most of the variation in optical properties separated different sites and different habitat types (Table 3). Specifically, DOM from sites with Sphagnum had higher SUVA, HIX, Em\text{max}, and C1 and C4 relative intensities, while rich fens had higher BIX, F\text{max}/A\text{abs}, and C2 and C3 relative intensities (Fig. 3; exact values are given in Table A.2, and terms are defined in...
One-way ANOVAs revealed that all of these differences were significant except for the differences in HIX (Table 3), likely because two of the Sphagnum-influenced samples (Bog1.0.40 and EOS.0.38) had unusually low HIX (Fig. 3) due to high tryptophan-like fluorescence. Fewer significant trends were observed with depth (Table 3). Among the depth trends that were observed, C1 increased with depth across all sites, C1 and C5 increased with depth in sites with Sphagnum, and Emmax increased with depth in rich fens (Fig. 3). No significant differences were observed between sampling dates (data not shown).

### 3.4. Correlation of optical properties with FT-ICR MS

Both of the principal components analyses of variables measured in this study (Fig. 4) showed a similar pattern in sample clustering in which rich fens are clearly separated from the other sites along PC1, with higher PC1 scores in rich fens. In the first PCA (Fig. 4a), which examined general trends in elemental composition and optical properties, PC1 correlates positively with H/C, N/C, S/C, Fmax/A340, and BIX, and correlates negatively with O/C, MW, Emmax, SUVA, and HIX. In the second PCA (Fig. 4b), which used more specific DOM compound classes and PARAFAC components, PC1 correlates positively with the protein, AS.carb, lignin, lipid, and UH compound classes, as well as with PARAFAC components C2, C3, and C5, and correlates negatively with tannin-like compounds and PARAFAC components C1 and C4. PC2 explained a small proportion of the variance in both PCAs (14% and 15%, respectively), and components PC2 and above showed no significant relationships with habitat type, site, or depth. This large proportion of unexplained variance (48% in Fig. 4a and 58% in Fig. 4b, since only PC1 was related to sample characteristics) is likely due to charge competition in FT-ICR MS, which results in nonlinear responses of observed compound relative abundances relative to true compound concentrations.

### Table 3

- **One-way ANOVA**
- **Paired t-test**

**General pore water chemistry:**
- pH: 1e−9/3e−8
- DOC: 2e−6/1e−7

**Elemental composition (FT-ICR MS):**
- MW: 0.02
- O/C: 0.02/0.03
- H/C: 6e−8/2e−7
- N/C: —/0.01
- S/C: 7e−5/0.001
- Protein: 0.002
- AS.carb: 6e−5/0.0008
- UH: —
- Lignin: 1e−5/0.0001
- Tannin: 0.005/0.004
- CA: —

**Optical properties:**
- SUVA: 0.002
- BIX: 8e−5/1e−5
- HIX: —
- Emmax: 0.009
- Fmax/A340: 0.03
- C1: 0.0003
- C2: 0.02
- C3: 0.0006
- C4: 6e−5
- C5: —

Abbreviations:
- DOC = dissolved organic carbon; MW = molecular weight; O/C, H/C, N/C, and S/C = elemental ratios. Compound classes (putative; see Table 2): AS.carb = aminosugar- and carbohydrate-like; UH = unsaturated hydrocarbon; CA = condensed aromatic. Optical properties (see Section 2.7.2): SUVA = specific UV absorbance; BIX = biological index; HIX = humification index; Emmax = emission wavelength of peak C maximum; Fmax/A340 = ratio of peak C fluorescence intensity to absorbance at 340 nm. PARAFAC components (see Section 2.7.3 and Table A.1): C1 = plant-derived humic-like; C2 and C3 = microbially-derived humic-like; C4 = high molecular weight aromatic humic-like or terrestrial reduced quinone-like; C5 = protein-like.
Fig. 2. FT-ICR MS results, presented as abundance-weighted average MW; O/C, H/C, N/C, and S/C; and relative abundances of FT-ICR MS compound classes (defined in Table 2) in each sample. Abbreviations of compound classes are as follows: AS.carb = aminosugar + carbohydrate, UH = unsaturated hydrocarbon, CA = condensed aromatic. Due to the presence of formulas that did not fit in any of the compound classes, their abundances do not add to exactly 100%. Different individual sites are denoted by color, and characteristics of the samples are given in Table 1. The % lipid-like compounds in sample E.1.26 is indicated by a number because it was much higher than that of the other samples. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

Table A.3, and regressions between FT-ICR MS results and PARAFAC component intensities are shown in Table A.4. The FT-ICR MS variables with the strongest correlations with each non-PARAFAC optical property (Table A.3) were all spectra-averaged molecular characteristics rather than compound class abundances. Among the PARAFAC
components, all components except C5 showed significant relationships with FT-ICR MS data (Table A.4). More detailed summaries of these results are provided in Section 4.

4. DISCUSSION

4.1. Differences in elemental composition between bogs and rich fens

Van Krevelen and molecular size difference plots of Stordalen bog and rich fen samples from a range of depths (10–70 cm for bogs and 7–70 cm for fens) indicate higher H/C, lower O/C, and lower molecular sizes in rich fens compared to bogs (Fig. 1). Similar differences are observed across the whole range of bog and rich fen samples (Fig. 2), including those not included in the more detailed plots shown in Fig. 1. These results extend a previous comparison of single bog and rich fen samples from ~30 cm (specifically, samples SOS.1.31 and E.1.26; Hodgkins et al., 2014) to six bog and six rich fen samples spanning five sites and a broad range of depths (Table 1).
4.2. Bulk optical properties are consistent with elemental composition

Based on linear regressions (Table A.3), most of the measured optical properties showed trends that were consistent with the elemental composition parameters that are expected from the types of molecules identified. This was despite the possibility of matrix effects affecting DOM optical properties (Sharpless and Blough, 2014). Although all of the samples were freshwater and likely had similar redox states (few differences were observed with depth; Table 3), pH differences between site types (Table 1) could have affected the differences in optical properties observed here (Tfaily et al., 2011; pH differences were unlikely to have affected FT-ICR MS results because all samples were acidified to pH = 2 prior to extraction). These caveats highlight the importance of using both methods (elemental composition and optical properties) when evaluating differences in DOM chemistry between samples.

Consistent with Stewart and Wetzel’s (1981) finding that the fluorescence/absorbance ratio of DOM correlates negatively with MW, $F_{\text{max}}/A_{340}$ of our samples was negatively correlated with average MW, representing the strongest correlation for this optical property ($R^2 = 0.27, p = 0.006$; Table A.3). Optical properties also generally predicted DOM aromaticity, but with more ambiguous results. Based on SUVA, $E_{\text{max}}$, BIX, HIX, and H/C (Figs. 2, 3, and 4a), sites with *Sphagnum* appear to contain more aromatic structures. However, the condensed aromatic (CA) compound class did not appear to show any variation between different site types (Fig. 2, Table 3), likely because it contributes little to the DOM composition observed with negative ESI (Figs. 1a and 2). CA also did not correlate with most of the optical properties, except for the expected positive correlation with SUVA ($R^2 = 0.22, p = 0.01$; Table A.3), suggesting that the trends in the other optical properties related to aromaticity ($E_{\text{max}}, \text{BIX, and HIX}$) are due to changes in other compounds. In contrast to CA, abundance-weighted average H/C and (H + O)/C, which both increase with decreasing aromaticity (the better-correlated ratio depends on whether O is predominately bonded to C by single or double bonds), were both strongly negatively correlated with SUVA, $E_{\text{max}}$, and BIX (Table A.3). These results suggest one (or both) of two conclusions. The most condensed aromatic structures, represented by the CA class, may not be easily measured by FT-ICR MS with negative ESI, likely due to their relative lack of ionizable carboxylic acid groups. Alternately, the overall aromaticity of DOM at Stordalen may be dominated by compounds with a moderate degree of aromaticity (such as lignin- and tannin-like compounds) rather than by highly condensed polyaromatic compounds (represented by the CA class).

Both PCAs—the first analyzing average molecular characteristics (Sleighther et al., 2010; Wagner et al., 2015) and bulk optical properties (Tfaily et al., 2015) (Fig. 4a), and the second analyzing compound class abundances and PARAFAC components (Tfaily et al., 2015) (Fig. 4b)—show a similarity in sample clustering in which rich fens are clearly separated from the other sites along PC1. This result highlights the robustness of both methods of measuring DOM elemental composition (average molecular characteristics vs. compound class abundances) and of both methods of measuring optical properties (commonly applied indices of SUVA, BIX, HIX, $E_{\text{max}}$, and $F_{\text{max}}/A_{340}$ vs. PARAFAC components). Thus, if any artifacts were caused by the compound class boundaries (which cannot be precisely defined) or PARAFAC components (which were defined based on data from a different peatland), these effects were likely minimal.
4.3. Possible identities of PARAFAC components

Probable compound class assignments for the PARAFAC components were determined based on the strongest positive linear correlations (Table A.4) with each single compound class defined for FT-ICR MS data (shown in Table 2 and Fig. 1a), as well as with various combinations of adjacent compound classes. In correlations of compound class abundances with PARAFAC components, a positive correlation can be interpreted more easily than a negative correlation because it suggests that both measures may represent the same set of compounds. Although these correlations do not provide absolute proof that the compound classes and PARAFAC components represent equivalent sets of compounds, the best-correlated sets of compounds are consistent with the properties expected for each component (Table A.1). Specifically, both C1 and C4 correlate best with a combination of CA and tannin-like compounds, which are more likely to be found in plant-derived DOM. In contrast, C2 correlates best with AS.carb compounds, while C3 correlates best with a combination of lipid, protein, AS.carb, and UH compounds, all of which have higher H/C consistent with microbial production. C5 did not correlate with any compound classes identifiable by FT-ICR MS (Table A.4), possibly because this region of the spectrum can be associated not just with proteins (more abundant in rich fens), but also with polyphenols such as tannins (more abundant in sites with Sphagnum (Maie et al., 2007; Hernes et al., 2009; Yamashita et al., 2011).

Across all samples, there was a strong negative correlation between the abundances of C1 and C3 ($R^2 = 0.92$, $p < 0.0001$; Fig. 5). This result is consistent with the trends in C1 and C3 in the GLAP, where C1 decreased with depth while C3 increased, and suggests that the degradation of C1 may be associated with microbial production of C3 (Tfaily et al., 2015). Similar to peatland pore water, microbial processing in lakes and rivers tends to drive PARAFAC component intensities toward fluorophores with shorter emission wavelengths; however, differences are often observed from peatland DOM in the exact excitation and emission wavelengths and overall shapes of the components (Mann et al., 2012; Walker et al., 2013; Kellerman et al., 2015). These differences could be caused by the following: (a) photobleaching in open water, which would not occur in subsurface peat pore water; (b) different pathways of DOM degradation under aerobic (open water) and anaerobic (peat pore water) conditions; and/or (c) contribution of algal-derived DOM to fluorescence in open water. In addition to these effects, the PARAFAC approach itself may also yield components that cannot be directly compared between environments. The assumption within PARAFAC that different species fluoresce independently is questionable, given that DOM fluorescence properties are influenced by charge transfer interactions between different fluorophores, dissolved ions (including pH differences), and dissolved oxygen (Sharpless and Blough, 2014). It is thus important to note that the correlations between compound classes and PARAFAC components described here are merely co-occurrences, and should not be interpreted as direct structural assignments.

4.4. Sphagnum is a primary driver of DOM properties

Based on two separate PCAs using a wide range optical properties and elemental composition data (Fig. 4), rich fens are clearly separated from the other sites, with no systematic trends between the remaining site types (collapsed palsa, bog, and poor fen) that have Sphagnum as a dominant plant species. Although a few depth trends were observed (Table 3), these trends are generally weak, as they were not consistently observed over the entire depth range in profiles with >2 depths (Figs. 2 and 3), and none of the principal components showed significant correlations with depth. This result suggests that the presence or absence of dense Sphagnum is the main driver of DOM elemental composition and optical properties at Stordalen. This interpretation is consistent with the much higher DOC concentrations in sites with Sphagnum (Table 1), which are likely due to the accumulation of recalcitrant organic compounds released by Sphagnum (Chanton et al., 2008).

DOM in sites with Sphagnum had a higher relative abundance of tannin-like and other high-O/C compounds compared to rich fens (Figs. 2 and 4). These compounds may represent Sphagnum-derived phenolics that could contribute to the suppression of decomposition at these sites (van Breemen, 1995; Verhoeven and Toth, 1995; Verhoeven and Liefveld, 1997; Freeman et al., 2004). In contrast, the more saturated, lower-oxygen compounds in rich fens may reflect the lower release of phenolics by sedges compared to Sphagnum, while the higher N/C and S/C in rich fens (Figs. 2 and 4a) may be driven by the higher N and S content of vascular plants (Hornibrook et al., 2000). Sphagnum-derived acidic carbohydrates (e.g., sphagnum), which would fall into the AS.carb class and are thought to suppress decomposition (Painter, 1983, 1991),
had very low observed abundance in all sites (Figs. 1a and 2). The low observed abundance of AS.carb compounds was likely due to their high polarity, which would have prevented their extraction with PPL cartridges.

The higher abundance of high-oxygen tannin-like compounds, which are likely to include phenolic moieties, in sites with *Sphagnum* than in rich fens, is consistent with the enzymatic latch mechanism (Freeman et al., 2001) proposed for keeping DOC concentrations high and decomposition rates low at sites with *Sphagnum*. According to this mechanism, the phenolics that inhibit decomposition are degraded by phenol oxidase, and this enzyme is inhibited by anaerobic conditions (Freeman et al., 2001, 2004) and by acidic pH (Williams et al., 2000; Tahvanainen and Haraguchi, 2013; Xiang et al., 2013). The more acidic pH of the sites with *Sphagnum* (Table 1) may thus protect phenolic compounds from decomposition by inhibiting phenol oxidase, and these phenolics may then inhibit the degradation of other compounds. Complexation of phenol oxidase with sphagnum or sphagnum acid represents one possible mechanism for its inhibition under acidic conditions, such that this enzyme may be inhibited by these specific acids rather than by pH alone. Since slow decomposition would prevent the degradation of these organic acids, the buildup of these acids, and the resulting inhibition of phenol oxidase and subsequent buildup of phenolics, may represent a self-amplifying feedback mechanism in which each compound group (*Sphagnum*-derived acids and phenolics) builds up due to the presence of the other. Both of these compound groups may then inhibit overall decomposition rates at sites with *Sphagnum*.

4.5. Microbial activity influences rich fen DOM composition

The DOM optical properties suggest that compared to the other sites, DOM from rich fens likely includes a higher proportion of microbially-produced material. Although all BIX values were <0.6, indicating that the majority of DOM in all samples was plant-derived, the higher BIX values in rich fens (Figs. 3 and 4a) suggest a higher contribution of microbially-derived DOM in these sites compared to the sites with *Sphagnum* (Huguet et al., 2009). This finding is consistent with those of Hodgkins et al. (2014), who observed higher rates of decomposition in rich fen peat compared to collapsed palsa, bog, and poor fen peat. PARAFAC components C2 and C3, which likely represent microbially-derived material (Tfaily et al., 2015), also had higher relative intensities in rich fens compared to the other sites (Figs. 3 and 4b). We can thus assume that the higher BIX and slightly lower HIX values in rich fens (Figs. 3 and 4a) indicate greater production of microbially-derived material (Kalbitz et al., 2003b; Birdwell and Engel, 2010; Tfaily et al., 2015) rather than a lower degree of DOM humification. However, the difference in HIX between sites was not significant (Table 3), indicating possible contribution of DOM humification to the variability in HIX between samples (Kalbitz et al., 2003b).

Given that microorganisms in peat are known to immobilize N in their biomass (Damman, 1988), the higher N/C in rich fens (Figs. 2 and 4a) may reflect higher microbial activity in addition to higher N/C from plant inputs at these sites. This possibility is supported by the correlation of higher N/C with optical properties indicative of microbially-produced DOM (specifically, positive correlations with BIX, C2, and C3, and negative correlations with Emmax and C1; Tables A.3 and A.4). This mechanism may amplify the effects of plant input on N/C (Hornibrook et al., 2000), as the comparatively N-rich DOM produced by sedges in rich fens would be further N-enriched by preferential microbial assimilation of N, followed by release of N-rich DOM from microbes via secretion and cell lysis. Given that S/C was also higher in rich fens (Figs. 2 and 4a) and was correlated with optical properties associated with microbial DOM (including all optical properties correlated with N/C, as well as negative correlations with SUVA and HIX; Tables A.3 and A.4), S enrichment may also occur by a similar mechanism (Novák et al., 1999). Although PCA showed a spread in rich fens along PC2 (Fig. 4a) in which some samples (E.1.26, E.2.7, and E.2.70) were specifically enriched in S/C while the remaining samples were specifically enriched in N/C, there is little evidence that these differences are related to site or depth.

The lower MW of rich fen DOM, as indicated by both average MW and F_{max}/A_{240}, may reflect more advanced decomposition at these sites. In the size-reactivity model (Amon and Benner, 1996; Burdige and Gardner, 1998) (summarized in Section 1.2), HMW-DOM is broken down into LMW-DOM, with the cutoff between these size classes defined at either 1000 Da or 3000 Da. Within the LMW-DOM size class, the effects of decomposition are generally more varied, with some studies reporting a decrease in molecular size with decomposition (e.g., Kalbitz et al., 2003b; Tfaily et al., 2013, 2015) while others report an increase (e.g., Gruber et al., 2006; Hur, 2011). Although the mass range measured in this study (m/z = 170–800 Da) falls entirely within the LMW-DOM size class, the association of lower MW with optical properties indicative of higher microbial activity (BIX, C2, and C3; Tables A.3 and A.4) suggests that the size reactivity model may apply to decomposition of the LMW-DOM pool in peatlands. Specifically, the smaller DOM molecules found in rich fens may reflect mLMW-DOM and/or pLMW-DOM that is partially degraded into mLMW-DOM, while the larger molecules found in sites with *Sphagnum* may reflect less degraded, more recalcitrant pLMW-DOM. In rich fens, more efficient decomposition of pLMW-DOM may be enabled by a greater diversity of reactions encoded by a more diverse microbial community (Mondav et al., 2014), which would increase the number of possible pathways for DOM degradation.

Our results complement and are consistent with those of Kellerman et al. (2015), who used a combination of FTICR MS and optical properties to examine DOM from 109 Swedish lakes spanning a wide range of climates and catchments. Kellerman et al. (2015) found that DOM from lakes with longer water residence times (and thus more autochthonous DOM production) had a lower oxidation state, lower aromaticity, smaller apparent MW, and higher N and S content compared to DOM from lakes with a
stronger terrestrial influence. Our data suggests that transition from partially-thawed peatlands with abundant Sphagnum to fully-thawed rich fens with little Sphagnum may result in similar changes in DOM composition as observed in lakes with long water residence times, with both leading to an increased proportion of microbially-produced DOM.

5. CONCLUSIONS

Using a combination of ultrahigh resolution mass spectrometry and comprehensive analysis of UV/Vis and fluorescence spectra, we found consistent patterns in DOM chemistry across a variety of peatland types at Stordalen Mire. In partially to fully waterlogged sites at Stordalen, the presence of dense Sphagnum moss appears to control DOM composition, such that DOM in sites with Sphagnum contains more aromatic, higher-oxygen, and higher-MW compounds. Some of these compounds, particularly tannin-like phenolics, may accumulate due to the inhibition of phenol oxidase by acidic pH. In conjunction with sphagnum and other inhibitory compounds, these phenolics then slow decomposition rates at sites with Sphagnum, keeping DOC concentrations high while reducing the rate of carbon release as CH₄ and CO₂. Conversely, in rich fens that contain little to no living Sphagnum, DOM has lower aromaticity and greater N/C and S/C consistent with more microbially-derived material. This result is consistent with Ward and Cory’s (2015) finding that lower aromatic content in permafrost DOM is associated with higher bacterial activity. At Stordalen, the higher rates of microbial activity in rich fens translate into greater CH₄ and CO₂ production at these sites (Hodgkins et al., 2014).

Taken together, these results indicate that DOM optical properties and elemental composition at Stordalen Mire vary mostly along a plant community succession associated with permafrost thaw (Johansson et al., 2006; Hodgkins et al., 2014; McCalley et al., 2014; Mondav et al., 2014), specifically the presence or absence of abundant Sphagnum that appears in the earlier thaw stages. In this way, Sphagnum acts as a “keystone” genus that disproportionately controls the biodegradability of DOM. Because Sphagnum and its litter in peat soils are thought to store more carbon than any other plant genus (Clymo and Hayward, 1982), any effects of climate change on this plant’s distribution, and consequent changes in DOM lability, could have major ramifications on a global scale. This is especially true at Arctic and boreal latitudes, which have more Sphagnum and are warming more rapidly than temperate and tropical latitudes. In permafrost peatlands with similar thaw successions as Stordalen, shifts from Sphagnum-dominated sites with thawing permafrost to fully-thawed rich fens without Sphagnum may lead to enhanced biodegradability of peatland DOM, and hence greater CH₄ emissions. Conversely, in areas where warming leads to the expansion of Sphagnum bogs, a negative climate feedback may result from increased carbon accumulation and lower CH₄ emissions (Tolonen and Turunen, 1996). Larger scale estimations of the prevalence of Stordalen-like thaw sequences in other Arctic wetlands may help to determine the global significance and direction of this warming feedback.

ACKNOWLEDGEMENTS

This research was funded by the US Department of Energy Office of Biological and Environmental Research under the Genomic Science program (Awards DE-SC0004632 and DE-SC0010580). Infrastructure for sampling was provided by the Abisko Scientific Research Station, and mass spectra were obtained at the National High Magnetic Field Laboratory FT-ICR Facility in Tallahassee, FL (Project NSF DMR-1157490). David Podgorski received support from the FSU Future Fuels Institute. Scott Saleska and Virginia Rich received additional support through the Ecosystem Genomics Initiative from the University of Arizona Technology and Research Initiative Fund, via the program in Water, Environmental and Energy Solutions.

APPENDIX A. SUPPLEMENTARY DATA

Supplementary data associated with this article can be found, in the online version, at http://dx.doi.org/10.1016/j.gca.2016.05.015.

REFERENCES


Associate editor: Thomas S. Bianchi