1   Kinetics and identities of extracellular peptidases in subsurface sediments of the White

2   Oak River Estuary, NC

Running title: Extracellular enzymes in subsurface sediments

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Abstract

Anoxic subsurface sediments contain communities of heterotrophic microorganisms that oxidize organic carbon at extraordinarily slow rates. In order to assess the mechanisms by which subsurface microorganisms access detrital sedimentary organic matter, we measured kinetics of a range of extracellular peptidases in anoxic sediments of the White Oak River estuary, NC. Nine distinct peptidase substrates were enzymatically hydrolyzed at all depths. Potential peptidase activities (V_{max}) decreased with increasing sediment depth, although V_{max} expressed on a per cell basis was approximately the same at all depths. Half-saturation constants (K_m) decreased with depth, indicating that subsurface enzymes are adapted to low substrate concentrations. Potential activities of extracellular peptidases acting on molecules that are enriched in degraded organic matter (D-phenylalanine and L-ornithine) increased relative to enzymes that act on L-phenylalanine, further suggesting microbial community adaptation to access degraded organic matter. Nineteen classes of exported peptidases were identified in genomic data from the same site, of which genes for class C25 (gingipain-like) peptidases represented more than 40% at each depth. Methionine aminopeptidases, zinc carboxypeptidases, and class S24-like peptidases, which are involved in single-stranded DNA repair, were also abundant. These results suggest a subsurface heterotrophic microbial community that primarily accesses low-quality detrital organic matter via a diverse suite of well-adapted extracellular enzymes.

IMPORTANCE

Burial of organic carbon in marine and estuarine sediments represents a long-term

sink for atmospheric carbon dioxide. Globally, ~40% of organic carbon burial occurs in

anoxic estuaries and deltaic systems. However, the ultimate controls on the amount of

organic matter that is buried in sediments, versus oxidized into CO_2, are poorly

constrained. Here we used a combination of enzyme assays and metagenomic analysis to

identify how subsurface microbial communities catalyze the first step of proteinaceous

organic carbon degradation. Our results show that microbial communities in deeper

sediments are adapted to access molecules characteristic of degraded organic matter,

suggesting that those heterotrophs are adapted to life in the subsurface.

Introduction

A large fraction of the microorganisms in subsurface sediments are heterotrophs that metabolize aged, microbially altered organic matter (Lomstein *et al.*, 2009; Jørgensen and Marshall, 2016; Biddle *et al.*, 2006). These communities' metabolisms can be more than a million-fold more slower than cells in culture (Hoehler and Jørgensen, 2013; Lomstein *et al.*, 2009). A recent meta-analysis showed that only about 12% of cells in marine sediments belonged to cultured species, while 27% belong to phyla that contain no cultured representatives (Lloyd *et al.*, 2018). Consequently, the mechanisms by which these microorganisms access detrital organic matter are poorly understood.

In surface environments, where carbon transformations are relatively rapid, heterotrophic microorganisms gain energy by metabolizing a combination of small molecules (<600-1000 Da), which can be taken up directly via general uptake porins (Benz and Bauer, 1988) and macromolecules, which must be broken down outside of the cell by extracellular enzymes. Most freshly-produced organic matter is macromolecular, and large molecules tend to be more bioavailable than small ones (Benner and Amon, 2015), so the nature and activity of extracellular enzymes present in surface environments is a major control on the rate of microbial carbon oxidation in such environments.

It is not clear whether microbial extracellular enzymes play the same role in subsurface sediments. It is conceivable that macromolecules are broken down primarily by non-enzymatic mechanisms in sediments. For instance, in soils, MnO_4 catalyzes the depolymerization of proteins without requiring enzymes. Certain bacterial species can use TonB-dependant transporters to transport polysaccharides that are substantially larger

than 600 Da into the periplasm (although enzymatic hydrolysis is still required prior to

uptake into the cytoplasm; Reintjes *et al.*, 2017). Furthermore, some of the unique aspects

of subsurface sediments suggest that extracellular enzymes might not be an effective

strategy to obtain carbon or energy. In order for the production of extracellular enzymes

to be part of a viable metabolic strategy, each enzyme must, over its lifetime, provide the

cell with at least as much carbon or energy as was required to synthesize the enzyme

(Vetter *et al.*, 1998; Allison, 2005; Schimel and Weintraub, 2003). In subsurface

sediments, where cell division times may be on the order of decades to millenia, enzyme

lifetimes would need to be correspondingly longer to remain 'profitable'. Since enzyme

lifetimes are finite, there must exist a community metabolic rate below which

extracellular enzyme lifetimes are too short to become profitable. That limit is difficult to

quantify because enzyme lifetimes in any environment are poorly constrained (e.g., Steen

and Arnosti, 2011). Thus, it is plausible that extracellular enzyme-mediated carbon

acquisition is impractical in sediments in which metabolic rates are particularly slow.

While extracellular enzyme activity in surface sediments has frequently been

reported, few reports exist of extracellular enzyme activity from deeper than 20 cm below

the seafloor (cmbsf) (Arnosti *et al.*, 2014). Enzyme activity has been reported in

sapropels up to 389 cmbsf in sapropels in the Eastern Mediterranean Sea (Coolen and

Overmann, 2000; Coolen *et al.*, 2002) and in sediment from 600-630 cmbsf in Aarhus

Bay sediments (Lloyd *et al.*, 2013a), as well as in a few other subsurface environments,

such as the interior of seafloor basalts at the Loihi seamount (Jacobson Meyers *et al.*,

2014). Furthermore, an analysis of transcriptomes from subsurface sediments of the Peru

Margin revealed diverse exported peptidases and carbohydrate-active enzymes, which

decreased in relative abundance with increasing depth (Orsi *et al.*, 2018).

In order to better understand how heterotrophic microorganisms in subsurface

sediments access organic matter, we assayed a diverse set of peptidases (protein-

degrading enzymes) in sediment cores from the White Oak River estuary, NC. We paired

these assays with analysis of the potential for extracellular peptidase production from

existing metagenomic data sets. We chose this site because the porewater geochemistry

and microbiology of these sediments has been well-characterized (Martens and

Goldhaber, 1978; Kelley *et al.*, 1990; Baker *et al.*, 2015; Lazar *et al.*, 2016; Lloyd *et al.*,

2011) and because they contain abundant Bathyarchaeota and Thermoprofundales

archaea, which appear to be capable of metabolizing detrital organic matter (Kubo *et al.*,

2012; Lloyd *et al.*, 2013a; Meng *et al.*, 2014; Zhou *et al.*, 2018b). We focused on

peptidases because protein degradation appears to be an important metabolism for some

subsurface archaea (Lloyd *et al.*, 2013a) and because peptidases were more active than

other enzymes in similar environments (Coolen and Overmann, 2000; Jacobson Meyers

et al., 2014). Because environmental samples contain a wide range of distinct peptidases

at variable activities (Obayashi and Suzuki, 2005; Steen and Arnosti, 2013) we measured

Figure 1: Saturation curves for six substrates measured using the single-cuvette reader methodology at each of six depths. Dark circles indicate "live" samples, open triangles indicate autoclaved controls. Dark lines indicate nonlinear least-squares fits to the Michaelis-Menten rate law. Substrate abbreviations are given in the column headings and are defined in Table 1. Sediment depths are listed on row headings in centimeters below sediment-water interface.

116   the hydrolysis of eleven different substrates, which may be hydrolyzed by structurally

117   and genetically diverse extracellular peptidases.

Results

PEPTIDASE KINETICS

Combining data from all three sampling dates, unambiguous hydrolysis of eleven

different peptidase substrates was observed. All peptidase substrates assayed with the

more-sensitive single-cuvette methodology were hydrolyzed much faster in untreated

sediments than in autoclaved controls. Kinetics of substrate hydrolysis were qualitatively

consistent with the Michaelis-Menten rate law, $v_0 = \frac{V_{max} \times [S]}{K_m + [S]}$ (Fig 1). Estimated V_{max}

values ranged from 40 to 3400 nmol g^{-1} sediment hr^{-1} (median 310, interquartile range

190 to 560 nmol g^{-1} sediment hr^{-1}). Throughout the core, AAF-AMC, GGR-AMC, and

Gly-AMC were hydrolyzed the fastest, and Arg-AMC was hydrolyzed slowest (Fig S1).

Summed V_{max} values decreased with depth from 9.09 μmol AMC g^{-1} sed hr^{-1} at the

surface to 1.24 μmol AMC g^{-1} sed hr^{-1}, or 13% of the surface value, at 82.5 cmbsf.

Estimated K_m values ranged from 36.1 μM to 1310 μM (median 138 μM, interquartile

range 102 to 326 μM), and trended downward (i.e., to greater substrate affinity) with

increasing depth (Fig 2). K_m values for hydrolysis of Leu-AMC were the greatest (i.e.,

lowest substrate affinity) while K_m values for hydrolysis of BocVPR-AMC, GGR-AMC,

and Arg-AMC were the least.

Figure 2: A: The sum of all peptidase V_{max} values, relative to the value at 4.5 cm, versus sediment depth. Error bars represent propagated error of the estimate of V_{max} for each substrate. B: Summed V_{max} relative to cell count. Error bars represent propagated error from summed Vmax and cell counts, and is dominated by cell count uncertainty. C: Organic carbon oxidation rates modeled from sulfate and methane profiles. D: Summed V_{max} relative to modeled carbon oxidation rates. Error bars represent error in summed V_{max} relative to organic carbon oxidation rates, for which uncertainty was not modeled.

In a separate core, hydrolysis rates of D-Phe-AMC, L-Phe-AMC, and L-Orn-

AMC were assessed. These were measured using a plate-reader technique that proved

insufficiently precise to accurately measure V_{max} or K_m, so we have only reported v_0,

which was measured at a high substrate concentration (400 μM) and therefore

approximates V_{max}. Ratios of v_0 for D-Phe-AMC : L-Phe-AMC hydrolysis and L-Orn-

AMC : L-Phe-AMC hydrolysis rates increased approximately linearly downcore (Fig 3).

MICROBIAL ABUNDANCE, CELL-SPECIFIC PEPTIDASE ACTIVITY, AND ORGANIC

CARBON OXIDATION RATES

143   Concordantly with potential activities, cell counts decreased more-or-less steadily

144   downcore from 4.5×10^8 cells ml^{-1} wet sediment at 1.5 cmbsf to 7.4×10^7 cells ml^{-1} wet

145   sediment at 82.5 cmbsf. Consequently, cell-specific total potential peptidase activity was

146   roughly constant at 32 +/- 14 amol AMC cell^{-1} hr^{-1} (Fig 2b) with no significant trend as a

147   function of depth. Most of the error in cell-specific peptidase activities results from

148   variance in cell counts rather than in V_{max} estimations.

149   Organic carbon oxidation rates were estimated using a 2-G model driven by

150   porewater methane and sulfate concentrations. The total organic carbon oxidation rate at

82.5 cmbsf was approximately 0.17% relative to that at 4.5 cmbsf (the top of the model domain), a decrease of almost 3 orders of magnitude (Fig 2c). Thus, summed V_{max} relative to G increased more than 200-fold in the deepest sediments relative to surface sediments.

Figure 3: a.) Ratio of v_0 for D-phenylalanine aminopeptidase: L-phenylalanine aminopeptidase and b.) L-ornithine aminopeptidase : L-phenylalanine aminopeptidase.

161   Samples for genomic analysis were taken from three broad sedimentary zones: the

162   sulfate reduction zone (SRZ, 8-12 cmbsf), sulfate-methane transition zone (SMTZ, two

distinct samples from nearby locations,

24-32 cmbsf and 26-30 cmbsf), and the

methane-rich zone (52-54 cmbsf; data

originally published in Baker *et al.*,

2015). A total of 3739 genes encoding

extracellular peptidases were identified

among metagenomes from the three

depth zones examined, including 685

from SRZ, 1994 from SMTZ, and 1060

Figure 4: K_m values of extracellular peptidases as a function of depth. Details of substrates and the peptidases they correspond to are in Table 1.

from MRZ. Of the genes encoding for

peptidases, 0-71% (depending on class of peptidase, algorithm and sediment depth)

contained a signal peptide and are likely secreted by the SEC-dependent transport system

(supplementary data). Among the genes associated with signal peptides, members of

peptidase family C25, belonging to the gingipain family, were by far the most abundant

at all depths, accounting for 41-45% of all SP-associated peptidases (Fig. 4). Genes

annotated as encoding extracellular methionine aminopeptidases and zinc

carboxypeptidases were also abundant (13%-19%). Together, these peptidase classes

accounted for 73%, 76%, and 73% of exported peptidases in the SRZ, SMTZ, and MRZ

respectively. The composition of protein families was generally consistent with depth,

particularly among the more-abundant peptidases. Five peptidase annotations were an

exception to this trend: peptidase family M1, peptidase family M20/M25/M40, peptidase

family M3, M61 glycyl aminopeptidase, and thermophilic metalloprotease (M29) were

found in much lower abundance at the SMTZ than the MRZ or SRZ. Given that those

correspond to differences of one or a few total reads, these are well within the range of

noise.

Discussion

189   IDENTITIES OF EXTRACELLULAR PEPTIDASES PRESENT IN WHITE OAK RIVER

190   ESTUARY SEDIMENTS

191   Kinetics of fluorogenic substrate hydrolysis was consistent with the Michaelis-

Figure 5: Exported peptidases identified in WOR sediments. Depths SRZ, SMTZ, and MRZ correspond to the sulfate-reducing zone (8-12 cmbsf), sulfate-methane transition zone (24-32 cmbsf) and methane-rich zone (52-54 cmbsf), respectively.

Menten rate law and hydrolysis rates were dramatically slower in autoclaved controls than in "live" treatments, indicating that the substrates were hydrolyzed by enzymes rather than by abiotic factors. The enzyme substrates used here encompassed a diverse range of amino acid and peptide chemistries, including polar and non-polar R groups at the P1 site (i.e., the amino

acid N-terminal to the scissile bond) and substrates with and without steric protecting

groups, which must have been hydrolyzed by endopeptidases (which cleave proteins from

within) and aminopeptidases (which cleave proteins from the N terminus) respectively.

Peptide bonds adjacent to a diverse set of amino acid residues were cleaved, including

glycine (the smallest amino acid), phenylalanine (among the largest amino acids),

arginine (positively charged at porewater pH) and leucine (uncharged, hydrophobic),

consistent with the presence of a diverse range of extracellular peptidases throughout the

core.

The metagenomic results also indicated the potential for a diverse range of

secreted peptidases, produced by a broad range of taxa, throughout the sediment column.

The metagenomic results represent a minimum estimate for the genomic potential for

extracellular peptidase production, because they rely on the assumption that only those

peptidases associated with signal peptides (SPs) are secreted. Non-SP-based enzyme

secretion pathways may also contribute to the pool of extracellular enzymes, including

Sec-independent secretion systems (Bendtsen *et al.*, 2005) and release of internal

enzymes into the extracellular medium by viral lysis (Danovaro *et al.*, 2008; Breitbart *et*

al., 2004).

The dominance of genes for exported gingipain-like endopeptidases (class C25) at

all depths is consistent with rapid hydrolysis rates of fluorogenic substrates for

endopeptidases. Gingipains are endopeptidases with preference for arginine at the P1

position (i.e., the N-terminal side of the hydrolyzed bond), which would be active

towards the substrates GGR-AMC and Boc-VPR-AMC. Those were among the fastest-

hydrolyzed substrates at each depth (Fig. 1), indicating that genes for C25 peptidases

were likely expressed. Previously, gingipains have been identified in Thermoprofundales

(formerly Marine Benthic Group D) and in Bathyarchaeota, and appear to be widespread

in marine sediments (Zhou *et al.*, 2018a; Lloyd *et al.*, 2013b; Orsi *et al.*, 2018). The M28

family, also among the most abundant annotations, contains a diverse range of

aminopeptidases and carboxypeptidases including leucine aminopeptidase, consistent

with the observed hydrolysis of Leu-AMC (Rawlings *et al.*, 2018). Genes for D-

aminopeptidases genes were observed, consistent with hydrolysis of D-Phe-AMC.

Other abundant genes were annotated as methionine aminopeptidase, zinc

carboxypeptidase, a C-terminal domain from bacterial pre-peptidases, and peptidases

from the MEROPS families M24, S24, M50, and M28. Potential activities of these

peptidases were not assayed. Zinc carboxypeptidases (M20) cleave enzymes from the

carboxy-terminus and have strong specificity for Gly at the P1 position (i.e., the position

C-terminal to the scissile bond), but little preference for the residue at the P1' position

(the position C-terminal to the scissile bond; in a carboxypeptidase this would be the C-

terminus of the protein). Methioine aminopeptidases (M24) are metallopeptidases with

preference for glycine at the P1 position.

S24 and M50 peptidases are less likely to be directly relevant to organic matter

processing. S24 peptidases are involved in the SOS response for single-stranded DNA

repair (Rawlings *et al.*, 2018). M50 peptidases are membrane-bound enzymes that act as

sporulation factors in Bacillus subtilis, and possibly other Bacteria (Green and Cutting,

2000; Feng *et al.*, 2007), and which are not secreted. However, DNA repair (Orsi *et al.*,

2013; Bird *et al.*) and spore formation (O'Sullivan *et al.*, 2015; Lomstein *et al.*, 2012)

both appear to be important survival mechanisms for microorganisms in subsurface

sediments. The bacterial C-terminal pre-peptidase domain is often found in secreted

peptidases, but is removed prior the peptidase becoming active, and could be associated

with a wide range of peptidases (Yeats *et al.*, 2003).

Each of these annotations is plausible in terms of what is known about peptidase

activities in sediments, and the annotations and observed activities generally agreed with

each other. We did not assay for carboxypeptidases (e.g. MEROPS family M20) or

methionine aminopeptidase, but carboxypeptidases have previously been observed to be

active in estuarine sediments (Pantoja *et al.*, 1997) and the generally broad substrate

specificities of extracellular aminopeptidases suggests that methionine aminopeptidases

could have contributed to the hydrolysis of the other aminopeptidase substrates (Steen *et*

al., 2015). However, a note of caution is also warranted when interpreting peptidase

annotations from deeply branching microorganisms: the high diversity of hydrolases

makes precise annotations difficult, and the exact substrate specificities of the peptidases

in these samples may differ somewhat from those inferred from the annotations

(Michalska *et al.*, 2015). Thus, while these annotations are environmentally plausible and

generally consistent with the fluorogenic enzyme assays, they should nevertheless be

viewed with some skepticism.

Heterotrophic microorganisms in subsurface sediments have little access to fresh

organic matter. In the cores described here, which represented ~275 years of sediment

deposition, organic matter reactivity decreased by at least three orders of magnitude

between the surface and 82.5 cmbsf (Fig 2c). It is challenging to determine what fraction

of high- vs low-molecular weight organic matter subsurface microorganisms metabolize.

However, the fact that cell-specific V_{max} was more-or-less constant downcore (Fig 2b)

suggests that the heterotrophic community relied on complex organic matter to a similar

degree at all depths. The cell-specific V_{max} values for Leu-AMC hydrolysis, 21-51 amol

cell^{-1} hr^{-1}, are comparable to previous measurements in active environments such as

surface sediments (2-100 amol cell^{-1} hr^{-1}) and seawater (mostly less than 100 amol cell^{-1}

hr^{-1}, but with some measurements up to 10 nmol cell^{-1} hr^{-1}; Vetter and Deming, 1994 and

references therein), consistent with communities that relied primarily on organic carbon

derived from macromolecules.

The ratio of ΣV_{max} : OC oxidation rate is sensitive to the mix of enzymes included

in the sum, and to the substrate specificity of enzymes assayed (some enzymes will be

capable of hydrolyzing multiple substrates). The absolute value of that sum, therefore, is

not particularly meaningful. The trend, however, is informative: as sediment depth

increased, the potential activity of extracellular peptidases decreased much more slowly

than the actual rate of organic carbon oxidation, so the ratio of ΣV_{max} : OC oxidation rate

increased dramatically (Fig 2d). V_{max} is a proxy for enzyme concentration, so the

observed increase in ΣV_{max} : OC oxidation rate combined with the trend in cell-specific

ΣV_{max} suggests that deeper heterotrophic communities exhibited similar demand for

detrital OM, but that those enzymes returned bioavailable hydrolysate at a much slower

rate because substrate concentrations were lower.

The White Oak River subsurface communities were similar to their surface

counterparts in terms of reliance on extracellular enzymes for bioavailable organic

carbon, although subsurface metabolisms were considerably slower. However, enzyme

kinetics and potential activities of D-phenylalanine aminopeptidase, L-phenylalanine

aminopeptidase, and L-ornithine aminopeptidase all suggested microbial community

adaptation to old, degraded organic matter in deeper sediments.

Most amino acids are biosynthesized as L-stereoisomers. As organic matter ages,

the ratio of D-amino acids to L-amino acids (D:L ratio) increases with depth, due to

abiotic racemization and increased abundance of D-amino acids derived from bacterial

cell walls (Bada *et al.*, 1970; Lomstein *et al.*, 2009). Accordingly, the potential activity of

D-phenylalanyl aminopeptidase increased relative to that of L-phenylalanyl

aminopeptidase, indicating an increased capacity to access degraded organic matter.

Ornithine, which is a product of the release of urea from arginine, is another marker for

degraded organic matter, while phenylalanine is more characteristic of fresher organic

matter (Dauwe *et al.*, 1999), and the relative potential activity of L-phenylalanine

aminopeptidase : ornithine aminopeptidase followed the same increasing trend with

depth. Finally, the decrease of K_m values with increasing depth indicates peptidases that

function more efficiently at lower substrate concentrations. It is intuitive that the

concentration of enzyme-labile organic matter concentrations would decrease downcore,

and the observed increase in ΣV_{max} : OC oxidation rate provides direct evidence of that.

Taken together, these three observations provide strong evidence for a subsurface

heterotrophic microbial community that is increasingly adapted to persist using degraded

organic matter at increasing depth.

This evidence raises the ecological question of *how* selective pressure produces a

heterotrophic community adapted to degraded organic matter. Modeling and genomic

observations in other sedimentary environments suggest that microbial growth rates are

too slow for community adaptation by enhanced growth rates of more successful taxa;

rather, communities in deeper sediments consist of taxa that were deposited at the

sediment-water interface and died at the slowest rates (Starnawski *et al.*, 2017; Bradley *et*

al., 2018). If those findings can be generalized to environments characterized by

recalcitrant organic matter, that poses a question: in which aquatic environments are

microorganisms capable gaining reproductive advantage by growing on recalcitrant

organic carbon? The studies cited above addressed sites at which sedimentation rates,

microbial respiration, and likely cell doubling times were considerably slower than in the

sediments described here, so even if growth (as opposed to persistence) on recalcitrant

organic carbon is not possible in those environments, it may have been in the White Oak

River sediments. Alternately, microbial taxa may gain adaptations to metabolize

recalcitrant organic matter in environments where labile organic matter is more abundant

and growth rates are higher. This scenario would imply that organisms which primarily

metabolize more labile organic matter would gain some selective advantage by also

metabolizing recalcitrant organic matter. Further analysis of the mechanisms by which

subsurface heterotrophs access degraded sedimentary organic matter may yield insights

into how microorganisms survive in low-energy environments, and into the controls on

organic carbon preservation in sediments over geological timescales.

Materials and Methods

STUDY SITE

Samples were collected from Station H in the White Oak River Estuary, 34°

44.490' N, 77° 07.44' W, first described by Gruebel and Martens (1984). The White Oak

River Estuary occupies a drowned river valley in the coastal plain of North Carolina.

Station H is characterized by salinity in the range of 10 to 28 and water depth on the

order of 2 m. The flux of ΣCO_2 across the sediment-water interface was 0.46 ± 0.02

mmol m^{-2} hr^{-1} (measured in May of 1987), primarily due to organic carbon oxidation via

sulfate reduction. The sediment accumulation rate averages 0.3 cm yr^{-1} (Kelley et al.,

1990). Total organic carbon content is approximately 5%. For this study, push cores of

40-85 cm were collected from Station H by swimmers on May 28, 2013, and October 22,

2014. In 2013, cores were transported to the nearby Institute of Marine Sciences

(University of North Carolina) at Morehead City, where they were sectioned and

processed for enzyme activities, porewater geochemistry, and cell counts within 6 hours

of sample collection. Porewater sulfate in 2013 was depleted by 43.5 cm, and methane

peaked at 79.5 cm (Fig. S1). In 2014, cores were transported on the day of sampling to

the University of Tennessee, Knoxville, stored at 4 °C, and processed for enzyme

activities the following day. Samples for metagenomic analysis were collected separately

in October 2010 from three sites (sites 1, 2, and 3, as previously described by Baker et al.,

2015), all of which are within 550 m of Station H.

ENZYME ASSAYS

Enzyme assays were performed using different protocols in 2013 (data presented

in Figs 1-3) versus 2014 (data presented in Fig 4). In 2013, enzyme assays were

performed according to a protocol similar to the one described in Lloyd et al (2013b).

Cores were sectioned into 3 cm intervals. The following intervals were selected for

enzyme assays: 0-3 cm, 3-6 cm, 27-30 cm, 57-60 cm, and 81-83 cm. Each section was

homogenized, and approximately 0.5 ml wet sediment was transferred into separate 5 ml

amber glass serum vials, which had been pre-weighed and preloaded with 4 ml anoxic

artificial seawater (Sigma Sea Salts, salinity = 15, pH=7.5) Samples were weighed again

to determine the precise mass of wet sediment added, and then an appropriate quantity of

20 mM peptidase substrate stock dissolved in DMSO was added, up to 90 μL, for final

substrate concentrations of 0, 25, 50, 75, 100, 200, or 300 μM. Substrates are listed in

Table 1. Triplicate incubations with 400 μM Arg-AMC, Gly-AMC, Leu-AMC and Gly-

Gly-Arg-AMC were also created, but these were omitted for Ala-Ala-Phe-AMC and Boc-

Phe-Val-Arg-AMC because the latter two substrates are considerably more expensive.

Each serum vial was vortexed, briefly gassed with N_2 to remove oxygen introduced with

the sample, and approximately 1.3 ml slurry was immediately removed, transferred to a

microcentrifuge tube, and placed on ice to quench the reaction. The precise time of

quenching was recorded. This was centrifuged at 10,000 × g within approximately 15

minutes. The supernatant was transferred to a methacrylate cuvette and fluorescence was

measured with a Turner Biosystems TBS-380 fluorescence detector set to UV mode

(λ_{ex}=365-395 nm, λ_{em}=465-485 nm). Samples were then incubated at 16 °C,

approximately the *in situ* temperature, and the sampling procedure was repeated after

approximately 3 hours. The rate of fluorescence production was calculated as the increase

in fluorescence for each sample divided by the elapsed time between sample quenching.

Killed controls were made using homogenized, autoclaved sediments from 35-45 cmbsf.

However, we note that autoclaving does not completely destroy sediment enzymes

because sorption to mineral surfaces stabilizes enzyme structure, vastly increasing their

ability to maintain a functional conformation at high temperatures (Stursova and

Sinsabaugh, 2008; Carter *et al.*, 2007; Schmidt, 2016). We therefore used the autoclaved

samples as a qualitative control for the null hypothesis that enzymes were responsible for

none of the observed substrate hydrolysis, rather than as a quantitative method to

distinguish enzymatic substrate hydrolysis from potential abiotic effects. In some

sediments, a large fraction of fluorophore can sorb to particles, requiring a correction to

observed fluorescence (Coolen *et al.*, 2002; Coolen and Overmann, 2000), but we

observed negligible sorption of fluorophore to the White Oak River sediments.

In 2014, enzymes were assayed using a protocol based on the approach of Bell et

al (2013), which was designed for soil enzyme assays. In this approach, peptidase

substrates were mixed with sediment-buffer slurries in 2-mL wells of a deep-well plate.

These plates were periodically centrifuged and 250 μL aliquots of supernatant were

transferred into a black 96-well microplate. Fluorescence was read using a BioTek

Cytation 3 microplate reader (λ_{ex} = 360 nm, λ_{em} = 440 nm). Results from this method

proved considerably noisier than the single-cuvette method used in 2013, so kinetic

parameters (V_{max} and K_m) were not calculated for these data. Nevertheless, results were

qualitatively similar to those from 2013, and we have reported V_{max} from 2014 as v_0

measured at 400 µM substrate concentration, which was saturating. In June 2014, the

following substrates were assayed: AAF-AMC, Arg-AMC, Boc-VPR-AMC, D-Phe-

AMC, Gly-AMC, Leu-AMC, L-Phe-AMC, Orn-AMC, Z-Phe-Arg-AMC, and Z-Phe-Val-

Arg-AMC. In October 2014, L-Phe-AMC, D-Phe-AMC, and Orn-AMC were assayed

according to the same protocol in 3-cm core sections at 1.5, 4.5, 7.5, 10.5, 19.5, 22.5,

25.5, 28.5, 34.5, 37.5, 40.5, 43.5, 49.5, 52.5, 58.5, and 61.5 cmbsf.

Peptidase kinetic data were analyzed using R. All raw data and scripts related to

enzyme analysis are posted at http://github.com/adsteen/WOR_enz_2013_2014. For

samples taken using the more sensitive single-cuvette method, Michaelis-Menten

parameters were estimated from nonlinear least squares fits to kinetic data. In the case of

Leu-AMC at 4.5 centimeters below seafloor, kinetic data could not successfully be fit to a

Michaelis-Menten function, so no K_m was reported and the value of v_0 at the highest

substrate concentration was substituted for V_{max}. For analysis of correlations, data sets

were qualitatively evaluated for homoskedasticity and normality of residuals using q-q

plots and plots of residuals vs fitted values. When untransformed data met those criteria,

the null hypothesis of no correlation was tested using linear least-squares regressions.

When untransformed data failed to meet those criteria, data were first log-transformed. In

cases in which log-transformed data were wither heteroskedastic or residuals were non-

normally distributed, data were rank-transformed and correlations were tested using

Spearman's ρ.

GEOCHEMICAL AND MICROBIOLOGICAL MEASUREMENTS

Sediment porosity was measured by mass after drying at 80 °C, according to the

equation

$$\phi = \frac{m_w/\rho_w}{m_w/\rho_w + \frac{m_d - S \times m_w/1000}{\rho_{ds}}}$$

Here, m_w represents mass lost after drying, ρ_w represents the density of pure

water, m_d represents the mass of the dry sediment, S represents salinity in g kg^{-1}, and ρ_{ds}

represents the density of dry sediment (assumed to be 2.5 g cm^{-3}). Porewater sulfate

concentrations were measured using a Dionex Ion Chromatograph (Sunnyvale, CA) in

porewater that was separated by centrifugation in 15 ml centrifuge tubes at $5000 \times g$ for 5

minutes, filtered at 0.2 μm, and acidified with 10% HCl. Methane was measured using 3

ml sediment subsamples that were collected from a cutoff syringe, entering through the

side of a core section, immediately after core extrusion. Subsamples were deposited

immediately in a 20 ml serum vial containing 1 ml, 0.1 M KOH. These were immediately

stopped and shaken to mix sediment with KOH. Methane was later measured by

injecting 500 μl of bottle headspace into a GC-FID (Agilent, Santa Clara, CA) using a

headspace equilibrium method (Lapham *et al.*, 2008).

GEOCHEMICAL MODELING

Organic carbon remineralization rates as a function of depth were estimated by

applying a multi-component reaction-transport model to depth distributions of sulfate and

methane concentration. The model is based on equations described in Boudreau

(Boudreau, 1996) and includes only sulfate reduction and methane production due to lack

of data regarding oxic and suboxic processes. Thus, the model is limited to depths greater

than 4.5 cm where sulfate reduction and methane production are the dominant processes,

and bioirrigation and bioturbation may be assumed to be negligible. The organic matter

remineralization rate is parameterized using the multi-G model first proposed by

Jørgensen (1978); a two-component model was sufficient to accurately simulate the

sulfate and methane data. For solutes, the upper boundary conditions were measured

values at 4.5 cm while the lower boundary conditions (200 cm) were set to zero-gradient.

The flux of reactive organic carbon to 4.5 cm was calculated from the sulfate flux across

the 4.5 cm horizon and an estimate of methane burial below the lower boundary (the

methane flux at the upper boundary was observed to be zero), with an assumed oxidation

state of reactive carbon of -0.7. The model contains four adjustable parameters that are

set to capture the major details of measured sulfate and methane data: first-order rate

constants for both fractions of the reactive carbon pool; the partitioning factor for both

fractions, and the rate constant for methane oxidation.

CELL ENUMERATION

Cells were enumerated by direct microscopic counts. One mL of sediment was

placed in a 2-mL screw-cap tube with 500 μl of 3% paraformaldehyde in phosphate

buffered saline (PBS), in which it was incubated overnight before being centrifuged for 5

minutes at 3000 × *g*. The supernatant was removed and replaced with 500 µl of PBS,

vortexed briefly and centrifuged again at 3000 × *g*. The supernatant was subsequently

removed and replaced with a 1:1 PBS:ethanol solution. Sediments were then sonicated at

20% power for 40 seconds to disaggregates cells from sediments and diluted 40-fold into

PBS prior to filtration onto a 0.2 µm polycarbonate filter (Fisher Scientific, Waltham,

MA) and mounted onto a slide. Cells were stained with 4',6-diamidino-2-phenylindole

(DAPI) and enumerated by direct counts using a Leica Epifluorescence Microscope.

METAGENOMIC ANALYSIS

To resolve the taxonomic distribution of extracellular peptidases we searched a

pre-existing White Oak River *de novo* assembled and binned metagenomic dataset (Table

S2; Baker *et al.*, 2015) for genes that were assigned extracellular peptidase functions.

These assignments were based on best matches to extracellular peptidases in KEGG,

pfam, and NCBI-nr (non-redundant) databases using the IMG annotation pipeline

(Markowitz et al., 2014). Genes were additionally screened for signal peptidase motifs

using the following programs: PrediSI setting the organism group to gram-negative

bacteria (Hiller *et al.*, 2004), PRED-Signal trained on archaea (Bagos *et al.*, 2009), the

standalone version of PSORT v.3.0 trained against archaea (Yu *et al.*, 2010), and SignalP

4.1 using gram-negative bacteria as the organism group (Petersen *et al.*, 2011). All

programs were used with default settings if not stated otherwise. Binned genomes from

three different depth zones of White Oak River sediments were examined. The sulfate-

rich zone (SRZ) genomes were obtained from sites 2 and 3 core sections 8-12 and 8-10

cm, respectively. The sulfate-methane transitions zone (SMTZ) genomes were recovered

from site 2 and 3 and depths of 30-32 cm and 24-28 cm. The methane-rich zone (MRZ)

was from site 1 and 52-54 cm.

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Conflict of interest

497   The authors declare no competing financial interests in relation to the work

498   described.

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655.

Table

Table 1: Substrates used in this study and the enzymes that hydrolyze them. AMC stands

for aminomethylcoumarin, the moiety that becomes fluorescent after hydrolysis of the

peptide bond. All amino acids are in the L- stereoconformation unless otherwise noted.

Enzymes are described as "putative" because the substrate specificity of many

environmental peptidases is fairly broad, so multiple peptidases may hydrolyze any given

substrate.

Substrate	Abbreviation	Putative enzyme
L-arginine-7-amido-4-methylcoumarin	Arg-AMC	arginyl aminopeptidase
L-glycine-7-amido-4-methylcoumarin	Gly-AMC	glycyl aminopeptidase
L-leucine-7-amido-4-methylcoumarin	Leu-AMC	leucyl aminopeptidase
Carboxybenzoyl-glycine-glycine-arginine-7-amido-4-methylcoumarin	Z-GGR-AMC	gingipain and other endopeptidases
Alanine-alanine-phenylalanine-7-amido-4-methylcoumarine	AAF-AMC	clostripain and other endopeptidases
Boc-valine-proline-arginine-AMC	Boc-VPR-AMC	gingipain and other endopeptidases
D-phenylalanine-AMC	D-Phe-AMC	D-phenyalanine aminopeptidase
L-phenylalanine-AMC	L-Phe-AMC	L-phenylalanine-aminopeptidase
Ornithine-AMC	Orn-AMC	ornithine aminopeptidase

Figures and Legends

Figure 1: Saturation curves for six substrates measured using the single-cuvette reader methodology at each of six depths. Dark circles indicate "live" samples, open triangles indicate autoclaved controls. Lines indicate nonlinear least-squares fits to the Michaelis-Menten rate law. Substrate abbreviations are given in the column headings and are defined in Table 1. Sediment depths are listed on row headings in centimeters below sediment-water interface.

Figure 2: a.) The sum of all peptidase Vmax values, relative to the value at 4.5 cm, versus

sediment depth. Error bars represent propagated error of the estimate of V_{max} for each

substrate. b.) Summed V_{max} relative to cell count. Error bars represent propagated error

from summed Vmax and cell counts, and is dominated by cell count uncertainty. c.)

Organic carbon oxidation rates modeled from sulfate and methane profiles. d.) Summed

V_{max} relative to modeled carbon oxidation rates. Error bars represent error in summed

V_{max} relative to organic carbon oxidation rates, for which uncertainty was not modeled.

Fig 3: a.) Ratio of v_0 for D-phenylalanine aminopeptidase: L-phenylalanine

aminopeptidase and b.) L-ornithine aminopeptidase : L-phenylalanine aminopeptidase.

Fig 4: K_m values of extracellular peptidases as a function of depth. Details of substrates

and the peptidases they correspond to are in Table 1.

Fig 5: Frequency of reads for genes of various classes of extracellular peptidases that

were associated with signal peptidases, relative to all genes for extracellular peptidases at

that depth. SRZ, SMTZ, and MRZ correspond to "sulfate reducing zone" (8-12 cmbsf),

"sulfate-methane transition zone" (24-32 cmbsf) and "methane-rich zone" (24-28 cmbsf)

respectively.

Supplemental Information

Fig S1: V_{max} and K_m values.

Supplemental file: annotations of all peptidases included in this manuscript.

depth, cmbsf

a

$\sum V_{max}$, relative to surface

b

$\sum V_{max}$ per cell, amol cell^{-1} hr^{-1}

c

OC oxidation rate, μM hr^{-1}

d

$\sum V_{max}$:OC oxidation rate