Nature and kinetics of extracellular peptidases in subsurface sediments of the White Oak

River estuary, NC

Andrew D. Steen[1,*], Richard T. Kevorkian[2], Jordan T. Bird[2], Nina Dombrowski[3], Brett J.

Baker[3], Shane M. Hagen[1], Katherine H. Mulligan[1,4], Jenna M. Schmidt[1], Austen T.

Webber[1], Marc J. Alperin[5]

[1]Department of Earth and Planetary Sciences, University of Tennessee, Knoxville,

Tennessee, USA

[2]Department of Microbiology, University of Tennessee, Knoxville, Tennessee, USA

[3]Department of Marine Science, University of Texas-Austin, Marine Science Institute,

Port Aransas, Texas, USA

[4]Department of Biology, University of North Carolina at Chapel Hill, Knoxville,

Tennessee, USA

[5]Department of Marine Sciences, University of North Carolina at Chapel Hill, USA

[*]Corresponding: email: asteen1@utk.edu

18   mail: 306 EPS Building, 1412 Circle Drive, Knoxville, TN 37996-1410

19   tel: (865) 974-4014, fax: 865 974 2638

Abstract

Microbial communities inhabiting subsurface sediments contain abundant heterotrophs, which oxidize organic matter to obtain carbon and energy. Subsurface sediments contain very low concentrations of bioavailable compounds, and it is not clear what fraction of sedimentary organic matter heterotrophs are able to access. To gain a more mechanistic understanding of heterotrophy in subsurface sediments of the White Oak River, NC, we examined the genetic potential for extracellular peptidase production encoded within metagenome-assembled genomes, and experimentally assayed the kinetics of a wide range of extracellular peptidases in bulk sediments. We identified genes coding for at least 15 classes of extracellular peptidases and observed enzyme-catalyzed hydrolysis of 11 different peptidase substrates. Potential activities (V_{max}) of extracellular peptidases decreased downcore, but cell-specific V_{max} was relatively constant and similar to values observed in seawater phytoplankton blooms. Half-saturation constants also decreased downcore, and the relative contribution of enzymes relevant to degraded organic matter increased with increasing depth. Together, these results suggest a subsurface community that accesses organic matter using mechanisms that are basically similar to those in surface environments, but which is adapted to the highly degraded organic matter that is present in the subsurface.

Introduction

Marine sediments are one of the largest microbial environments on earth

(Kallmeyer *et al.*, 2012). Many sedimentary microbes appear to be heterotrophs, slowly

metabolizing organic matter (Jørgensen and Marshall, 2016; Biddle *et al.*, 2006), but the

mechanisms by which these heterotrophs access old, unreactive organic carbon remain

poorly characterized.

In surface environments, where photosynthesis fuels carbon fixation,

heterotrophic microorganisms gain energy from 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. Because most freshly-produced organic matter is macromolecular

and large molecules tend to be more bioavailable than small ones (Benner and Amon,

2015), 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. Extracellular peptidase activity has been identified in sapropels up

to 389 cm below seafloor (cmbsf) in the eastern Mediterranean Sea (Coolen and

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

sediments (Lloyd *et al.*, 2013b) and in the interior of seafloor basalts at the Loihi

seamount (Jacobson Meyers *et al.*, 2014). However, the nature and relative importance of

extracellular enzymes in subsurface environments remains poorly constrained because

these studies examined few samples, and only limited enzymatic classes and sample

numbers were assayed. It is possible that, like heterotrophs in surface environments,

heterotrophs in subsurface environments mainly gain access to organic carbon via

extracellular enzymes. On the other hand, some of the unique aspects of subsurface

sediments suggest that extracellular enzymes might not be an effective strategy to obtain

carbon or energy. First, subsurface sediments contain markedly fewer bioavailable

compounds such as amino acids and sugars than do surface sediments (Burdige, 2007).

Therefore, these compounds may be insufficiently abundant to be viable heterotrophic

substrates. Second, 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

metabolic rates may be orders of magnitude slower at the surface, enzyme lifetimes

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

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

lifetimes are too short to become profitable. Unfortunately that limit is difficult to

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

and Arnosti, 2011).

It is also possible that extracellular enzymes are not the primary mechanism by

which heterotrophs access sedimentary organic matter. More exotic mechanisms, such as

abiotic liberation by reactive species formed from radioactive decay of naturally-present

radioisotopes (e.g. Blair *et al.*, 2007) are conceivable, but have not been demonstrated.

The goals of this work are to determine whether subsurface heterotrophic communities

make substantial use of extracellular enzymes to access organic matter, and if so, to

characterize the set of enzymes used.

We investigated genes for extracellular enzymes and the activities of

corresponding enzymes in sediments of the White Oak River, NC by estimating the

presence of peptidase families in available metagenomic sequencing data and

complementing this by measuring the hydrolysis rates of eleven potential peptidase

substrates. This site was chosen 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 Marine Benthic Group D archaea, which appear to be

capable of metabolizing detrital organic matter (Kubo *et al.*, 2012; Lloyd *et al.*, 2013b;

Meng *et al.*, 2014). We focused on peptidases because protein degradation appears to be

an important metabolism for some subsurface archaea (Lloyd *et al.*, 2013b) 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 the hydrolysis of eleven different substrates that

may be hydrolyzed by structurally and genetically diverse extracellular peptidases. By

measuring potential activities (i.e., the capacity of the enzyme to catalyze hydrolysis if

substrate concentrations were not limiting) and substrate affinities of microbial

extracellular enzymes, we illuminated some of the mechanisms by which subsurface

heterotrophic communities access organic carbon.

Materials and Methods

 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, and 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, June 14, 2014,

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.

 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

SignalIP 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. Results are provided

in Supplementary Table 1.

In total, 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. Many of these

genes were binned to Bacteria (Baker et al. 2015) and Archaea community members

(Baker *et al.*, 2016; Lazar *et al.*, 2016; Seitz *et al.*, 2016). Taxonomic assignments of

peptidases identified in the community were based on this binning information. However,

since not all of the peptidases were binned, we used top matches to NCBI to identify the

unbinned genes. The majority of the Archaea present in the shallow (8-12 cm) sulfate-

rich zone were confidently binned, thus were used to determine the relative contributions

of archaeal extracellular peptidases (Fig 1b). A smaller proportion (68% of SRZ, 24% of

SMTZ, and 27% of MRZ) of the bacterial peptidase genes were confidently binned,

therefore, classification was based on top BLAST hits to NCBI. These classifications

were then refined using the bin assignments.

ENZYME ASSAYS

Enzyme assays were performed using different protocols in 2013 versus 2014. 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.

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"

(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 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.

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 × 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

stoppered 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).

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

$3000 \times g$. The supernatant was removed and replaced with 500 μl of PBS, vortexed

briefly and centrifuged again at $3000 \times g$. 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.

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 (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.

Results and Discussion

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 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 Table 3). Among

the genes with signal peptidases, zinc carboxypeptidases, peptidases of class C25 and

genes of the clostripain family are most likely to be secreted. Alternative secretion

pathways, which do not rely on signal peptides and therefore would not be identified by

the algorithms we used, are also known to exist. These include Sec-independent secretion

systems (Bendtsen *et al.*, 2005) and release of cytoplasmic enzymes by cell lysis. The

relative importance of these pathways in sediments has not been assessed. At all three

depths, peptidases of class C25, belonging to the gingipain family, were the most

abundant extracellular peptidases (Fig 1a). Gingipains are endopeptidases (i.e., enzymes

that cleave proteins mid-chain rather than from the N- or C-termini) with strong

specificity for the residue arginine on the N-terminal side of the scissile bond (Rawlings

and Barrett, 1999). Bathyarchaeota, which are abundant in sediments of the White Oak

River, use extracellular gingipain (among other peptidases) to degrade detrital protein-

like organic matter (Kubo *et al.*, 2012; Lloyd *et al.*, 2013b). Genes annotated as encoding

extracellular methionine aminopeptidases and zinc carboxypeptidases were also

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

genes for clostripain (another endopeptidase with preference for Arg N-terminal to the

scissile bond) and S24 peptidases (a regulatory peptidase involved in the SOS stress

response) were slightly enriched in the SMTZ.

In SRZ, Bacteria accounted for 61% of genes for extracellular peptidases for

which a lineage could be assigned, increasing to 69% in MRZ (Fig 1b). Archaeal

peptidases decreased from 39% in SRZ to 30% in MRZ, while Eukaryota accounted for

<1.5% at all depths. Consistent with this distribution, bacteria make up ca. 40-60% of

cells in White Oak River sediments as determined by CARD-FISH counts, with no

pronounced depth trend (Lloyd *et al.*, 2013a). Since the majority of the archaeal

peptidases in the SRZ were assigned to genomic bins, we were able to accurately classify

them. Interestingly, over half of the archaeal SRZ peptidases belong to marine benthic

group D (MBG-D) genomic bins (Fig 1c)(Lazar *et al.*, 2016). Additionally, a large

portion of the peptidases in the SRZ belong to newly described Archaea belonging to the

Asgard superphylum (Zaremba-Niedzwiedzka *et al.*) including Lokiarchaeota (Spang *et*

al., 2015) and Thorarchaeota (Seitz *et al.*, 2016). Little is known about the ecological

roles of these novel Archaea. It has recently been shown that they contain metabolic

pathways for the degradation of proteins and acetogenesis (Seitz *et al.*, 2016). Sources of

bacterial peptidases varied less with depth than sources of archaeal peptidases, with

Proteobacteria, Bacteroidetes, and Firmicutes among the dominant phyla (Fig 1d). 13-

20% of extracellular peptidase genes belonged to bacteria but could not be assigned a

phylum, and 19-22 phyla contributed extracellular peptidases at each depth.

Deltaproteobacteria and Gammaproteobacteria dominated the SRZ community at the

White Oak River (Baker *et al.*, 2015), and accordingly the largest portion of the

extracellular peptidases also belong to them. These phylogenetic groups decrease with

depth in the SMTZ and MRZ, however, Deltaproteobacteria contribute a large portion of

extracellular peptidases downcore. Although these groups are commonly thought to rely

on sulfur and nitrogen respiration, these genomes were shown to contain metabolic

pathways for the degradation and fermentation of organic carbon (Baker *et al.*, 2015).

Additionally, several phyla thought to be involved in fermentation of detrital carbon

including *Clostridia*, *Bacteroidetes*, *Planctomycetes*, and the candidate phyla radiation

(CPR) constitute larger portions of the peptidases in the SMTZ and MRZ.

Bioinformatic tools are a powerful way to investigate the potential of microbial

communities to oxidize complex organic molecules, but these tools do not provide

information about the expression level of genes or the in situ activities of gene products

and annotation algorithms often fail to identify the precise function of hydrolases,

particularly in deeply-branching lineages (e.g., Michalska *et al.*, 2015). Thus, we

measured the potential activities of a wide range of extracellular peptidases in the WOR

sediments. In 2013 (when the assay protocol used was more sensitive), all six peptidase

substrates tested were hydrolyzed faster in untreated sediments than in autoclaved

controls (Fig 2, Fig S2). Kinetics of substrate hydrolysis were qualitatively consistent

with the Michaelis-Menten rate law,

$v_0 = (V_{max} + [S])/(K_m + [S])$,

which is characteristic of purified hydrolases as well as mixtures of isofunctional

enzymes in environmental samples (Steen *et al.*, 2015; Sinsabaugh *et al.*, 2014).

Together, these lines of evidence show that the observed substrate hydrolysis was due to

extracellular peptidases rather than abiotic factors. Combining data from all three

sampling dates, unambiguous hydrolysis of eleven different peptidase substrates was

observed.

The diverse substrates used in this study were apparently cleaved by a wide

variety of peptidases, including aminopeptidases (peptidases that cleave an N-terminal

amino acid from a protein) and endopeptidases (peptidases that cleave internal peptide

bonds). 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). Individual extracellular peptidases can often accept a fairly broad range of

substrates, and sometimes a substrate may primarily be hydrolyzed by an enzyme that

exhibits maximal activity towards a different substrate. For instance, in pelagic samples

from Bogue Sound, an estuary in North Carolina, the substrate Leu-AMC, which is

putatively a substrate for leucine aminopeptidase, was hydrolyzed more by arginine

aminopeptidase than by leucine aminopeptidase (Steen *et al.*, 2015). Nevertheless, the

large diversity of substrates that were hydrolyzed in this study suggests a diverse set of

extracellular peptidases present in sediments, as has previously been observed in pelagic

samples (Obayashi and Suzuki, 2008, 2005; Steen and Arnosti, 2013).

340   Bulk potential enzyme activities (V_{max} values) decreased with increasing depth

(Fig 3a). However, no trend was evident in V_{max} values expressed per cell (Fig 3b), and

V_{max} expressed relative to bulk organic carbon oxidation rate increased downcore by more

than two orders of magnitude (Figs 3c-d). V_{max} per cell was approximately 100-200 amol

cell^{-1} hr^{-1} throughout the core, comparable to previous measurements made in a surface

sediments (2-100 amol cell^{-1} hr^{-1}) and surface 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).

Modeled organic carbon oxidation due to sulfate reduction and methane production

decreased from 9315 μmol C (g wet sediment)$^{-1}$ hr^{-1} at 4.5 cmbsf (the top of the model

domain) to 18 μmol C (g wet sediment^{-1}) hr^{-1} at 82.5 cmbsf, a decrease of a factor of 518.

The sum of V_{max} of all peptidases measured in 2013 decreased from 94.7 μmol g sed^{-1} hr^{-1}

at 1.5 cmbsf to 12.8 μmol g sed^{-1} hr^{-1} at 82.5 cmbsf depth, a decrease of a factor of 7.4.

The value of the V_{max}:OC oxidation rate ratio is sensitive to the precise set of enzyme

included in the sum (it is likely that some peptidases exist in the sediment that were not

measured, and that some promiscuous enzymes are double-counted because they

hydrolyze more than one substrate), but the trend is clear: as sediment depth increased,

the potential activity of extracellular peptidases increased faster than the actual rate of

organic carbon oxidation. V_{max} is a proxy for the concentration of enzymes in an

environment, so the observed increase in V_{max} relative to OC oxidation rate suggests that

subsurface microbial communities produced similar quantities of enzyme per cell as

surface communities, but that those enzymes returned less bioavailable organic matter,

presumably due to lower substrate concentrations.

362   In the sense of bulk activities, therefore, subsurface heterotrophic communities in

WOR seem to be similar their surface counterparts in terms of reliance on extracellular

enzymes to access organic matter, although metabolisms are slower in the subsurface.

However, there are indications from enzyme kinetics and activities of specific enzymes

that subsurface communities are specialized for their environment. Peptidase K_m values

decreased with increasing depth (Fig 4, Fig S3, ANCOVA, $F(5, 22)=4.44$, $p < 0.05$).

When substrate concentrations are considerably less than K_m, as they likely are in

subsurface sediments, *in situ* substrate hydrolysis rates are controlled more by K_m than by

V_{max} (Steen and Ziervogel, 2012; Cornish-Bowden, 2012), so decreasing K_m values may

be an adaptive response to low bioavailable substrate concentrations in subsurface

sediments (Sinsabaugh *et al.*, 2014).

373   The trends in K_m warrant some discussion of how K_m values are measured and

374   interpreted. As with all environmental enzyme assays, our measurements relied on

375   measuring the rate of reaction of an artificial substrate that was added to the sample. The

376   sample also contained some quantity of natural substrate (i.e., proteins or peptides). This

377   naturally-present substrate can be viewed as a competitive inhibitor of the added artificial

378 substrate (Cornish-Bowden, 2012). As a consequence, the K_m we measure ($K_{m,app}$) is

379 actually the sum of the true enzyme K_m plus the concentration of *in situ* enzyme substrate

380 ([S_{is}]); $K_{m,app} = K_m + [S_{is}]$. It is therefore possible that the observed decrease in $K_{m,app}$

381 actually reflects a change in the concentration of natural peptidase substrates with depth,

382 rather than a change in substrate affinity of peptidases. [S_{is}] is extremely difficult to

383 measure (or even to define precisely) because it reflects the sum of the concentrations of

384 all of the individual molecules which can act as substrate for a given peptidase, weighted

385 by the degree to which the peptidase is capable of accessing each substrate. [S_{is}] almost

386 certainly decreased downcore, since a decrease in the concentration of enzymatically-

387 hydrolyzable, protein-like organic matter is a diagnostic feature of aged organic matter

388 (Amon *et al.*, 2001; Dauwe *et al.*, 1999). However, it is very unlikely that [S_{is}] could have

389 decreased enough to cause more than a negligible fraction of the observed decreases in

390 K_m, which were ranged from 110 μM (GGR-AMC) to 990 μM (AAF-AMC). Sediments

391 of Aarhus Bay, which are similar to those of the White Oak River in terms of organic

392 matter content and grain size, are characterized by porewater dissolved combined amino

393 acid concentrations in the range of 50-150 μM L^{-1} (Pedersen *et al.*, 2001). Only a fraction

394 of that dissolved protein-like material can act as substrate for any given peptidase. 10-

395 40% of total combined amino acids in sediments were accessible to a peptidase which

396 was specifically selected because of its ability to hydrolyze diverse proteins (Dauwe *et*

397 *al.*, 1999), and a smaller fraction may be available to the individual peptidases measured

398 in this study. Furthermore, K_m values were measured in a 9:1 buffer:sample slurry, so

399 [S_{is}] in the slurry would be one-tenth of *in situ* [S_{is}]. By multiplying assumed porewater

400 substrate concentrations (from Aarhus Bay) by the fraction that is enzymatically

hydrolysable and dividing by 10 to account for dilution in the slurry, we estimate that

changes in $[S_{is}]$ could account for, at most, a 0.5-6 µM decrease in $K_{m,app}$. The remainder

of the 110-990 µM decrease must be due to changes in the true substrate affinity K_m,

indicating that subsurface enzymes hydrolyze low concentrations of protein considerably

more efficiently than surface enzymes.

The suite of specific peptidases that are active in deeper sediments also seems to

reflect adaptation to more degraded organic matter. In order to assess the degree to which

sedimentary extracellular peptidases target more recalcitrant organic matter, rather than

the extremely small pool of relatively labile organic matter, in 2014 we compared

potential activities of D-phenylalanyl aminopeptidase (D-PheAP) and ornithine

aminopeptidase (OrnAP) to those of L-phenylalanyl aminopeptidase (L-PheAP). D-

phenylalanine and ornithine are both markers for degraded organic matter. Most amino

acids are biosynthesized as L-stereoisomers. D-stereoisomers in sedimentary organic

matter can be produced via abiotic racemization of biomass (Bada and Schroeder, 1975;

Steen *et al.*, 2013) or by bacterial reprocessing of phytoplankton-derived OM (Pedersen

et al., 2001; Kaiser and Benner, 2008; Lomstein *et al.*, 2006). Ornithine is non-

proteinogenic amino acid, which does not exist in high concentration in fresh biomass,

but which can be produced in sediments via deamination of arginine and therefore

indicates OM degradation (Hare, 1968; Lee and Cronin, 1984). L-phenylalanine is among

the most recalcitrant amino acids (Dauwe and Middelburg, 1998) but is more labile than

D-phenylalanine or ornithine, so we take L-PheAP activity as a marker of the

community's ability to access relatively fresh OM.

D-PheAP:L-PheAP and OrnAP:L-PheAP ratios increased significantly with depth

(Fig 5, D-PheAP:L-Phe-AP: $p<0.01$, $r^2=0.35$, n=16; OrnAP:L-PheAP: $p < 0.05$, $r^2=0.26$,

n=16), indicating that subsurface communities expressed peptidases that release amino

acids from relatively recalcitrant organic matter. These results do not indicate the actual

flux of organic matter to communities: V_{max} reflects a potential rate at saturating substrate

concentrations, not an *in situ* rate. Nevertheless, these results suggest that deeper

heterotrophic communities seek to access the larger pool of more recalcitrant organic

matter rather than the smaller pool of more labile organic matter.

In summary, genetic analysis and enzyme assays reveal that a diverse set of

extracellular peptidases are present throughout the top 83 cm of White Oak River estuary

sediments, and that these enzymes are produced by a wide range of Bacteria and Archaea.

Surface and subsurface communities express similar quantities of extracellular peptidase

per cell, but subsurface peptidases catalyze reactions much more slowly, likely because

concentrations of enzyme-labile substrates are much lower in the subsurface. This implies

selective pressure on subsurface communities to produce extracellular enzymes that

function more effectively in the presence of low concentrations of more-degraded organic

matter. Accordingly, subsurface enzymes showed higher affinity (lower K_m) than surface

enzymes, and the mix of subsurface enzymes shifted towards enzymes such as D-

phenylalanine aminopeptidase and ornithine aminopeptidase. These shifts are consistent

with community-level adaptation to heterotrophic communities to subsurface organic

matter, although we lack data as to whether that adaptation is caused by evolution of

surface communities as sediment is buried or selective survival of taxa that are more

capable of accessing degraded organic matter. Further analysis of the mechanisms by

which subsurface heterotrophs access sedimentary organic matter may yield continued

insights into how heterotrophic microorganisms live in low-energy environments such as

subsurface sediments.

Acknowledgements

We thank Michael Piehler for access to his laboratory facilities at the University of North Carolina Institute of Marine Sciences, the captain of the R/V Capricorn for sampling assistance, and Terry Hazen for use of lab equipment at the University of Tennessee. We thank Oliver Jeffers for reminding ADS that science is cool. Funding for KHM was provided by NSF grant DBI-1156644 to Steven W. Wilhelm. Funding for ADS was provided by NSF grant OCE-1431598 and a C-DEBI subaward. This work is C-DEBI Contribution number <<to be determined.>>

Conflict of interest

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

References

References

Allison SD. (2005). Cheaters, diffusion and nutrients constrain decomposition by

microbial enzymes in spatially structured environments. *Ecol Lett* **8**: 626–635.

Amon RMW, Fitznar H-P, Benner R. (2001). Linkages among the bioreactivity, chemical

composition, and diagenetic state of marine dissolved organic matter. *Limnol*

Oceanogr **46**: 287–297.

Bada JL, Schroeder RA. (1975). Amino acid racemization reactions and their

geochemical implications. *Naturwissenschaften* **62**: 71–79.

Bagos PG, Tsirigos KD, Plessas SK, Liakopoulos TD, Hamodrakas SJ. (2009).

Prediction of signal peptides in archaea. *Protein Eng Des Sel* **22**: 27–35.

Baker BJ, Lazar CS, Teske AP, Dick GJ. (2015). Genomic resolution of linkages in

carbon, nitrogen, and sulfur cycling among widespread estuary sediment bacteria.

Microbiome **3**: 14.

Baker BJ, Saw JH, Lind AE, Lazar CS, Hinrichs K-U, Teske AP, *et al.* (2016). Genomic

inference of the metabolism of cosmopolitan subsurface Archaea, Hadesarchaea. *Nat*

Microbiol **1**: 16002.

Bell CW, Fricks BE, Rocca JD, Steinweg JM, McMahon SK, Wallenstein MD. (2013).

High-throughput fluorometric measurement of potential soil extracellular enzyme

activities. *J Vis Exp* e50961.

Bendtsen JD, Kiemer L, Fausbøll A, Brunak S. (2005). Non-classical protein secretion in

bacteria. *BMC Microbiol* **5**: 58.

Benner R, Amon RMW. (2015). The size-reactivity continuum of major bioelements in

the ocean. *Ann Rev Mar Sci* **7**: 185–205.

Benz R, Bauer K. (1988). Permeation of hydrophilic molecules through the outer

membrane of gram-negative bacteria. Review on bacterial porins. *Eur J Biochem* **176**:

1–19.

Biddle JF, Lipp JS, Lever MA, Lloyd KG, Sørensen KB, Anderson R, *et al.* (2006).

Heterotrophic Archaea dominate sedimentary subsurface ecosystems off Peru. *Proc*

Natl Acad Sci U S A **103**: 3846–51.

Blair CC, D'Hondt S, Spivack AJ, Kingsley RH. (2007). Radiolytic hydrogen and

microbial respiration in subsurface sediments. *Astrobiology* **7**: 951–70.

Boudreau BP. (1996). A method-of-lines code for carbon and nutrient diagenesis in

aquatic sediments. *Comput Geosci* **22**: 479–496.

Burdige DJ. (2007). Preservation of organic matter in marine sediments: controls,

mechanisms, and an imbalance in sediment organic carbon budgets? *Chem Rev* **107**:

467–85.

Carter DO, Yellowlees D, Tibbett M. (2007). Autoclaving kills soil microbes yet soil

enzymes remain active. *Pedobiologia (Jena)* **51**: 295–299.

Coolen MJL, Cypionka H, Sass AM, Sass H, Overmann J. (2002). Ongoing modification

of Mediterranean Pleistocene sapropels mediated by prokaryotes. *Science* **296**: 2407–

10.

Coolen MJL, Overmann J. (2000). Functional Exoenzymes as Indicators of Metabolically

Active Bacteria in 124,000-Year-Old Sapropel Layers of the Eastern Mediterranean

Sea. *Appl Environ Microbiol* **66**: 2589–2598.

Cornish-Bowden A. (2012). Fundamentals of enzyme kinetics. Fourth Edition. Wiley-

VCH Verlag & Co. KGaA: Weinheim.

Dauwe B, Middelburg JJ. (1998). Amino acids and hexosamines as indicators of organic

matter degradation state in North Sea sediments. *Limnol Oceanogr* **43**: 782–798.

Dauwe B, Middelburg JJ, Van Rijswijk P, Sinke J, Herman PMJ, Heip CHR. (1999).

Enzymatically hydrolyzable amino acids in North Sea sediments and their possible

implication for sediment nutritional values. *J Mar Res* **57**: 109–134.

Gruebel KA, Martens CS. (1984). Radon-222 tracing of sediment-water chemical

transport in an estuarine sediment. *Limnol Oceanogr* **29**: 587–597.

Hare PE. (1968). Geochemistry of Proteins, Peptides, and Amino Acids. In: Eglinton G,

Murphy MTJ (eds). *Organic Geochemistry: Methods and Results*. Springer-Verlag:

Berlin, Heidelberg, pp 438–462.

Hiller K, Grote A, Scheer M, Münch R, Jahn D. (2004). PrediSi: Prediction of signal

peptides and their cleavage positions. *Nucleic Acids Res* **32**. e-pub ahead of print, doi:

10.1093/nar/gkh378.

Jacobson Meyers ME, Sylvan JB, Edwards KJ. (2014). Extracellular enzyme activity and

microbial diversity measured on seafloor exposed basalts from Loihi seamount

indicate the importance of basalts to global biogeochemical cycling. *Appl Environ*

Microbiol **80**: 4854–64.

Jørgensen BB. (1978). A comparison of methods for the quantification of bacterial sulfate

reduction in coastal marine sediments: II. Calculation from mathematical models.

Geomicrobiol J **1**: 29–47.

Jørgensen BB, Marshall IPG. (2016). Slow Microbial Life in the Seabed. *Ann Rev Mar*

Sci **8**: 311–32.

Kaiser K, Benner R. (2008). Major bacterial contribution to the ocean reservoir of detrital

organic carbon and nitrogen. *Limnol Oceanogr* **53**: 99–112.

Kallmeyer J, Pockalny R, Adhikari RR, Smith DC, D'Hondt S. (2012). Global

distribution of microbial abundance and biomass in subseafloor sediment. *Proc Natl*

Acad Sci U S A **109**: 16213–6.

Kelley CA, Martens CS, Chanton JP. (1990). Variations in sedimentary carbon

remineralization rates in the White Oak River estuary, North Carolina. *Limnol*

Oceanogr **35**: 372–383.

Kubo K, Lloyd KG, F Biddle J, Amann R, Teske A, Knittel K. (2012). Archaea of the

Miscellaneous Crenarchaeotal Group are abundant, diverse and widespread in marine

sediments. *ISME J* **6**: 1949–65.

Lapham LL, Chanton JP, Martens CS, Sleeper K, Woolsey JR. (2008). Microbial activity

in surficial sediments overlying acoustic wipeout zones at a Gulf of Mexico cold

seep. *Geochemistry, Geophys Geosystems* **9**: Q06001.

Lazar CS, Baker BJ, Seitz K, Hyde AS, Dick GJ, Hinrichs K-U, *et al.* (2016). Genomic

evidence for distinct carbon substrate preferences and ecological niches of

Bathyarchaeota in estuarine sediments. *Environ Microbiol* **18**: 1200–11.

Lee C, Cronin C. (1984). Particulate amino acids in the sea: Effects of primary

productivity and biological decomposition. *J Mar Res* **42**: 1075–1097.

Lloyd KG, Alperin MJ, Teske A. (2011). Environmental evidence for net methane

production and oxidation in putative ANaerobic MEthanotrophic (ANME) archaea.

Environ Microbiol **13**: 2548–2564.

Lloyd KG, May MK, Kevorkian RT, Steen AD. (2013a). Meta-analysis of quantification

methods shows that archaea and bacteria have similar abundances in the subseafloor.

Appl Environ Microbiol **79**: 7790–9.

Lloyd KG, Schreiber L, Petersen DG, Kjeldsen KU, Lever M a, Steen AD, *et al.* (2013b).

Predominant archaea in marine sediments degrade detrital proteins. *Nature* **496**: 215–

8.

Lomstein B, Jorgensen B, Schubert C, Niggemann J. (2006). Amino acid biogeo- and

stereochemistry in coastal Chilean sediments. *Geochim Cosmochim Acta* **70**: 2970–

2989.

Martens CS, Goldhaber MB. (1978). Early diagenesis in transitional sedimentary

environments of the White Oak River Estuary, North Carolina. *Limnol Oceanogr* **23**:

428–441.

Meng J, Xu J, Qin D, He Y, Xiao X, Wang F. (2014). Genetic and functional properties

of uncultivated MCG archaea assessed by metagenome and gene expression analyses.

ISME J **8**: 650–9.

Michalska K, Steen AD, Chhor G, Endres M, Webber AT, Bird J, *et al.* (2015). New

aminopeptidase from 'microbial dark matter' archaeon. *FASEB J* **29**: 4071–4079.

Obayashi Y, Suzuki S. (2008). Occurrence of exo- and endopeptidases in dissolved and

particulate fractions of coastal seawater. *Aquat Microb Ecol* **50**: 231–237.

Obayashi Y, Suzuki S. (2005). Proteolytic enzymes in coastal surface seawater:

Significant activity of endopeptidases and exopeptidases. *Limnol Oceanogr* **50**: 722–

726.

Pedersen A-GU, Thomsen TR, Lomstein BA, Jørgensen NOG. (2001). Bacterial

influence on amino acid enantiomerization in a coastal marine sediment. *Limnol*

Oceanogr **46**: 1358–1369.

Petersen TN, Brunak S, von Heijne G, Nielsen H. (2011). SignalP 4.0: discriminating

signal peptides from transmembrane regions. *Nat Methods* **8**: 785–6.

Rawlings ND, Barrett AJ. (1999). MEROPS: The peptidase database. *Nucleic Acids Res*

27: 325–331.

Schimel JP, Weintraub MN. (2003). The implications of exoenzyme activity on microbial

carbon and nitrogen limitation in soil: a theoretical model. *Soil Biol Biochem* **35**:

549–563.

Schmidt J. (2016). Microbial Extracellular Enzymes in Marine Sediments: Methods

Development and Potential Activities in the Baltic Sea Deep Biosphere. University of

Tennessee - Knoxville.

Seitz KW, Lazar CS, Hinrichs K-U, Teske AP, Baker BJ. (2016). Genomic

reconstruction of a novel, deeply branched sediment archaeal phylum with pathways

for acetogenesis and sulfur reduction. *ISME J* **10**: 1696–1705.

Sinsabaugh RL, Belnap J, Findlay SG, Shah JJF, Hill BH, Kuehn KA, *et al.* (2014).

Extracellular enzyme kinetics scale with resource availability. *Biogeochemistry* **121**:

287–304.

Spang A, Saw JH, Jørgensen SL, Zaremba-Niedzwiedzka K, Martijn J, Lind AE, *et al.*

(2015). Complex archaea that bridge the gap between prokaryotes and eukaryotes.

Nature **521**: 173–179.

Steen AD, Arnosti C. (2013). Extracellular peptidase and carbohydrate hydrolase

activities in an Arctic fjord (Smeerenburgfjord, Svalbard. *Aquat Microb Ecol* **69**: 93–

99.

Steen AD, Arnosti C. (2011). Long lifetimes of β-glucosidase, leucine aminopeptidase,

and phosphatase in Arctic seawater. *Mar Chem* **123**: 127–132.

Steen AD, Jørgensen BB, Lomstein BA. (2013). Abiotic Racemization Kinetics of

Amino Acids in Marine Sediments. *PLoS One* **8**. e-pub ahead of print, doi:

10.1371/journal.pone.0071648.

Steen AD, Vazin JP, Hagen SM, Mulligan KH, Wilhelm SW. (2015). Substrate

specificity of aquatic extracellular peptidases assessed by competitive inhibition

assays using synthetic substrates. *Aquat Microb Ecol* **75**: 271–281.

Steen AD, Ziervogel K. (2012). Comment on the review by German et

al.(2011)'Optimization of hydrolytic and oxidative enzyme methods for ecosystem

studies'[Soil Biology & Biochemistry 43: 1387--1397]. *Soil Biol Biochem* **48**: 196–

197.

Stursova M, Sinsabaugh RL. (2008). Stabilization of oxidative enzymes in desert soil

may limit organic matter accumulation. *Soil Biol Biochem* **40**: 550–553.

Vetter YA, Deming JW. (1994). Extracellular enzyme activity in the Arctic Northeast

Water polynya. *Mar Ecol Prog Ser* **114**: 23–34.

Vetter Y, Deming J, Jumars P, Krieger-Brockett B. (1998). A Predictive Model of

Bacterial Foraging by Means of Freely Released Extracellular Enzymes. *Microb Ecol*

36: 75–92.

Yu NY, Wagner JR, Laird MR, Melli G, Rey S, Lo R, *et al.* (2010). PSORTb 3.0:

improved protein subcellular localization prediction with refined localization

subcategories and predictive capabilities for all prokaryotes. *Bioinformatics* **26**:

1608–1615.

Zaremba-Niedzwiedzka K, Caceres E, Saw J, Backstrom D, Juzokaite L, Vancaester E, *et*

al. Metagenomic exploration of Asgard arcahea illuminates the origin of eukaryotic

cellular complexity.

Figure Legends

Fig 1: (A) Frequency of genes for various classes of extracellular peptidases, relative to

all genes for extracellular peptidases; (b) sources of extracellular peptidases by domain

and depth; and sources of (c) Archaeal and (d) Bacterial peptidases by phylum and depth.

Fig 2: Peptidase saturation curves, showing Michealis-Menten kinetics consistent with

enzymatic rather than abiotic substrate hydrolysis. Open triangles indicate autoclaved

controls.

Fig 3: Peptidase activities compared to microbial abundance and activity. Panel A:

summed V_{max} of the six peptidases measured in 2013, expressed relative to the value at

1.5 cm. Error bars represent standard error of the estimate of the rate of fluorophore

production. B: Summed V_{max} relative to cell abundance. Error bars represent error

propagated from error of Vmax and standard deviation of cell counts. C: Modeled

organic carbon oxidation rate. D: Summed V_{max} relative to organic carbon oxidation rate.

Error bars represent standard error of the estimate of the rate of fluorophore production.

Fig 4: Estimated K_m values as a function of sediment depth. Colored lines indicate linear

regressions for individual substrates, while the black line and shaded area represent a

regression for all substrates taken together.

655
656

657   Fig 5: Left panel: Ratio of D-phenylalanyl aminopeptidase V_{max} to L-phenylalanyl

658   aminopeptidase V_{max} versus depth. Right panel: Ratio of ornithyl aminopeptidase V_{max} to

659   L-phenylalanyl aminopeptidase V_{max}. The shaded band indicates the 95% confidence

660   interval of the fitted values.