1 **Metabolic roles of uncultivated bacterioplankton lineages in the northern Gulf of Mexico**
2 **"Dead Zone"**
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28 Running title: Decoding microbes of the Dead Zone
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Abstract
Marine regions that have seasonal to long-term low dissolved oxygen (DO) concentrations,
sometimes called 'dead zones,' are increasing in number and severity around the globe with
deleterious effects on ecology and economics. One of the largest of these coastal dead zones
occurs on the continental shelf of the northern Gulf of Mexico (nGOM), which results from
eutrophication-enhanced bacterioplankton respiration and strong seasonal stratification. Previous
research in this dead zone revealed the presence of multiple cosmopolitan bacterioplankton
lineages that have eluded cultivation, and thus their metabolic roles in this ecosystem remain
unknown. We used a coupled shotgun metagenomic and metatranscriptomic approach to
determine the metabolic potential of Marine Group II Euryarchaeota, SAR406, and SAR202. We
recovered multiple high-quality, nearly complete genomes from all three groups as well as those
belonging to Candidate Phyla usually associated with anoxic environments- Parcubacteria (OD1)
and Peregrinibacteria. Two additional groups with putative assignments to ACD39 and
PAUC34f supplement the metabolic contributions by uncultivated taxa. Our results indicate
active metabolism in all groups, including prevalent aerobic respiration, with concurrent
expression of genes for nitrate reduction in SAR406 and SAR202, and dissimilatory nitrite
reduction to ammonia and sulfur reduction by SAR406. We also report a variety of active
heterotrophic carbon processing mechanisms, including degradation of complex carbohydrate
compounds by SAR406, SAR202, ACD39, and PAUC34f. Together, these data help constrain
the metabolic contributions from uncultivated groups in the nGOM during periods of low DO
and suggest roles for these organisms in the breakdown of complex organic matter.
53

Importance
Dead zones receive their name primarily from the reduction of eukaryotic macrobiota (demersal
fish, shrimp, etc.) that are also key coastal fisheries. Excess nutrients contributed from
anthropogenic activity such as fertilizer runoff result in algal blooms and therefore ample new
carbon for aerobic microbial metabolism. Combined with strong stratification, microbial
respiration reduces oxygen in shelf bottom waters to levels unfit for many animals (termed
hypoxia). The nGOM shelf remains one of the largest eutrophication-driven hypoxic zones in the
world, yet despite its potential as a model study system, the microbial metabolisms underlying
and resulting from this phenomenon—many of which occur in bacterioplankton from poorly
understood lineages—have received only preliminary study. Our work details the metabolic
potential and gene expression activity for uncultivated lineages across several low DO sites in
the nGOM, improving our understanding of the active biogeochemical cycling mediated by these
"microbial dark matter" taxa during hypoxia.
67

Introduction

Hypoxia (dissolved oxygen [DO] below 2 mg·L^{-1}/~62.5 μmol·kg^{-1}) is dangerous or lethal to a wide variety of marine life, including organisms of economic importance (1). Hypoxia results from oxygen consumption by aerobic microbes combined with strong stratification that prevents reoxygeneation of bottom waters. These taxa are fueled primarily by autochthonous organic matter generated from phytoplankton responding to nitrogen input (1). Hypoxic zones have become more widespread globally through the proliferation of nitrogen-based fertilizers and the resulting increases in transport to coastal oceans via runoff (2). In the nGOM, nitrogen runoff from the Mississippi and Atchafalaya Rivers leads to bottom water hypoxia that can extend over 20,000 km^2- one of the world's largest seasonal "dead zones" (1). Action plans to mitigate nGOM hypoxia have stressed that increasing our "understanding of nutrient cycling and transformations" remains vital for plan implementation (3). These needs motivated our current study of the engines of hypoxic zone nutrient transformation: microorganisms.

Much of our current knowledge regarding microbial contributions to regions of low DO comes from numerous studies investigating naturally occurring, deep-water oxygen minimum zones (OMZs), such as those in the Eastern Tropical North and South Pacific, the Saanich Inlet, and the Arabian, Baltic, and Black Seas (4-11). In many of these systems, continual nutrient supply generates permanent or semi-permanent decreases in oxygen, sometimes to the point of complete anoxia (4). During these conditions, anaerobic metabolisms, such as nitrate and sulfate reduction and anaerobic ammonia oxidation, become prevalent (5, 9, 11-13). In contrast, nGOM hypoxia is distinguished by a seasonal pattern of formation, persistence, and dissolution (1); benthic contributions to bottom water oxygen consumption (14, 15); and a shallow shelf that places much of the water column within the euphotic zone (16). While parts of the nGOM hypoxic zone can become anoxic (1, 17), many areas maintain low oxygen concentrations even during peak hypoxia while the upper water column remains oxygenated (18-20).

The first studies of bacterioplankton assemblages during nGOM hypoxia showed nitrifying Thaumarchaea dominated (21) and could be highly active (22), suggesting a major role for these taxa in nGOM nitrogen cycling. However, many more poorly understood organisms from cosmopolitan, but still uncultivated "microbial dark matter" (23) lineages, such as Marine Group II Euryarchaeota (MGII), SAR406, and SAR202, also occurred in abundance (21, 22). While the likely functions of some of these groups have become clearer recently, all of them contain multiple sublineages that may have distinct metabolic roles. For example, the SAR202 lineage of Chloroflexi contains at least five subclades with distinct ecological profiles (24, 25), and the best understood examples have been examined in the context of complex carbon degradation in the deep ocean (25). Likewise, SAR406 represents a distinct phylum with numerous sublineages, and the bulk of metabolic inference comes from taxa in deep water OMZs (23, 26-28).

None of these groups have been studied in detail in shallow coastal waters, particularly in the context of seasonal hypoxia. Thus, we pursued a combined metagenomic/metatranscriptomic approach to i) elucidate the specific contributions of these uncultivated lineages to

biogeochemical cycling in the nGOM during hypoxia, ii) evaluate the relative similarity of these
organisms to their counterparts elsewhere, and iii) determine if other uncultivated lineages had
eluded previous microbial characterization in the region due to confounding factors such as
primer bias (29), 16S rRNA gene introns (30), or low abundance. Metagenomic binning
recovered 20 genomes across seven uncultivated lineages including MGII, SAR406, and
SAR202, and also from Candidate Phyla previously uncharacterized in the nGOM: Parcubacteria
(23), Peregrinibacteria (31), and possibly PAUC34f (32) and ACD39 (33). Our results provide
the first information on the likely potential function and activity of these taxa during hypoxia in
the shallow nGOM and suggest novel roles for some of these groups that possibly reflect
sublineage-specific adaptations.
118
119   **Results**
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Study area
Our previous work used 16S rRNA gene amplicon data and qPCR to examine correlations
between whole microbial communities, nutrients, and DO across the geographic range of the
2013 seasonal hypoxia (21). Here we selected six of those samples from offshore of the region
between Atchafalaya Bay and Terrebonne Bay (D', D, and E transects). These sites ranged
considerably in DO concentration ($\sim$2.2 – 132 μmol$\cdot$kg^{-1}), and we chose them to facilitate a
detailed investigation of the metabolic repertoire of individual taxa across the span of suboxic (1-
20 μmol$\cdot$kg^{-1} DO) to oxic (> 90 μmol$\cdot$kg^{-1} DO) (5) water. Microbial samples from these sites
were collected at the oxygen minimum near the bottom. Site depth ranged from 8 – 30 m, with
the hypoxic (< 2 mg$\cdot$L^{-1}/62.5 μmol$\cdot$kg^{-1}) layer (at sites D2, D3, E2A, and E4) extending up to
$\sim$5 m off the bottom (Table S1).
132
Metagenomic assembly yielded high quality genomes from multiple uncultivated lineages
Our initial assembly and binning efforts recovered 76 genomes. Using a concatenated ribosomal
protein tree that included members of the Candidate Phylum Radiation (CPR) (34) (Fig. S1),
CheckM (35) (Fig. S2), 16S rRNA genes and other single copy markers where available, and
analyses of individual gene taxonomy (Fig. S3), we assigned 20 genomes to uncultivated
"microbial dark matter" groups. These were six Marine Group II Euryarchaeota (MGII), five
Marinimicrobia (SAR406), three in the SAR202 clade of Chloroflexi, and within Candidate
Phyla (CP), one Parcubacteria (OD1), two Peregrinibacteria, and putatively, one ACD39 and two
PAUC34f (Table 1, Supplemental Information). We further defined the MGII, SAR406, and
SAR202 genomes into sublineages based on average amino acid identity (AAI), GC content,
clade structure in the ribosomal protein tree, and 16S rRNA genes (Supplemental Information).
SAR406 genomes belonged to two groups, A and B, corresponding to the previously established
Arctic96B-7 and SHBH1141 16S rRNA gene clades (27). The three SAR202 genomes belonged
to the previously established subclade I 16S rRNA gene clade (24). All genomes, with the
exception of the Parcubacteria Bin 40, had estimated contamination of less than 6%, and in the

majority of cases, less than 2%. Four of the six MGII genomes had estimated completeness (via
CheckM) of greater than 61%, four of the five SAR406 greater than 73%, and all three SAR202
genomes were estimated to be greater than 83% complete. All CP lineages had at least one
genome estimated to be greater than 71% complete (Table 1).
152
Unique roles for the ubiquitous MGII, SAR406, and SAR202 lineages in nGOM hypoxia
MGII comprised over 10% of the total community in some samples from 2013, and one MGII
OTU also had a strong negative correlation with DO during 2013 hypoxia (21). Within our
metagenomics dataset, MGII were more abundant in lower oxygen samples than in fully oxic
ones, and the most abundant of the lineages reported here (Fig. S7). The majority encoded for
aerobic, chemoheterotrophic metabolism, with no predicted genes for nitrogen or sulfur
respiration except for a putative nitrite reductase (*nirK*) in a single genome- Bin 15 (Fig.1, Table
S1). MGII genomic abundance correlated well with transcriptional abundance in most samples
(Fig. 3), and we specifically found MGII cytochrome c oxidase expression throughout, though
the levels and patterns differed depending on the gene and the source genome (Fig. 4, Table S1).
Expression of the *nirK* gene occurred in the D2 and E2A samples- both suboxic. All but the most
incomplete genome encoded for ammonia assimilation, making this a likely nitrogen source.
Aggregate metabolic construction from multiple bins also indicated a complete TCA cycle,
glycolysis via the pentose phosphate pathway, and gluconeogenesis (Fig. 1). Carbohydrate active
enzyme (CAZy) genes can provide critical information on the relationships between microbes
and possible carbon sources (36). We found few and these were largely restricted to
glycosyltransferases (GT) in families 2 and 4, with activities related to cellular synthesis. In
general, CAZy expression occurred for at least one gene in every genome and we detected
expression of GT cellular synthesis genes in the E2A sample (Fig. 5), likely indicating actively
growing cells.
SAR406 represented over 5% of the population in some locations during hypoxia in
2013, and one abundant OTU was negatively correlated with DO (21). Metagenomic read
recruitment to the SAR406 bins confirmed this trend, with greater recruitment in the suboxic
samples relative to dysoxic or oxic (Fig. S7). Total RNA recruitment was strongest to Bins 45
and 51-1, though most bins showed a RNA to DNA recruitment ratio > 1 in at least one sample,
indicating these taxa were likely active (Fig. 3). Despite their affinity for low oxygen
environments, the SAR406 genomes encoded a predicted capacity for aerobic respiration (Fig.
1), and we found expression of cytochrome c oxidases in even the lowest oxygen samples (Fig.
4). The Group B genomes encoded both high and low-affinity cytochrome c oxidases (37),
whereas the high affinity (cbb$_3$-type) oxidases were not recovered in the Group A genomes
(Table S1), which may indicate sublineage-specific optimization for different oxygen regimes.
Sublineage variation also appeared in genes for the nitrogen and sulfur cycles. Group B
genomes all contained predicted nitrous oxide reductases (*nosZ*) and *nrfAH* genes for
dissimilatory nitrite reduction to ammonium (defined here as DNRA, although this acronym
frequently refers to nitrate, even though that is a misnomer (38)). The *nrfA* genes formed a

monophyletic group with *Anaeromyxobacter dehalogenans* 2CP-1, an organism with
demonstrated DNRA activity (38) (Fig. S8A). The genes also contained conserved motifs
diagnostic of the *nrfA* gene (38) (Fig. S8B, C). We observed expression of *nrfAH* and *nosZ* at the
sites with the lowest DO concentrations (D2, E2A, and E4) and expression appeared to have a
negative relationship with DO concentration (Fig. 4). The Bin 51-1 Group A genome contained
predicted *narHI* genes for dissimilatory nitrate reduction, which we did not find in the Group B
genomes. We observed expression of SAR406 *narHI* only in the lowest DO sample from station
E2A (Fig. 4). Two Group B SAR406 genomes had predicted *phsA* genes for thiosulfate reduction
to sulfide (and/or polysulfide reduction (39)), as previously described from fosmid sequences
(27). We detected transcripts for these genes only in samples E2A and E4, the two lowest DO
samples (Fig. 4). Many of the anaerobic respiratory genes were co-expressed with cytochrome c
oxidases, indicating a potential for either co-reduction of these alternative terminal electron
acceptors or poising of these organisms for rapid switching between aerobic and anaerobic
metabolism (40).
All SAR406 genomes had numerous genes for heterotrophy. We found CAZy genes in
all major categories except polysaccharide lyases, and expression for most of these genes in both
Group A and Group B genomes in one or more samples (Fig. 5). Notable carbohydrate
compounds for which degradation capacity was predicted include cellulose (glycoside hydrolase
(GH) families GH3, GH5; carbohydrate binding module (CBM) family CBM6), starch (GH13),
agar and other sulfated galactans (GH2, GH16), chitin (GH18), xylan (GH30, CBM9), and
peptidoglycan (GH23, GH103, CBM50). The genomes contained putative transporters for a
variety of dissolved organic matter (DOM) components including nucleosides, amino and fatty
acids, and oligopeptides (Table S1). We also found numerous outer membrane transporters,
including cation symporters; Outer Membrane Receptors (OMR- TonB-dependent), which play
important roles in transport of metals, vitamins, colicins, and other compounds; Outer Membrane
Factors (OMF); and most genomes also had large numbers of duplicated genes (24 in Bin 45-2),
identified via hidden Markov model searches against the SFam database (41), annotated as "Por
secretion system C-terminal sorting domain-containing protein," some of which were associated
with GH16. These genes likely play a role in sorting C-terminal tags of proteins targeted for
secretion via the Por system, which is essential for gliding motility and chitinase secretion in
some Bacteroidetes (42). The extensive gene duplication may indicate expanded and/or
specialized sorting functionality, and suggests an emphasis on protein secretion in this group.
Expression of a membrane-bound lytic murein transglycosylase D (GH23) involved in
membrane remodeling also supports the idea of active and growing cells from Group A in all
samples (Table S1).
We detected Chloroflexi 16S rRNA gene sequences during 2013 hypoxia at up to 5% of
the community (21), and recovered three mostly complete SAR202 Chloroflexi genomes in this
work. Although present at lower abundance than MGII and SAR406 (Fig. S7), these genomes
showed relatively high activity in some samples (Fig. 3). Like subclade III and V, subclade I
organisms likely respire oxygen. However, we also found *napAB* and *nosZ* genes for nitrate and

nitrous oxide reduction, respectively (Fig. 1). As in SAR406, we detected concurrent expression
of these genes with cytochrome c oxidases in the lowest DO samples (Fig. 4) (Fig. 4, Table S1).
The SAR202 genomes have numerous transporters, many with predicted roles in organic
matter transport, which supports previous observations of DOM uptake (43). In particular,
SAR202 genomes had considerably more Major Facilitator Superfamily (MFS) transporters than
the other genomes in this study (Table S1) and those of the subclade III genomes (25), and SFam
searches revealed a the majority of these shared annotation as a "Predicted arabinose efflux
permease" (SFam 346742). MFS genes transport numerous diverse substrates, such as sugars and
amino acids, through coupling with an ion gradient, and can be associated with either uptake or
export of compounds (44). SAR202 genomes also had between 53 and 66 predicted ABC
transporters.
The SA202 genomes encoded a number of duplicated genes in specific gene families.
The largest gene family expansion that we observed was associated with SFam 6706, with
between 46 and 48 genes in this family encoded in each genome. Most of these (121/142) were
annotated as either a "galactonate dehydratase" or a "L-alanine-DL-glutamate epimerase."
Galactonate dehydratase catalyzes the first step of the pathway to utilize D-galactonate in central
carbon metabolism via the pentose phosphate pathway. The large number of genes in these
categories likely indicates some divergence for alternative roles as this group belongs broadly to
the COG4948 "L-alanine-DL-glutamate epimerase or related enzyme of enolase superfamily."
All genomes also had numerous dehydrogenases as reported for the subclade III genomes (25).
Specifically, SFams 346640 and 1639 were the third and fourth most abundant, with 16-18 and
13-15 genes in each family, respectively, across the three genomes. Genes in these families were
annotated as "short-chain alcohol dehydrogenase family," "3-alpha (or 20-beta)-hydroxysteroid
dehydrogenase," "meso-butanediol dehydrogenase," and others. These match the annotations of
the subclade III genomes, and suggest a similar role in conversion of alcohols to ketones (25).
The SAR202 genomes have comparatively few CAZy genes relative to the other genomes. GH15
and GH63 suggest starch degradation, and GH105 pectin degradation, and we detected
expression of multiple genes in these categories across samples (Fig. 5, Table S1).
256
Other Candidate Phylum organisms in nGOM hypoxia
In contrast to the abundant and cosmopolitan MGII, SAR406, and SAR202 clades, we also
recovered genomes from several groups that were either previously undetected in the nGOM or
very rare. Although these taxa likely do not contribute the biomass of more populous clades,
their genomes provide important insight into their functional potential during hypoxia. The Bin
13 genome (possibly ACD39) also had the highest relative activity compared to all the other
genomes in our study (Fig. 3), underlining the point that low abundance does not automatically
equate to low metabolic impact. Bin 13 had predicted aerobic respiration with both high and low-
affinity cytochrome c oxidases (Fig. 2). The low affinity oxidases contributed more reads in the
samples where we could detect expression (Table S1). The genome contained numerous
predicted CAZy genes in the glycosyltransferase and glycoside hydrolase categories, spread

across multiple families in each (Table S1). Notable degradation capacity included starch (GH13) and peptidoglycan (GH23, GH103, GH104).

Bin 13 had ~ 80 ABC transporter genes, and similarly to the SAR406 genomes, numerous outer membrane transporters, including the OMR and OMF families. We predict complete glycolysis/gluconeogenesis pathways and a TCA cycle. We recovered paralogous pilus subunit genes, chemotaxis genes, and a partial flagellar assembly. Furthermore we detected relatively high expression of the *flgLN* flagellin genes in samples D2, D3, E2A, and E4 (Table S1) suggesting active motility in these environments. Several other Bin 13 genes were among the most highly expressed in all samples, but could only be classified as hypothetical (Table S1). Similarly, the three most populous SFams in Bin 13, according to number of genes (n=16, 15, and 13) also linked to genes annotated as hypothetical proteins with either tetratricopeptide, HEAT, TPR, or Sel1 repeats. Although currently obscure, these and the highly expressed hypothetical genes represent important targets for future research into the function of this group.

Bins 50 and 48 were lower in abundance than SAR202 genomes (Fig. S7, PAUC34f), with no observable trend associated with oxygen levels (Fig. S7). These genomes encoded flagellar motility, aerobic respiration, glycolysis via the pentose-phosphate pathway, gluconeogenesis, assimilatory sulfate reduction, and DNRA (Fig. 2). The *nrfA* subunit from both genomes grouped in the same monophyletic clade as those from SAR406 (Fig. S8A), and had similar conserved motifs (Fig. S8B, C). However, we note that the *nrfAH* gene sets for Bins 50 and 48 occurred on relatively short contigs (5650 and 5890 bp, respectively), so the metabolic assignment cannot be corroborated as definitively as that for SAR406. The Bin 50 genome was among the more active in our analysis (Fig. 3), and we detected highest expression of cytochrome c oxidase components in samples E2A and E4 (Fig. 4). DNRA gene expression was low but observable in the same samples. We also recovered a partial gene for the ribulose-bisphosphate carboxylase (RuBisCO) large subunit, but this fragment was on a very short contig (3954 bp), and we did not detect expression in any of our samples, so we cannot rule out that this gene occurred on a contaminating contig.

The Bin 50 and 48 genomes had abundant CAZy genes in all categories, suggesting a highly flexible metabolic repertoire for carbon acquisition. They contain possible capacity for breakdown of starch (GH13, CBM48), peptidoglycan (GH23, CBM50), fructose-based oligosaccharides (GH32), and hemicellulose (GH2, GH3, GH43). Notably these genomes were the only ones with predicted polysaccharide lyases (PL) among those compared (with the exception of a single predicted PL gene in SAR406- Table S1). PL genes cleave uronic-acid containing polysaccharides (45). These organisms seem particularly adapted for pectin (PL1, PL2, PL9, PL10, PL11, PL22, GH78) and alginate (PL15, PL17) degradation- both compounds are common cell wall components of green and brown algae, respectively.

In line with the algal cell wall degradation ability, we detected a large expansion (102 genes in Bin 50) of sulfatase genes in SFam 1534, annotated predominantly as either "arylsulfatase A" or "choline-sulfatase." Arylsulfatases cleave sulfate esters, usually to supply microbes with a source of sulfur, and can be located intracellularly or in membranes (46).

Choline-sulfatases cleave choline sulfate to choline and sulfate, with downstream use for the
former as a carbon source or osmoprotectant and the latter as a sulfur source (47). Given the
predicted assimilatory sulfate reduction pathway in Bins 50 and 48, this is a logical means to
obtain sulfur for the group. We observed large expansions in galactonate and other dehydratases
(as in SAR202, above- SFam 6706 n=42 in Bin 50), as well as numerous ABC transporter
permeases (SFam 4442), which match the transporter predictions via IMG: 117 predicted genes
for ABC transporters in all. These genomes also had numerous OMF and OMR transporter genes
(Table S1). The large number of transporters and protein family expansions correspond to the
relatively large expected genome sizes (between 5 and 6 Mbp).
We also recovered genomes associated with CPR taxa usually associated with anoxic
environments: two Peregrinibacteria and one from the Uhrbacteria subclade of the Parcubacteria
(formerly OD1). All three genomes could be assigned taxonomically with high confidence based
on their positions in the ribosomal protein tree (Fig. S1) and via gene annotations (Fig. S3). We
note that although the Peregrinibacteria bins (16 and 39) had very low predicted contamination,
the Parcubacteria Bin 40 has 15% predicted contamination (75% of which we attribute to strain
heterogeneity in the bin) (Table 1). Recovery of Parcubacteria from a coastal marine system is
unusual, but not unprecedented. Parcubacteria single-cell genomes have been identified in
marine and brackish sources (23), and we previously identified 26 rare OTUs assigned to the
phylum in nGOM hypoxia (21). That number of OTUs may explain why we observed 20 single
copy marker genes present in two copies in Bin 40 (Table S1).
In contrast to Parcubacteria, Peregrinibacteria have thus far only been found in terrestrial
subsurface aquifers (31, 33, 48, 49), and remained undetected in our amplicon survey (21). Both
groups occurred in low relative abundance to the other taxa in this study, and showed the lowest
activity (Fig. 3). Consistent with previous reports of obligate fermentative metabolism by
Parcubacteria and Peregrinibacteria (23, 30, 31, 48), we identified no respiratory pathways for
these taxa (Fig. 2) and they trended towards greater abundances in the lowest DO samples (Fig.
S7). In spite of relatively high predicted genome completion, we found very few CAZy genes,
and those were mostly restricted to glycosyltransferases (Table S1) probably involved in capsular
polysaccharide synthesis. While these organisms had low relative abundance to the other groups
(Fig. S7), we did observe activity in some samples (Fig. 3- E2, E2A, D1).
338
Discussion
This work provides the first reconstruction of multiple nearly complete genomes from
uncultivated bacterioplankton during nGOM hypoxia. Although we define roles for MGII,
SAR406, SAR202, Bin 13 and Bins 50/48 as aerobic heterotrophs, we also observed concurrent
expression of genes associated with anaerobic metabolism in SAR406 (nitrate reduction, DNRA,
nitrous oxide reduction, and sulfur reduction), SAR202 (nitrate and nitrous oxide reduction),
MGII (nitrite reduction), and Bins 50/48 (DNRA) in suboxic samples with the lowest measured
DO concentrations. Simultaneous utilization of multiple electron acceptors with different redox
potentials likely indicates an abundant supply of electron donors (50), may denote niche

partitioning within group sublineages at a finer level of taxonomic resolution than we observed,
or indicate poising of taxa for rapidly changing chemical gradients (40). An organism's set of
CAZy genes often gives insights into its biology, in particular into nutrient sensing and
acquisition. All taxa examined in this study had predicted chemoorganoheterotrophic
metabolism, and the CAZy genes found in these genomes suggest that SAR406, SAR202, Bin
13, and Bins 50/48 participate in the degradation of complex organic matter resulting from the
detritus of larger organisms. This matches the general model of hypoxic zone oxygen
consumption resulting from sinking organic matter provided by algal blooms in surface waters
(1). The observed activity of obligate fermentative groups Parcubacteria and Peregrinibacteria
also suggests that anoxic pockets occur in the water column where these organisms can thrive.
Marine group II (MGII) is a broadly distributed archaeal clade, with members found in
different marine (51, 52), and sedimentary (53), environments. Previous work during 2012 and
2013 hypoxia indicated a proliferation of archaeal taxa in both the Thaumarchaea and MGII
phyla (21, 22). The prevalence of MGII among lower oxygen samples in the hypoxic zone is
somewhat surprising, considering that they are commonly associated with aerobic environments
(52). However, oxygen was still present in even the lowest DO samples (Fig. 3), and MGII
success likely had more to do with the carbon content than oxygen levels. These nGOM MGII
appear to be metabolically similar to those described in previous work: MGII have been shown
to be dominant in water column environments associated with blooms in productivity, for
example at deep-sea hydrothermal plumes (51). Thus, the increased availability of organic matter
(proteins and carbohydrates), thought to be preferred substrates for MGII (54, 55), probably
explains their abundance.
Another cosmopolitan group found in our samples was SAR406 or Marine Group A.
These organisms were discovered over 20 years ago (28, 56), and the clade has recently been
proposed as the phylum "Marinimicrobia" (23). SAR406 occur in numerous marine (5, 23, 26,
28, 57), sedimentary (23), and even oil reservoir (58) environments. They are prevalent in deeper
ocean waters (28, 57, 59) and prefer lower oxygen concentrations in OMZs (5, 26, 60). Our
genomes had larger estimated genome sizes- 2.6-2.7 Mbp (Group A) and 2.8-3.5 Mbp (Group
B)- compared to 1.1-2.4 Mbp from single-cell genomes (23). Overall GC content, however, was
in the range of the 30-48% reported for fosmids (27) and single-cell genomes (23). The lower
GC Group A genomes specifically had a similar GC content to the Arctic96B-7 fosmids,
matching their predicted phylogenetic affiliation (see below) (27).
Our data now also define roles for them in the eutrophication-driven hypoxia of the
nGOM. Previous metabolic reconstructions of SAR406 predicted aerobic metabolism (23) and
sulfur reduction (27), which our data confirm, although the sulfur reduction genes were only
found in Group B organisms (Table S1). Our genomes also suggest multiple nitrogen cycling
roles that appear to be organized by sublineages within the phylum, and sublineage specific
presence of both high and low affinity cytochrome c oxidases. The Group B organisms group
with the early diverging SBH1141 clade (27), for which no previous genome data exist. Group B
organisms contained both types of cytochrome c oxidases, *nosZ* and *nrfAH* genes, whereas

Group A organisms, sister to the Arctic96B-7 clade, contained the low affinity cytochrome c oxidases only, and additionally *narHI* genes not found in Group B. The unique roles predicted for these taxa are not surprising given the diversity of the SAR406 clade and the genetic distances between Group A and B (Fig. S4). The fosmids associated with the Arctic69B-7 clade contained genes for oxidative stress and sulfur reduction (27), although we only found sulfur reduction genes in the distantly related Group B genomes. The ArcticB96-7 clade may be diverse enough to encompass differing metabolic strategies, but the variable presence of *phsA* genes in this group may simply be due to incomplete genomic data. In addition to sublineage-specific respiratory characteristics, our results also generate specific hypotheses about organic matter metabolism in SAR406: likely degradation capacity for cellulose, starch, agar, xylan, and peptidoglycan; transport of nucleosides, amino and fatty acids, and oligopeptides; and substantial gene duplication associated with protein secretion for possible extracellular metabolism. Together these data suggest that during nGOM hypoxia SAR406 members degrade complex carbohydrates fueled by aerobic respiration, and supplemented with facultative anaerobic respiration of nitrate, nitrite, or sulfur compounds.

      Members of the SAR202 clade of Chloroflexi also inhabit a wide variety of marine environments (24), frequently in deeper waters (24, 43, 57, 59, 61) and remain functionally understudied because genome data for SAR202 have been lacking. Landry and colleagues recently described the properties for several single-cell genomes representing SAR202 subclades III and V recovered from the mesopelagic (25). Our genomes have generally higher GC content and much lower expected genome sizes than those predicted by Landry *et al.*, although these calculations are likely complicated by the relative incompleteness of their genomes (8-47%). The Landry *et al.* genomes indicated a role for SAR202 in the oxidation of recalcitrant dissolved organic matter, and specifically cyclic alkanes, via flavin mononucleotide monooxygenases (FMNOs) and different dehydrogenases that occurred in paralogous groups (25). We observed many of the same gene expansions, namely that of MFS transporters and short-chain dehydrogenases (and related genes), but we did not recover any FMNOs of SFams 4832 or 4965, suggesting subclade and/or niche-specific adaptations. Furthermore, we observed *napAB* and *nosZ* genes for nitrate and nitrous oxide reduction (and expression of these genes), which were not reported for subclade III or V. Our nGOM hypoxia SAR202 genomes had CAZy genes implicating them in degradation of complex compounds such as chitin and pectin. The emerging picture of these taxa from both shallow hypoxic waters and the mesopelagic is one of recalcitrant carbon degraders, with overlapping suites of paralogous genes, but that may be specialized for specific compounds more commonly available in their respective habitats.

      This study has also developed roles for CP taxa in a shallow marine water column during hypoxia. The most active organism in our survey based on the ratio of RNA to DNA reads recruited, Bin 13, putatively belongs to a group with little genomic data- ACD39. The original ACD39 genome was reconstructed from an aquifer community (33). Although this was only a partial genome, it shared some features with our putative ACD39 member, namely pilin and chemotaxis genes, those containing TPR and tetratricopeptide repeats, and CAZy genes for

degradation of complex compounds such as starch (33). Our study provides evidence that these
taxa have relatively large genomes (~4.8 Mbp), are active aerobes in nGOM hypoxia, and have
chemotaxis and motility genes that could facilitate scavenging and surface attachment. However,
most of the highly expressed genes in this organism were annotated as hypothetical proteins, so
much of the function of these organisms remains to be uncovered.
Bins 50 and 48 provide novel genome data for bacterioplankton in nGOM hypoxia,
although the exact taxonomic position of these bins remains in conflict. The ribosomal protein
tree provides evidence that these taxa belong to the Latescibacteria (WS3) (Fig. S1), but 16S
rRNA genes (Fig. S6) and our amplicon data point toward membership in the more poorly
understood PAUC34f clade. Since no previous genome data exist for PAUC34f, we cannot rule
out erroneous assignment in the ribosomal protein tree due to insufficient taxon selection. Bins
50 and 48 represented the largest genomes of the study, with estimated complete sizes of ~5-6
Mbp, and numerous genes suggesting degradation of a wider suite of complex organic matter
than any of the other genomes examined. For example, they were the only genomes with
numerous polysaccharide lyase genes, and these likely facilitate breakdown of algal cell wall
components like pectin and alginate. The Bin 50 genome was among the most active across all
samples (Fig. 3), and we detected expression of cytochrome c oxidase genes, and those for
DNRA, in both the Bin 50 and 48 genomes. Thus, we expect these organisms to have an aerobic,
potentially facultatively anaerobic, multifaceted chemoorganoheterotrophic metabolism with
roles in complex carbon compound degradation (like that of algal cell walls) and the nitrogen
cycle.
If these bins belong to PAUC34f, they represent the first genomic data for the group.
Although originally discovered, and commonly found, in marine sponges (32, 62-64), this
putative bacterial phylum (via GreenGenes/SILVA) has been detected as a rare group in other
marine invertebrates (65) and stream sediment (66), and we identified 18 distinct but rare
PAUC34f OTUs in nGOM hypoxia, compared to just three from WS3 (21). Although the
majority of studies suggest an endosymbiotic lifestyle for PAUC34f, our representative genome
data point towards a free-living existence with multiple terminal electron accepting processes,
motility genes for seeking more favorable conditions, and a large metabolic repertoire for
degradation of complex compounds. On the other hand, if these genomes represent WS3, the
sister clade to PAUC34f (Fig. S6), they have many similarities to the lifestyles inferred from
recent metagenomic investigations (67, 68). Specifically, while this group was previously
considered anaerobic (67), new data have supported an aerobic lifestyle for some members (48),
and revealed complete electron transport chains and both high and low affinity cytochrome c
oxidases (68). The Bin 50 and 48 genomes predict aerobic metabolism as well, although only
with low affinity cytochrome c oxidases. Farag *et al.* also found little evidence of these taxa in
host-associated environments, contrary to PAUC34f sequence data (68). The enrichment of PL
family genes in Bins 50 and 48, polysaccharide degradation capability in general, and specific
genes for degradation of cell wall components, all corroborate previous findings on WS3 as well
(68). Bin 50 had 78 annotated peptidases, nearly double that in all other genomes in the study

(Bin 48 had 46), which also concurs with metagenomic predictions for WS3 (68). Our genomes
differed from WS3 metagenomes principally in the predicted DNRA metabolism and the
dramatic expansion of sulfatases. Although sulfatases were observed in WS3 metagenomes (68),
they were not present in the numbers associated with Bin 50 (n=102). A large cadre of sulfatases
has been previously reported for *Lentisphaera* (n=267) and *Pirellula* (n=110) genomes (69, 70)
and suggests specialization for degradation of sulfate-esters to satisfy carbon and/or sulfur
requirements.
Although Parcubacteria and Peregrinibacteria occurred in low abundance (Fig. S7) and
we detected activity in only a few samples, their recovery in the hypoxic zone is notable because
these organisms have generally been associated with anoxic environments. Our predicted
genome sizes (~1.5 Mbp) corroborate previous reports of these organisms having small genomes
(31, 48). We did not observe any genes associated with nitrogen or sulfur redox transitions,
although we cannot rule these capabilities entirely due to incomplete genomes. Regardless, we
can hypothesize that Parcubacteria and Peregrinibacteria persist as members of the rare biosphere
until they can take advantage of microanoxic niches in the water column where they participate
in carbon cycling as obligately fermentative organisms.
Excluding Parcubacteria and Peregrinibacteria, the other uncultivated groups in the
nGOM hypoxic zone had one or more genomes that encoded cytochrome c oxidases (and other
electron transport chain components) for respiring oxygen, making these taxa likely only
facultative anaerobes. Pervasive aerobic metabolism in an oxygen-depleted water column may
seem counterintuitive, yet despite DO being as low as 2.2 μmol kg^{-1} in the E2A sample, oxygen
probably remained high enough to sustain aerobic microbes. As little as ~0.3 μmol kg^{-1} oxygen
inhibited denitrification in OMZ populations by 50% (71), and even *Eschericia coli* K-12 could
grow aerobically at oxygen concentrations as low as 3 nM (72). Thus, for many organisms,
active aerobic respiration likely persists even in suboxic waters during nGOM hypoxia.
Nevertheless, our data also suggests pervasive co-reduction of alternative terminal
electron acceptors (oxygen, nitrate, nitrite, nitrous oxide, and sulfur), sometimes within the same
organism (Fig. 4). Co-reduction of electron acceptors with different redox potentials across a
community could indicate microniches and/or aggregates in the water column where DO
concentrations drop below bulk values (40). Alternatively this can occur with an abundance of
electron donor, and overlapping redox processes have been reported in multiple environments,
including aquatic ones (50, 73). Concurrent expression of genes for multiple terminal electron
accepting processes within a single organism has been proposed as a means of improved
readiness for dynamic conditions, albeit at the cost of lower productivity (40). Given that many
uncultivated taxa likely perform multiple terminal electron accepting processes (and possibly do
so simultaneously), and we found a comparative cornucopia of genes for degradation of
chemoorganohetrotrophic energy sources, we hypothesize that niche differentiation within
uncultivated hypoxic zone bacterioplankton occurs predominantly via specialization for different
oxidizable substrates rather than for distinct roles in the canonical redox cascade (4, 5).

Importantly, many of the active uncultivated taxa also appeared adapted for degradation
of complex carbon substrates. Such compounds might comprise the bulk of available organic
matter during the later stages of hypoxia after initial oxygen depletion by microorganisms
feeding on more labile carbon sources. Selection for chemoorganotrophic microbes adapted to
utilize recalcitrant organic matter could also explain why organisms that do not require an
exogenous carbon source, such as the chemolithoautotrophic *Nitrosopumilus*, proliferate during
hypoxia (21, 22) compared to their levels during spring before DO decreases (74, 75). Temporal
data on the relative abundance and activity of these nGOM microbial dark matter organisms, and
of organic matter composition in the water column, will be critical to more fully understand the
relationship of bacterioplankton to the creation, maintenance, and dissolution of nGOM hypoxia.
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Materials and Methods
Sample selection and nucleic acid processing. Six samples representing hypoxic (n=4) and oxic
(n=2) DO concentrations were picked from among those previously reported (21) at stations D1,
D2, D3, E2, E2A, and E4 (Table S1). DO, and nutrient collection information is detailed in
Gillies *et al.*, 2015. Nucleic acids were collected as follows: At these six stations 10 L of
seawater was collected and filtered with a peristaltic pump. A 2.7 μM Whatman GF/D pre-filter
was used and samples were concentrated on 0.22 μM Sterivex filters (EMD Millipore). Sterivex
filters were immediately sparged, filled with RNAlater, and placed at -20˚C, at which they were
maintained until extraction. DNA and RNA were extracted directly off of the filter by placing
half of the Sterivex filter in a Lysing matrix E (LME) glass/zirconia/silica beads Tube (MP
Biomedicals, Santa Ana, CA) using the protocol described in Gillies *et al.* (2015) which
combines phenol:chloroform:isoamyalcohol (25:24:1) and bead beating. Genomic DNA and
RNA were stored at -80°C until purified. DNA and RNA were purified using QIAGEN
(Valencia, CA) AllPrep DNA/RNA Kit. DNA quantity was determined using a Qubit2.0
Fluorometer (Life Technologies, Grand Island, NY). RNA with an RNA integrity number (RIN)
(16S/23S rRNA ratio determined with the Agilent TapeStation) $\geq$ 8 (on a scale of 1-10, with 1
being degraded and 10 being undegraded RNA) was selected for metatranscriptomic sequencing.
Using a Ribo-Zero kit (Illumina) rRNA was subtracted from total RNA. Subsequently, mRNA
was reverse transcribed to cDNA as described in Mason *et al.* (2012) (76).
538
Sequencing, assembly, and binning. DNA and RNA were sequenced separately, six samples per
lane, with Illumina HiSeq 2000 chemistry to generate 100 bp, paired-end reads (180 bp insert
size) at the Argonne National Laboratory Next Generation Sequencing facility. The data are
available at the NCBI SRA repository under the BioSample accession numbers
SAMN05791315-SAMN05791320 (DNA) and SAMN05791321-SAMN05791326 (RNA).
DNA sequencing resulted in a total of 416,924,120 reads that were quality trimmed to
413,094,662 reads after adaptors were removed using Scythe
(https://github.com/vsbuffalo/scythe), and low-quality reads (Q < 30) were trimmed with Sickle

(https://github.com/najoshi/sickle). Reads with three or more Ns or with average quality score of
less than Q20 and a length < 50 bps were removed. Genomes were reconstructed using two
rounds of assembly. Metagenomic reads from all six samples were pooled, assembled, and
binned using previously described methods (77, 78). Briefly, quality filtered reads were
assembled with IDBA-UD (79) on a 1TB RAM, 40-core node at the LSU High Performance
Computing cluster SuperMikeII, using the following settings: -mink 65 –maxk 105 –step 10 –
pre_correction –seed_kmer 55. Initial binning of the assembled fragments was performed using
tetra-nucleotide frequency signatures using 5 kbp fragments of the contigs. Emergent self-
organizing maps (ESOM) were manually delineated and curated based on clusters within the
map. The primary assembly utilized all reads and produced 28,080 contigs ≥ 3 kb totaling
217,715,956 bp. Of these, 303 contigs were over 50 kb, 72 over 100 kb, and the largest contig
was just under 495 kb. Binning produced 76 genomes, of which 20 genomes were assigned to
lineages with uncultivated representatives using CheckM, ribosomal protein trees, and 16S
rRNA gene sequences (below).
561
562 *DNA and RNA mapping.* Metagenomic and metatranscriptomic sequencing reads from each
563 sample were separately mapped to binned contigs using BWA (80) to compare bin abundance
564 across samples and facilitate bin cleanup (below). Contigs within each bin were concatenated
565 into a single fasta sequence and BWA was used to map the reads from each sample to all bins.
566 All commands used for these steps are available in supplementary information.
567
Bin QC. Bins were examined for contamination and completeness with CheckM (35), and we
attempted to clean bins with > 10% estimated contamination using a combination of methods.
First, the CheckM modify command removed contigs determined to be outliers by GC content,
coding density, and tetranucleotide frequency. Next, in bins that still showed > 10%
contamination, contigs were separated according to comparative relative abundance of mean
DNA read coverage by sample. Final bins were evaluated with CheckM again to generate the
statistics in Table S1 and final bin placements in the CheckM concatenated gene tree (Fig. S2).
575
Ribosomal protein tree. The concatenated ribosomal protein tree was generated using 16
syntenic genes that have been shown to undergo limited lateral gene transfer (rpL2, 3, 4, 5, 6, 14,
15, 16, 18, 22, 24 and rpS3, 8, 10, 17, 19) (81). Ribosomal proteins for each bin were identified
with Phylosift (82). Amino acid alignments of the individual ribosomal proteins were generated
using MUSCLE (83) and trimmed using BMGE (84) (with the following settings: -m
BLOSUM30 –g 0.5). The curated alignments were then concatenated for phylogenetic analyses
and phylogeny inferred via RAxML v 8.2.8 (85) with 100 bootstrap runs (with the following
settings: mpirun -np 4 -npernode 1 raxmlHPC-HYBRID-AVX -f a -m PROTCATLG -T 16 -p
12345 -x 12345 -# 100). Note this is similar to the number utilized in a previous publication for
this tree with automated bootstrapping (86), and required just over 56 hours of wall clock time.
The alignment is available in SI.

Average amino acid identity. AAI was calculated with Get Homologues (87) v. 02032017, with
the following settings: –M –t 0 –n 16 –A.

Taxonomic assignment. Taxonomy for each bin was assigned primarily using the ribosomal
protein tree. However, for bins that did not have enough ribosomal proteins to be included in the
tree, or for which the placement within the tree was poorly supported, assignments were made
based on the concatenated marker gene tree as part of the CheckM analysis (Fig. S2), or via 16S
rRNA gene sequences, when available. 16S rRNA genes were identified via CheckM, and these
sequences were aligned against the NCBI nr database using BLASTN to corroborate CheckM
assignments. In the case of the SAR202 genomes, which did not have representative genomes in
either the ribosomal protein tree or the CheckM tree, the 16S rRNA gene sequences for two of
the three bins (43-1, 43-2) were available and aligned with the sequences used to define the
SAR202 clade (24) (Fig. S5). Alignment, culling, and inference were completed with MUSCLE
(83), Gblocks (88), and FastTree2 (89), respectively, with the FT_pipe script. The script is
provided in SI. The 16S rRNA gene tree for subclade assignment of SAR406 (Fig. S4) was
assembled by blasting the four 16S sequences predicted by CheckM against a local GenBank nt
database using blastn (v. 2.2.28+) (90), selected the top 100 non-redundant hits to each sequence,
and manually removing all hits to genome sequences. These were combined with previously
defined SAR406 subclade reference sequences (26), fosmid 16S sequences (27), single-cell
genome sequences (23), and run through alignment, culling, and inference with FT_pipe. Taxa
with identical alignments were removed with RAxML v 8.2.8 (85) using default settings, and the
final tree was inferred using FastTree2 (89). For putative CP genomes, taxonomy was also
evaluated by examining the taxonomic identification for each of the predicted protein sequences
after a BLASTP search against the NCBI nr database. Post-blast, the number of assignments to
the dominant one or two taxonomic names, along with the number of assignments to "uncultured
bacterium," was plotted for each genome according to the bit score quartile (Fig. S3). Quartiles
were determined in R using the summary function. Bin 56 has two ribosomal protein operons on
scaffold_2719/Ga0113622_1153 and scaffold_21777/Ga0113622_1009. In the ribosomal protein
tree, the former placed the organism in the *Planctomycetaceae*, while the latter (which was much
smaller) placed the organism in CP WS3. The majority of BLASTP annotations to the nr
database matched *Planctomycetaceae* taxa, as did the 16S rRNA gene sequences found in the
genome, so Bin 56 organism was designated a Planctomycetes and not WS3, and excluded from
this study. The 16S rRNA gene from Bin 50 was also used to infer taxonomic identity using and
established phylogeny for the WS3 clade (68) and relevant outgroups. The Bin 50 sequence was
blasted against the greengenes database (Dec. 2013) with megablast, and since many of the top
hits belonged to the PAUC34f clade, these were included with the sequences from Farag *et al.*
2017. Alignment, culling, and inference was completed with FT_pipe. Node labels were
constructed with the newick utilities (91) script nw_rename.

Metabolic reconstruction. Post-binning, genomes were submitted individually to IMG (92) for
annotation. Genome accession numbers are in Table S1, and all are publically available.
Metabolic reconstruction found in Table S1 and Figs S5-7/ S11-13, came from these annotations
and inspection with IMG's analysis tools, including KEGG pathway assignments and transporter
predictions. Transporters highlighted for DOM uptake were identified based on information at
the Transporter Classification Database (93). Carbohydrate-active enzymes (CAZymes) were
predicted using the same routines as those in operation for the updates of the carbohydrate-active
enzymes database (www.cazy.org) (94).
635

RPKM abundance of taxa and genes. Abundance of taxa within the sample was quantified by
evaluating mapped reads using Reads-Per-Kilobase-Per-Million (RPKM) normalization (95)
according to $A_{ij} = (N_{ij}/L_i) \times (1/T_j)$, where A_{ij} is the abundance of bin i in sample j, N_{ij} is the
number of reads that map to bin from sample j, L_i is the length of bin i in kilobases, and T_j is the
total number of reads in sample j divided by 10^6. These values were generated for all bins, with
only the data for the 20 uncultivated bins reported here. All contigs within a given bin were
artificially concatenated into "supercontigs" prior to mapping. N_{ij} was generated using the
samtools (80) idxstats function after mapping with BWA. The data in Fig. S7 were created by
summing (N_{ij}/L_i) for groups of taxa defined in Table S1 prior to multiplying by $(1/T_j)$. RNA
coverage was used to evaluate both bin and gene activity for all bins. Mean coverage for each
supercontig was calculated using bedtools (96) and bins were assigned a rank from lowest mean
recruitment (1) to highest mean recruitment (2). Bins with particularly high or low activity
(transcript abundance) relative to their abundance (genome abundance) were identified using
rank-residuals, calculated as follows: On a plot of DNA coverage rank vs. RNA coverage rank,
residuals for each bin or gene were calculated from the identity. As the rank-residuals followed a
Gaussian distribution, bins with a residual that was > 1 s.d. from the rank-residual mean were
classified as having higher-than-expected transcriptional activity; bins with a residual that was <
1 s.d. from the mean were classified as having lower-than-expected transcriptional activity.
RPKM values were also calculated for every gene in every bin analogously to that for bins, using
RNA mapping values extracted with the bedtools multicov function. Sample E2 was omitted
from gene-specific calculations as only 4588 transcriptomic reads mapped successfully from this
sample, compared to >100,000 from other samples. 17,827 of 140,347 genes had no evidence of
expression in any sample and so were removed from further analysis. 3,840 genes recruited reads
in all remaining samples. All calculations are available in Table S1 or the R markdown document
Per.gene.RPKM.Rmd in Supplemental Information. Table S1 includes only analyzed data for the
uncultivated bins reported in this study. Note that RPKM values indicate abundance
measurements across a small number of samples. While we can evaluate the relative expression
of genes for those samples, our dataset lacks sufficient power to evaluate estimates of
significance in differential expression.
665

nrfA sequence assessment. Initial annotation of our bins identified putative homologs to the *nrfAH* genes associated with dissimilatory nitrite reduction to ammonia. Since *nrfA*-type nitrite reductases can be misannotated due to homology with other nitrite reductases, annotation for these genes was curated with phylogenetic analysis using known *nrfA* genes (38) obtained via Dr. Welsh (personal communication). Alignment, culling, and inference were completed with the FT_pipe script. The tree was rooted on the designated outgroup octaheme nitrite reductase sequence from *Thioalkalivibrio nitratireducens* ONR. Node labels were constructed with the newick utilities (91) script nw_rename. Visualization of the alignment (Fig. S8B,C) to confirm the presence of the first CXXCK/CXXCH and highly conserved KXQH/KXRH catalytic site was completed with the MSAViewer (97) online using the un-culled *nrfA* alignment as input.

SFam homology searches. To identify group specific expansions in particular gene families, we performed a homology search of all predicted protein coding sequences in each bin against the Sifted Families (SFam) database (41) using hmmsearch (HMMER 3.1b (98)) with default settings except for the utilization of 16 cpus per search.

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Author contributions

JCT and OUM designed the study. LEG, NNR, and JCT collected samples. NNR provided processed oceanographic data. LEG and OUM extracted, quantified, and determined quality of nucleic acids. JCT, KWS, and BJB reconstructed the genomes. KWS, BJB, BT, BH, and JCT conducted downstream analyses. JCT led manuscript writing and all co-authors evaluated and contributed edits.

Competing financial interests

706   The authors declare no competing financial interests.

708 **Table 1.** Genome characteristics for the 20 bins associated with uncultivated lineages.

IMG Genome ID	Bin Id	Taxonomy	Compl %	Contam %	Strain het.	Scaff.	Longest scaff. (bp)	Size (bp)	Genes	GC (fract.)	Coding density (fract.)	Estim. Compl. Size (Mbp)
2651870035	43-1	Chloroflexi (SAR202)	90.3	0	0	69	161242	1972793	1882	0.52	0.88	2.2
2651870036	43-2	Chloroflexi (SAR202)	88.6	4.1	15.4	134	157345	2402386	2373	0.52	0.91	2.7
2651870034	43	Chloroflexi (SAR202)	83.2	0.1	100	179	90503	2475308	2392	0.53	0.89	3.0
2693429801	45	Marinimicrobia (SAR406) –B	89.8	5.5	71.4	124	105140	2811623	2487	0.47	0.93	3.1
2651870052	45-1	Marinimicrobia (SAR406) –B	85.2	0.1	100	144	100582	2410233	2235	0.49	0.93	2.8
2693429802	45-2	Marinimicrobia (SAR406) –B	79.3	1.8	12.5	287	61408	2811444	2554	0.46	0.94	3.5
2651870053	51-1	Marinimicrobia (SAR406) –A	73.6	1.7	50	75	191525	1901306	1835	0.39	0.95	2.6
2651870051	51	Marinimicrobia (SAR406) –A	21.3	0	0	115	16923	578802	686	0.41	0.95	2.7
2651870038	15	Euryarchaeota (MGII)	83.2	1.6	0	43	255599	1885130	1614	0.62	0.95	2.3
2651870039	17	Euryarchaeota (MGII)	81.9	0.1	50	88	137319	1803861	1564	0.43	0.96	2.2
2651870037	14	Euryarchaeota (MGII)	71.2	0.8	100	80	90069	1389909	1305	0.54	0.94	2.0
2651870040	18	Euryarchaeota (MGII)	61.1	0.8	100	155	20633	1033226	1025	0.50	0.96	1.7
2651870042	38	Euryarchaeota (MGII)	27.1	0	0	138	12893	615290	610	0.55	0.96	2.3
2651870041	17-1	Euryarchaeota (MGII)	16.6	2.8	33.3	123	16952	538052	566	0.41	0.95	3.2
2693429807	13	Unclassified (ACD39)	89.8	5.1	20	401	75104	4269849	3686	0.47	0.93	4.8
2693429799	40	Parcubacteria (OD1)	71.9	15.1	75	97	65104	1086283	1208	0.52	0.92	1.5
2693429797	16	Peregrinibacteria	83.2	0.3	100	67	59308	1384712	1318	0.39	0.93	1.7
2693429798	39	Peregrinibacteria	49.9	0.3	100	131	18806	747520	809	0.45	0.95	1.5
2693429804	50	Unclassified (PAUC34f)	84.8	1.4	0	455	76269	5346994	5484	0.58	0.92	6.3
2693429803	48	Unclassified (PAUC34f)	51.9	2.2	0	470	20176	2566149	2596	0.55	0.94	4.9

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711

712   **Figure 1.** Metabolic reconstruction of Marine Group II Euryarchaeota, SAR406, and SAR202,
713   based on the top three or four most complete genomes. Colors indicate pathway elements based
714   on the number of genomes in which they were recovered, according to the key. Black outlines
715   and/or arrows indicate genes that were not observed. Numbers correspond to annotations
716   supplied in Table S1.
717
718   **Figure 2.** Metabolic reconstruction of the Candidate Phylum members PAUC34f, Parcubacteria
719   (OD1), Peregrinibacteria, and ACD39 (Bin 13). Colors indicate pathway elements based on the
720   number of genomes in which they were recovered, according to the key. Black outlines and/or
721   arrows indicate genes that were not observed. Numbers correspond to annotations supplied in
722   Table S1.
723
Figure 3. Relative DNA to RNA recruitment rank for each genome, by sample. Colors indicate
the relative difference in the ratio of rank based on total RNA and DNA mapping. Red indicates
a higher RNA recruitment rank compared to DNA recruitment rank, and vice-versa for blue. +
and - symbols indicate bins where the rank-residual from the identity in RNA vs. DNA read
mapping was more or less than one standard deviation beyond 0, respectively. Dendrograms
were calculated using an Unweighted Pair Group Method with Arithmetic Mean (UPGMA) from
Euclidian distances of rank residuals across all samples and bins.
731
732   **Figure 4.** Expression of predicted respiratory genes. RPKM values of RNA recruitment for each
733   gene, by sample, are depicted with colors according to the key (yellow to blue follows increasing
734   intensity). Genes are grouped by bin, taxonomic affiliation, and specific respiratory process.
735   DNRA- dissimilatory nitrite reduction to ammonia.
736
737   **Figure 5.** Expression of predicted CAZy genes. RPKM values or RNA recruitment for each
738   gene, by sample, are depicted with colors according to the key (yellow to red follows increasing
739   intensity). Genes are grouped by bin, taxonomic affiliation, and general CAZy categories. CE-
740   carbohydrate esterase; GH- glycoside hydrolase; GT- glycosyltransferase; CBM- carbohydrate
741   binding module.
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Supplemental Information

Supplemental Text provides additional information on taxonomic assignments.

Table S1. Spreadsheet (Table_S1.xlsx) containing information on taxonomy, CheckM results, IMG statistics, partial metabolic reconstruction, gene annotations associated with Figures 1 and 2, transporter classifications, CAZy predictions, sample chemical data, RPKM values and gene neighborhoods for WS3 cytochrome c oxidases and *nrfA* genes from SAR406 and WS3.

Figure S1. Maximum likelihood tree of concatenated ribosomal protein coding genes. Values at internal nodes indicate bootstrap support (n=100). Scale bar indicates changes per position.

Figure S2. Phylogenetic placement of bins based on CheckM.

Figure S3. Annotations of protein-coding gene sequence best blastp hits in the nr database, divided into quartiles by bit score, for CP bacteria.

Figure S4. SAR406 16S rRNA gene phylogeny. Genes recovered from assembled bins have 45* or 51* designations. Values at nodes indicate Shimodaira-Hasegawa "like" values (89). Scale bar indicates changes per position.

Figure S5. SAR202 16S rRNA gene phylogeny. Genes recovered from assembled bins are indicated as 43-*. Subclades are designated according to Morris *et al.* 2004, and the tree is rooted according to Figure 1 in that publication. Values at nodes indicate Shimodaira-Hasegawa "like" values (89). Scale bar indicates changes per position.

Figure S6. 16S rRNA gene phylogeny of the WS3 clade (68) with added PAUC34f sequences from the GreenGenes database and the Bin 50 sequence. Tree is rooted on the Archaea according to Fig. S1 in Farag *et al.* 2017. Values at nodes indicate Shimodaira-Hasegawa "like" values (89). Scale bar indicates changes per position.

Figure S7. Metagenomic RPKM values for each group, comprised of aggregated values for each bin within the group. Values are plotted according to sample and colored according to the dissolved oxygen (DO) concentration from where the sample was taken.

Figure S8. Evaluation of predicted nrfA genes in SAR406 and Bins 50/48. A) Phylogenetic tree of predicted *nrfA* genes. Additional taxa are from Figure 3 in Welsh *et al.*, 2014. The tree was rooted at the midpoint. Values at nodes indicate Shimodaira-Hasegawa "like" values (89). Scale bar indicates changes per position. B&C) Conserved catalytic motifs within the *nrfA* gene. B) Black square surrounds the first heme-binding CXXCK/CXXCH motif. C) Black square

surrounds the catalytic KXQH/KXRH motif. The alignment follows highlighting found in Welsh

et al., 2014 (38). All genes numbers from this study are indicated as 26536*, corresponding to

rows 64 and 66-68.

Additional Supplemental Information such as scripts, workflows, and key files, including fasta

files for each tree, are provided as a link hosted at the Thrash Lab website:

http://thethrashlab.com/publications.

794   References Cited
795

1.   Rabalais NN, Turner RE, Dortch Q, Justic D, Bierman VJ, Wiseman Jr WJ. 2002.
Nutrient-enhanced productivity in the northern Gulf of Mexico: past, present and future.
Hydrobiologia 475:39-63.
2.   Diaz RJ, Rosenberg R. 2008. Spreading Dead Zones and Consequences for Marine
Ecosystems. Science 321:926-929.
3.   Anonymous. Action Plan 2008 for Reducing Mitigating, And Controlling Hypoxia in the
Northern Gulf of Mexico and Improving Water Quality in the Mississippi River Basin.
Mississippi River Gulf of Mexico Watershed Nutrient Task Force, U.S. Environmental
Protection Agency, Office of Wetlands, Oceans, and Watersheds: Washington, D.C.,
4.   Ulloa O, Canfield DE, DeLong EF, Letelier RM, Stewart FJ. 2012. Microbial
oceanography of anoxic oxygen minimum zones. Proceedings of the National Academy
of Sciences 109:15996-16003.
5.   Wright JJ, Konwar KM, Hallam SJ. 2012. Microbial ecology of expanding oxygen
minimum zones. Nature Reviews Microbiology 10:381-394.
6.   Grote J, Jost G, Labrenz M, Herndl GJ, Jurgens K. 2008. Epsilonproteobacteria
Represent the Major Portion of Chemoautotrophic Bacteria in Sulfidic Waters of Pelagic
Redoxclines of the Baltic and Black Seas. Applied and Environmental Microbiology
74:7546-7551.
7.   Glaubitz S, Kiesslich K, Meeske C, Labrenz M, Jurgens K. 2013. SUP05 Dominates the
Gammaproteobacterial Sulfur Oxidizer Assemblages in Pelagic Redoxclines of the
Central Baltic and Black Seas. Applied and Environmental Microbiology 79:2767-2776.
8.   Friedrich J, Janssen F, Aleynik D, Bange HW, Boltacheva N, Çagatay MN, Dale AW,
Etiope G, Erdem Z, Geraga M, Gilli A, Gomoiu MT, Hall POJ, Hansson D, He Y,
Holtappels M, Kirf MK, Kononets M, Konovalov S, Lichtschlag A, Livingstone DM,
Marinaro G, Mazlumyan S, Naeher S, North RP, Papatheodorou G, Pfannkuche O, Prien
R, Rehder G, Schubert CJ, Soltwedel T, Sommer S, Stahl H, Stanev EV, Teaca A,
Tengberg A, Waldmann C, Wehrli B, Wenzhöfer F. 2014. Investigating hypoxia in
aquatic environments: diverse approaches to addressing a complex phenomenon.
Biogeosciences 11:1215-1259.
9.   Lam P, Kuypers M. 2011. Microbial Nitrogen Cycling Processes in Oxygen Minimum
Zones. Annual review of marine science 3:317-345.
10.   Beman JM, Popp BN, Alford SE. 2012. Quantification of ammonia oxidation rates and
ammonia-oxidizing archaea and bacteria at high resolution in the Gulf of California and
eastern tropical North Pacific Ocean. Limnology and Oceanography 57:711-726.
11.   Saad EM, Longo AF, Chambers LR, Huang R, Benitez‐Nelson C, Dyhrman ST, Diaz
JM, Tang Y, Ingall ED. 2016. Understanding marine dissolved organic matter
production: Compositional insights from axenic cultures of Thalassiosira pseudonana.
Limnology and Oceanography AOP. doi: 10.1002/lno.10367.
12.   Canfield DE, Stewart FJ, Thamdrup B, De Brabandere L, Dalsgaard T, DeLong EF,
Revsbech NP, Ulloa O. 2010. A Cryptic Sulfur Cycle in Oxygen-Minimum-Zone Waters
off the Chilean Coast. Science 330:1375-1378.
13.   Stewart FJ, Ulloa O, DeLong EF. 2012. Microbial metatranscriptomics in a permanent
marine oxygen minimum zone. Environmental Microbiology 14:23-40.

14.   Roberts BJ, Doty SM. 2015. Spatial and Temporal Patterns of Benthic Respiration and
840   Net Nutrient Fluxes in the Atchafalaya River Delta Estuary. Estuaries and Coasts
841   38:1918-1936.
842   15.   McCarthy MJ, Carini SA, Liu Z, Ostrom NE, Gardner WS. 2013. Oxygen consumption
843   in the water column and sediments of the northern Gulf of Mexico hypoxic zone.
844   Estuarine, Coastal and Shelf Science 123:46-53.
845   16.   Schaeffer BA, Sinclair GA, Lehrter JC, Murrell MC, Kurtz JC, Gould RW, Yates DF.
846   2011. An analysis of diffuse light attenuation in the northern Gulf of Mexico hypoxic
847   zone using the SeaWiFS satellite data record. Remote Sensing of Environment 115:3748-
848   3757.
849   17.   Rabalais NN, Turner RE, Gupta BKS, Boesch DF, Chapman P, Murrell MC. 2007.
850   Hypoxia in the northern Gulf of Mexico: Does the science support the Plan to Reduce,
851   Mitigate, and Control Hypoxia? Estuaries and Coasts 30:753-772.
852   18.   Rabalais NN, Turner RE, Wiseman Jr WJ. 2001. Hypoxia in the Gulf of Mexico. Journal
853   of Environmental Quality 30:320-329.
854   19.   Rabalais NN, Turner RE, Wiseman JWJ. 2002. Gulf of Mexico hypoxia, AKA "The dead
855   zone". Annual Review of Ecology and Systematics 33:235-63.
856   20.   Wiseman Jr WJ, Rabalais NN, Turner RE, Dinnel SP, MacNaughton A. 1997. Seasonal
857   and interannual variability within the Louisiana coastal current: stratification and
858   hypoxia. Journal of Marine Systems 12:237-248.
859   21.   Gillies LE, Thrash JC, deRada S, Rabalais NN, Mason OU. 2015. Archaeal enrichment in
860   the hypoxic zone in the northern Gulf of Mexico. Environmental Microbiology 17:3847.
861   22.   Bristow LA, Sarode N, Cartee J, Caro-Quintero A, Thamdrup B, Stewart FJ. 2015.
862   Biogeochemical and metagenomic analysis of nitrite accumulation in the Gulf of Mexico
863   hypoxic zone. Limnology and Oceanography 60:1733-1750.
864   23.   Rinke C, Schwientek P, Sczyrba A, Ivanova NN, Anderson IJ, Cheng J-F, Darling A,
865   Malfatti S, Swan BK, Gies EA, Dodsworth JA, Hedlund BP, Tsiamis G, Sievert SM, Liu
866   W-T, Eisen JA, Hallam SJ, Kyrpides NC, Stepanauskas R, Rubin EM, Hugenholtz P,
867   Woyke T. 2013. Insights into the phylogeny and coding potential of microbial dark
868   matter. Nature 499:431-437.
869   24.   Morris RM, Rappé MS, Urbach E, Connon SA, Giovannoni SJ. 2004. Prevalence of the
870   Chloroflexi-Related SAR202 Bacterioplankton Cluster throughout the Mesopelagic Zone
871   and Deep Ocean. Applied and Environmental Microbiology 70:2836-2842.
872   25.   Landry Z, Swan BK, Herndl GJ, Stepanauskas R, Giovannoni SJ. 2017. SAR202
873   Genomes from the Dark Ocean Predict Pathways for the Oxidation of Recalcitrant
874   Dissolved Organic Matter. mBio 8:17.
875   26.   Allers E, Wright JJ, Konwar KM, Howes CG, Beneze E, Hallam SJ, Sullivan MB. 2012.
876   Diversity and population structure of Marine Group A bacteria in the Northeast subarctic
877   Pacific Ocean. The ISME Journal 7:256-268.
878   27.   Wright JJ, Mewis K, Hanson NW, Konwar KM, Maas KR, Hallam SJ. 2014. Genomic
879   properties of Marine Group A bacteria indicate a role in the marine sulfur cycle. The
880   ISME Journal 8:455-468.
881   28.   Gordon DA, Giovannoni SJ. 1996. Detection of stratified microbial populations related to
882   Chlorobium and Fibrobacter species in the Atlantic and Pacific oceans. Applied and
883   Environmental Microbiology 62.

29. Parada AE, Needham DM, Fuhrman JA. 2015. Every base matters: assessing small
subunit rRNA primers for marine microbiomes with mock communities, time series and
global field samples. Environmental Microbiology 18:1403-1414.
30. Brown CT, Hug LA, Thomas BC, Sharon I, Castelle CJ, Singh A, Wilkins MJ, Wrighton
KC, Williams KH, Banfield JF. 2015. Unusual biology across a group comprising more
than 15% of domain Bacteria. Nature 523:208-211.
31. Wrighton KC, Thomas BC, Sharon I, Miller CS, Castelle CJ, Verberkmoes NC, Wilkins
MJ, Hettich RL, Lipton MS, Williams KH, Long PE, Banfield JF. 2012. Fermentation,
Hydrogen, and Sulfur Metabolism in Multiple Uncultivated Bacterial Phyla. Science
337:1661-1665.
32. Hentschel U, Hopke J, Horn M, Friedrich AB, Wagner M, Hacker J, Moore BS. 2002.
Molecular evidence for a uniform microbial community in sponges from different oceans.
Applied and Environmental Microbiology 68:4431-4440.
33. Wrighton KC, Castelle CJ, Wilkins MJ, Hug LA, Sharon I, Thomas BC, Handley KM,
Mullin SW, Nicora CD, Singh A, Lipton MS, Long PE, Williams KH, Banfield JF. 2014.
Metabolic interdependencies between phylogenetically novel fermenters and respiratory
organisms in an unconfined aquifer. The ISME Journal 8:1452-1463.
34. Hug LA, Baker BJ, Anantharaman K, Brown CT, Probst AJ, Castelle CJ, Butterfield CN,
Hernsdorf AW, Amano Y, Ise K, Suzuki Y, Dudek N, Relman DA, Finstad KM,
Amundson R, Thomas BC, Banfield JF. 2016. A new view of the tree of life. Nature
Microbiology 1.
35. Parks DH, Imelfort M, Skennerton CT, Hugenholtz P, Tyson GW. 2015. CheckM:
assessing the quality of microbial genomes recovered from isolates, single cells, and
metagenomes. Genome Research 25:1043-1055.
36. Cantarel BL, Coutinho PM, Rancurel C, Bernard T, Lombard V, Henrissat B. 2009. The
Carbohydrate-Active EnZymes database (CAZy): an expert resource for Glycogenomics.
Nucleic Acids Research 37:1–6. doi:10.1093/nar/gkn663.
37. Pitcher RS, Watmough NJ. 2004. The bacterial cytochrome cbb3 oxidases. Biochimica et
biophysica acta 1655:388-399.
38. Welsh A, Chee-Sanford JC, Connor LM, Löffler FE, Sanford RA. 2014. Refined NrfA
Phylogeny Improves PCR-Based nrfA Gene Detection. Applied and Environmental
Microbiology 80:2110-2119.
39. Burns JL, DiChristina TJ. 2009. Anaerobic respiration of elemental sulfur and thiosulfate
by Shewanella oneidensis MR-1 requires psrA, a homolog of the phsA gene of
Salmonella enterica serovar typhimurium LT2. Applied and environmental microbiology
75:5209-5217.
40. Chen J, Hanke A, Tegetmeyer HE, Kattelmann I, Sharma R, Hamann E, Hargesheimer T,
Kraft B, Lenk S, Geelhoed JS, Hettich RL, Strous M. 2017. Impacts of chemical
gradients on microbial community structure. The ISME Journal 11:920-931.
41. Sharpton TJ, Jospin G, Wu D, Langille MGI, Pollard KS, Eisen JA. 2012. Sifting through
genomes with iterative-sequence clustering produces a large, phylogenetically diverse
protein-family resource. BMC Bioinformatics 13:264.
42. McBride MJ, Zhu Y. 2013. Gliding Motility and Por Secretion System Genes Are
Widespread among Members of the Phylum Bacteroidetes. Journal of Bacteriology
195:270-278.

43. Varela MM, Aken HM, Herndl GJ. 2008. Abundance and activity of Chloroflexi‐type
SAR202 bacterioplankton in the meso‐ and bathypelagic waters of the (sub)tropical
Atlantic. Environmental Microbiology 10:1903-1911.
44. Law CJ, Maloney PC, Wang D-N. 2008. Ins and Outs of Major Facilitator Superfamily
Antiporters. Annual Review of Microbiology 62:289-305.
45. Lombard V, Bernard T, Rancurel C, Brumer H, Coutinho PM, Henrissat B. 2010. A
hierarchical classification of polysaccharide lyases for glycogenomics. Biochemical
Journal 432:437-444.
46. Cregut M, Piutti S, Slezack-Deschaumes S, Benizri E. 2013. Compartmentalization and
regulation of arylsulfatase activities in Streptomyces sp., Microbacterium sp. and
Rhodococcus sp. soil isolates in response to inorganic sulfate limitation. Microbiological
Research 168:12-21.
47. Cregut M, Durand M-J, Thouand G. 2014. The Diversity and Functions of Choline
Sulphatases in Microorganisms. Microbial Ecology 67:350-357.
48. Anantharaman K, Brown CT, Hug LA, Sharon I, Castelle CJ, Probst AJ, Thomas BC,
Singh A, Wilkins MJ, Karaoz U, Brodie EL, Williams KH, Hubbard SS, Banfield JF.
2016. Thousands of microbial genomes shed light on interconnected biogeochemical
processes in an aquifer system. Nature Communications 7:13219.
49. Luef B, Frischkorn KR, Wrighton KC, Holman H-YN, Birarda G, Thomas BC, Singh A,
Williams KH, Siegerist CE, Tringe SG, Downing KH, Comolli LR, Banfield JF. 2015.
Diverse uncultivated ultra-small bacterial cells in groundwater. Nature Communications
6:6372.
50. Alewell C, Paul S, Lischeid G, Storck FR. 2008. Co-regulation of redox processes in
freshwater wetlands as a function of organic matter availability? Science of The Total
Environment 404:335-342.
51. Li M, Baker BJ, Anantharaman K, Jain S, Breier JA, Dick GJ. 2015. Genomic and
transcriptomic evidence for scavenging of diverse organic compounds by widespread
deep-sea archaea. Nature Communications 6:8933. DOI: 10.1038/ncomms9933.
52. Zhang CL, Xie W, Martin-Cuadrado A-B, Rodriguez-Valera F. 2015. Marine Group II
Archaea, potentially important players in the global ocean carbon cycle. Frontiers in
Microbiology 6:1108.
53. Orsi WD, Smith JM, Liu S, Liu Z, Sakamoto CM, Wilken S, Poirier C, Richards TA,
Keeling PJ, Worden AZ, Santoro AE. 2016. Diverse, uncultivated bacteria and archaea
underlying the cycling of dissolved protein in the ocean. The ISME Journal 10:2158-
2173.
54. Ouverney CC, Fuhrman JA. 2000. Marine planktonic archaea take up amino acids.
Applied and environmental microbiology 66:4829-4833.
55. Iverson V, Morris RM, Frazar CD, Berthiaume CT, Morales RL, Armbrust EV. 2012.
Untangling genomes from metagenomes: revealing an uncultured class of marine
Euryarchaeota. Science 335:587-590.
56. Fuhrman JA, McCallum K, Davis AA. 1993. Phylogenetic diversity of subsurface marine
microbial communities from the Atlantic and Pacific Oceans. Applied and Environmental
Microbiology 59:1294-1302.
57. Schattenhofer M, Fuchs BM, Amann R, Zubkov MV, Tarran GA, Pernthaler J. 2009.
Latitudinal distribution of prokaryotic picoplankton populations in the Atlantic Ocean.
Environmental Microbiology 11:2078-2093.

58. Hu P, Tom L, Singh A, Thomas BC, Baker BJ, Piceno YM, Andersen GL, Banfield JF.
2016. Genome-Resolved Metagenomic Analysis Reveals Roles for Candidate Phyla and
Other Microbial Community Members in Biogeochemical Transformations in Oil
Reservoirs. mBio 7:15.
59. DeLong EF, Preston CM, Mincer T, Rich V, Hallam SJ, Frigaard N-UU, Martinez A,
Sullivan MB, Edwards R, Brito BR, Chisholm SW, Karl DM. 2006. Community
genomics among stratified microbial assemblages in the ocean's interior. Science (New
York, NY) 311:496-503.
60. Fuchs BM, Woebken D, Zubkov MV, Burkill P, Amann R. 2005. Molecular
identification of picoplankton populations in contrasting waters of the Arabian Sea.
Aquatic Microbial Ecology 39:145-157.
61. Treusch AH, Vergin KL, Finlay LA, Donatz MG, Burton RM, Carlson CA, Giovannoni
SJ. 2009. Seasonality and vertical structure of microbial communities in an ocean gyre.
The ISME Journal 3:1148-1163.
62. Weigel BL, Erwin PM. 2017. Effects of reciprocal transplantation on the microbiome and
putative nitrogen cycling functions of the intertidal sponge, Hymeniacidon heliophila.
Scientific Reports 7:43247.
63. Jeong I-H, Kim K-H, Park J-S. 2013. Analysis of bacterial diversity in sponges collected
off Chujado, an Island in Korea, using barcoded 454 pyrosequencing: Analysis of a
distinctive sponge group containing Chloroflexi. Journal of Microbiology 51:570-577.
64. Cuvelier ML, Blake E, Mulheron R, McCarthy PJ, Blackwelder P, Thurber RL, Lopez
JV. 2014. Two distinct microbial communities revealed in the sponge Cinachyrella.
Frontiers in Microbiology 5:581.
65. Erwin PM, Pineda M, Webster N, Turon X, López-Legentil S. 2013. Down under the
tunic: bacterial biodiversity hotspots and widespread ammonia-oxidizing archaea in coral
reef ascidians. The ISME Journal 8:575-588.
66. Costa PS, Reis MP, Ávila MP, Leite LR, de Araújo FMG, Salim ACM, Oliveira G,
Barbosa F, Chartone-Souza E, Nascimento AMA. 2015. Metagenome of a Microbial
Community Inhabiting a Metal-Rich Tropical Stream Sediment. PLOS ONE
10:e0119465.
67. Youssef NH, Farag IF, Rinke C, Hallam SJ, Woyke T, Elshahed MS. 2015. In Silico
Analysis of the Metabolic Potential and Niche Specialization of Candidate Phylum
"Latescibacteria" (WS3). PLOS ONE 10:10.1371/journal.pone.0127499.
68. Farag IF, Youssef NH, Elshahed MS. 2017. Global Distribution Patterns and Pangenomic
Diversity of the Candidate Phylum "Latescibacteria" (WS3). Applied and Environmental
Microbiology 83: e00521-17.
69. Thrash JC, Cho JC, Vergin KL, Morris RM, Giovannoni SJ. 2010. Genome Sequence of
Lentisphaera araneosa HTCC2155T, the Type Species of the Order Lentisphaerales in the
Phylum Lentisphaerae. Journal of Bacteriology 192:2938-2939.
70. Krüger M, Meyerdierks A, Glöckner FO, Amann R, Widdel F, Kube M, Reinhardt R,
Kahnt J, Böcher R, Thauer RK, Shima S. 2003. A conspicuous nickel protein in
microbial mats that oxidize methane anaerobically. Nature 426:878-881.
71. Dalsgaard T, Stewart FJ, Thamdrup B, De Brabandere L, Revsbech NP, Ulloa O,
Canfield DE, DeLong EF. 2014. Oxygen at Nanomolar Levels Reversibly Suppresses
Process Rates and Gene Expression in Anammox and Denitrification in the Oxygen
Minimum Zone off Northern Chile. mBio 5:e01966-14-e01966-14.

72. Stolper DA, Revsbech NP, Canfield DE. 2010. Aerobic growth at nanomolar oxygen
concentrations. Proceedings of the National Academy of Sciences 107:18755-18760.
73. Eggleston EM, Lee DY, Owens MS, Cornwell JC, Crump BC, Hewson I. 2015. Key
respiratory genes elucidate bacterial community respiration in a seasonally anoxic
estuary. Environmental Microbiology 17:2306-2318.
74. King GM, Smith CB, Tolar B, Hollibaugh JT. 2013. Analysis of composition and
structure of coastal to mesopelagic bacterioplankton communities in the northern gulf of
Mexico. Frontiers in Microbiology 3:doi: 10.3389/fmicb.2012.00438.
75. Tolar BB, King GM, Hollibaugh JT. 2013. An analysis of thaumarchaeota populations
from the northern gulf of Mexico. Frontiers in Microbiology 4:72.
76. Mason OU, Hazen TC, Borglin S, Chain PSG, Dubinsky EA, Fortney JL, Han J, Holman
H-YN, Hultman J, Lamendella R, Mackelprang R, Malfatti S, Tom LM, Tringe SG,
Woyke T, Zhou J, Rubin EM, Jansson JK. 2012. Metagenome, metatranscriptome and
single-cell sequencing reveal microbial response to Deepwater Horizon oil spill. The
ISME Journal 6:1715-1727.
77. Dick GJ, Andersson AF, Baker BJ, Simmons SL, Thomas BC, Yelton AP, Banfield JF.
2009. Community-wide analysis of microbial genome sequence signatures. Genome
Biology 10:R85.
78. Sharon I, Morowitz MJ, Thomas BC, Costello EK, Relman DA, Banfield JF. 2013. Time
series community genomics analysis reveals rapid shifts in bacterial species, strains, and
phage during infant gut colonization. Genome Reserach 23:111-120.
79. Peng Y, Leung HCM, Yiu SM, Chin FYL. 2012. IDBA-UD: a de novo assembler for
single-cell and metagenomic sequencing data with highly uneven depth. Bioinformatics
28:1420-1428.
80. Li H, Durbin R. 2009. Fast and accurate short read alignment with Burrows–Wheeler
transform. Bioinformatics 25:1754-1760.
81. Sorek R, Zhu Y, Creevey CJ, Francino MP, Bork P, Rubin EM. 2007. Genome-wide
experimental determination of barriers to horizontal gene transfer. Science 318:1449-52.
82. Darling AE, Jospin G, Lowe E, Matsen Iv FA, Bik HM, Eisen JA. 2014. PhyloSift:
phylogenetic analysis of genomes and metagenomes. PeerJ 2:e243.
83. Edgar RC. 2004. MUSCLE: multiple sequence alignment with high accuracy and high
throughput. Nucleic Acids Research 32:1792-1797.
84. Criscuolo A, Gribaldo S. 2010. BMGE (Block Mapping and Gathering with Entropy): a
new software for selection of phylogenetic informative regions from multiple sequence
alignments. BMC Evol Biol 10:210.
85. Stamatakis A. 2014. RAxML version 8: a tool for phylogenetic analysis and post-analysis
of large phylogenies. Bioinformatics 30:1312-1313.
86. Hug LA, Baker BJ, Anantharaman K, Brown CT, Probst AJ, Castelle CJ, Butterfield CN,
Hernsdorf AW, Amano Y, Ise K, Suzuki Y, Dudek N, Relman DA, Finstad KM,
Amundson R, Thomas BC, Banfield JF. 2016. A new view of the tree of life. Nature
Microbiology 1:16048.
87. Contreras-Moreira B, Vinuesa P. 2013. GET_HOMOLOGUES, a versatile software
package for scalable and robust microbial pangenome analysis. Applied and
Environmental Microbiology 79:7696-7701.
88. Castresana J. 2000. Selection of conserved blocks from multiple alignments for their use
in phylogenetic analysis. Molecular Biology and Evolution 17:540-552.

89.   Price MN, Dehal PS, Arkin AP. 2010. FastTree 2--approximately maximum-likelihood
1068   trees for large alignments. PLOS ONE 5:e9490.
1069   90.   Camacho C, Coulouris G, Avagyan V, Ma N, Papadopoulos J, Bealer K, Madden TL.
1070   2009. BLAST+: architecture and applications. BMC Bioinformatics 10:421.
1071   91.   Junier T, Zdobnov EM. 2010. The Newick utilities: high-throughput phylogenetic tree
1072   processing in the Unix shell. Bioinformatics 26:1669-1670.
1073   92.   Markowitz VM, Chen IMA, Palaniappan K, Chu K, Szeto E, Pillay M, Ratner A, Huang
1074   J, Woyke T, Huntemann M, Anderson I, Billis K, Varghese N, Mavromatis K, Pati A,
1075   Ivanova NN, Kyrpides NC. 2014. IMG 4 version of the integrated microbial genomes
1076   comparative analysis system. Nucleic Acids Research 42:D560-7.
1077   93.   Saier MH, Reddy VS, Tsu BV, Ahmed M, Li C, Moreno-Hagelsieb G. 2016. The
1078   Transporter Classification Database (TCDB): recent advances. Nucleic Acids Research
1079   44:D372-D379.
1080   94.   Lombard V, Ramulu H, Drula E, Coutinho PM, Henrissat B. 2014. The carbohydrate-
1081   active enzymes database (CAZy) in 2013. Nucleic Acids Research 42:D490-D495.
1082   95.   Mortazavi A, Williams BA, McCue K, Schaeffer L, Wold B. 2008. Mapping and
1083   quantifying mammalian transcriptomes by RNA-Seq. Nature Methods 5:621-628.
1084   96.   Quinlan AR, Hall IM. 2010. BEDTools: a flexible suite of utilities for comparing
1085   genomic features. Bioinformatics 26:841-842.
1086   97.   Yachdav G, Wilzbach S, Rauscher B, Sheridan R, Sillitoe I, Procter J, Lewis SE, Rost B,
1087   Goldberg T. 2016. MSAViewer: interactive JavaScript visualization of multiple sequence
1088   alignments. Bioinformatics 32:3501-3503.
1089   98.   Eddy SR. 2011. Accelerated Profile HMM Searches. PLOS Computational Biology
1090   7:e1002195.
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Figure 1

Figure 2

Figure 3

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