



Metagenomes from 25 Low-Abundance Microbes in a Partial Nitrification Anammox Microbiome

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ABSTRACT Microbial communities using anammox bacteria to remove nitrogen are increasingly important in wastewater treatment. We report on 25 metagenome-assembled genomes of low-abundance microbes from a partial nitrification anammox bioreactor system that have not been described previously. These data add to the body of information about this important wastewater treatment system.

Processes utilizing anammox bacteria for nitrogen removal have become popular for energy-saving wastewater treatment since they do not require oxygen or organic carbon for growth to convert ammonium (NH_4^+) and nitrite (NO_2^-) directly to nitrogen gas (N_2) (1–7). Microorganisms in partial nitrification anammox (PNA) bioreactors adapt to repetitive and rapidly changing microaerobic and anoxic environments (5). The most abundant and active functional groups involved in nitrogen cycling in PNA systems are anammox bacteria, ammonia-oxidizing bacteria, and nitrite-oxidizing bacteria, and some of the strains appear to be ubiquitous (8, 9); we are beginning to realize the extent of the metabolic versatility and interactions within these microbiomes (8–10). We reported on metagenome-assembled genomes (MAGs) of the most abundant and active microorganisms in a laboratory-scale PNA bioreactor treating reject water from a full-scale struvite recovery process (11). There, the bioreactor was inoculated with biomass from the York River Treatment Plant (Hampton Roads Sewerage District, Seaford, VA), which uses a PNA deammonification process to treat reject water from the solids dewatering facility (12).

This announcement includes 25 low-abundance MAGs with greater than 68% completion, grouped into 16 clusters, that were recovered from the same laboratory-scale PNA bioreactor and for which a specific role in the community has not been described (Table 1). These additional MAGs add to the expanding body of knowledge about microorganisms present in deammonification bioreactors at low abundance (e.g., each less than 1% based on DNA sequencing coverage and collectively less than 5%, determined as described previously [11, 13]).

Genomic DNA was extracted at multiple time points during operation of the bioreactor (days 77, 231, 350, and 454) using a modified phenol-chloroform method (11). The quality of the isolated DNA was determined using a Qubit 4 fluorometer (Thermo Fisher Scientific, MA, USA), a NanoDrop 2000 spectrophotometer (Thermo Fisher Scientific), and gel electrophoresis. Sequencing libraries were prepared using the TruSeq DNA PCR-free kit (Illumina, CA, USA) following the standard protocol as described (14), samples were sequenced on the HiSeq 2500 platform (Illumina) at the University of Wisconsin-Madison Biotechnology Center using 2×250 -bp reads, and low-quality reads (quality scores of less than 20 and sequence lengths of less than 100 bp) were removed using Sickle (15). A total of 18,106,239 reads

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TABLE 1 MAG statistics and genome accession numbers

| MAG identification | GenBank accession no. | GTDB taxonomy | Completeness (%) | Contamination (%) | Genome size (bp) | No. of scaffolds | <i>N</i> ₅₀ (bp) | GC content (%) | Sequencing depth (×) | No. of tRNAs | No. of 5S rRNAs | No. of 16S rRNAs | No. of 23S rRNAs |
|--------------------|-----------------------|-------------------------------|------------------|-------------------|------------------|------------------|-----------------------------|----------------|----------------------|--------------|-----------------|------------------|------------------|
| BCT_39 | JAJVIC0000000000 | Bacteroidia bacterium | 99.52 | 3.02 | 4,874,625 | 55 | 472,994 | 42.53 | 87 | 45 | 2 | 2 | 2 |
| BAC_79R | JAJVHK0000000000 | Bacterium | 98.88 | 4.49 | 5,858,271 | 32 | 291,821 | 61.52 | 15 | 52 | 0 | 0 | 1 |
| GAM_9R | JAJVIO0000000000 | Gammaproteobacteria bacterium | 97.62 | 2.41 | 3,557,069 | 39 | 193,302 | 66.55 | 32 | 49 | 1 | 0 | 0 |
| BAC_63R | JAJVIF0000000000 | Bacterium | 97.22 | 0 | 3,260,764 | 4 | 1,880,758 | 62.82 | 17 | 48 | 1 | 1 | 1 |
| TMB_89R | JAJVIN0000000000 | Thermoanaerobaculla bacterium | 97.01 | 0.85 | 6,170,929 | 69 | 148,422 | 65.82 | 16 | 48 | 1 | 1 | 1 |
| BAC_38 | JAJVIB0000000000 | Bacterium | 95.54 | 1.1 | 7,631,068 | 45 | 347,644 | 58.95 | 126 | 48 | 1 | 1 | 1 |
| ELM_99 | JAJVIP0000000000 | Elusimicrobia bacterium | 95.51 | 0 | 3,644,178 | 43 | 213,845 | 48.71 | 30 | 76 | 1 | 1 | 1 |
| PCS_19R | JAJVHX0000000000 | Physcisphaerae bacterium | 95.45 | 1.14 | 3,568,264 | 28 | 220,153 | 55.42 | 21 | 48 | 0 | 1 | 1 |
| OMN_62 | JAJVIE0000000000 | Omnitrophota bacterium | 95.16 | 0 | 2,713,171 | 25 | 1,594,128 | 61.88 | 33 | 48 | 1 | 1 | 1 |
| PCS_14R | JAJVHU0000000000 | Physcisphaerae bacterium | 94.89 | 1.14 | 4,668,787 | 70 | 104,852 | 65.74 | 52 | 49 | 0 | 1 | 1 |
| VRM_23R | JAJVHY0000000000 | Verrucomicrobiae bacterium | 94.59 | 3.07 | 4,819,195 | 36 | 167,251 | 59.69 | 18 | 50 | 1 | 1 | 1 |
| PCS_76R | JAJVHI0000000000 | Physcisphaerae bacterium | 94.32 | 0.57 | 5,487,514 | 72 | 107,025 | 67.82 | 16 | 54 | 1 | 1 | 1 |
| PCS_15R | JAJVHO0000000000 | Physcisphaerae bacterium | 93.18 | 1.7 | 5,249,578 | 50 | 159,786 | 67.13 | 59 | 61 | 0 | 1 | 1 |
| ANL_81R | JAJVHM0000000000 | Anaerolineae bacterium | 92.73 | 2.18 | 7,867,783 | 60 | 252,904 | 57.36 | 13 | 48 | 1 | 1 | 1 |
| PCL_122_1R | JAJVHS0000000000 | Physcisphaerales bacterium | 91.16 | 0.86 | 4,404,079 | 104 | 58,562 | 67.83 | 10 | 0 | 0 | 0 | 0 |
| PSM_69 | JAJVIG0000000000 | Pseudomonadales bacterium | 89.29 | 3.48 | 3,834,276 | 66 | 84,489 | 68.46 | 29 | 46 | 0 | 0 | 0 |
| GAM_33 | JAJVHZ0000000000 | Gammaproteobacteria bacterium | 89.03 | 0.65 | 4,271,281 | 128 | 64,536 | 64.03 | 45 | 52 | 1 | 0 | 0 |
| MYX_71R | JAJVHO0000000000 | Mycococcota bacterium | 88.03 | 0.21 | 3,968,878 | 78 | 77,580 | 63.12 | 11 | 44 | 1 | 1 | 0 |
| RDC_77 | JAJVJ0000000000 | Rhodocyclaceae bacterium | 85.88 | 1.14 | 2,815,254 | 67 | 57,548 | 64.68 | 18 | 39 | 0 | 0 | 0 |
| PLM_90R | JAJVID0000000000 | Planctomycetota bacterium | 85.47 | 1.71 | 6,468,668 | 221 | 61,976 | 64.89 | 9 | 65 | 0 | 1 | 1 |
| PYM_42R | JAJVIA0000000000 | Pyrinomonadales bacterium | 77.45 | 0.91 | 3,991,433 | 141 | 38,096 | 57.8 | 90 | 72 | 1 | 1 | 1 |
| ANL_34 | JAJVHT0000000000 | Anaerolineae bacterium | 76.31 | 0 | 4,595,932 | 231 | 33,931 | 56.33 | 49 | 38 | 0 | 0 | 0 |
| BAC_13 | JAJVHW0000000000 | Bacterium | 75.57 | 2.84 | 3,864,190 | 174 | 22,649 | 53.05 | 44 | 36 | 1 | 1 | 1 |
| PLM_16R | JAJVIL0000000000 | Planctomycetota bacterium | 69.09 | 0.29 | 2,770,384 | 154 | 25,696 | 72.1 | 42 | 51 | 1 | 1 | 1 |
| BRK_7R | JAJVIL0000000000 | Burkholderiaceae bacterium | | | | | 18,455 | 69.49 | 18 | 39 | 0 | 0 | 0 |

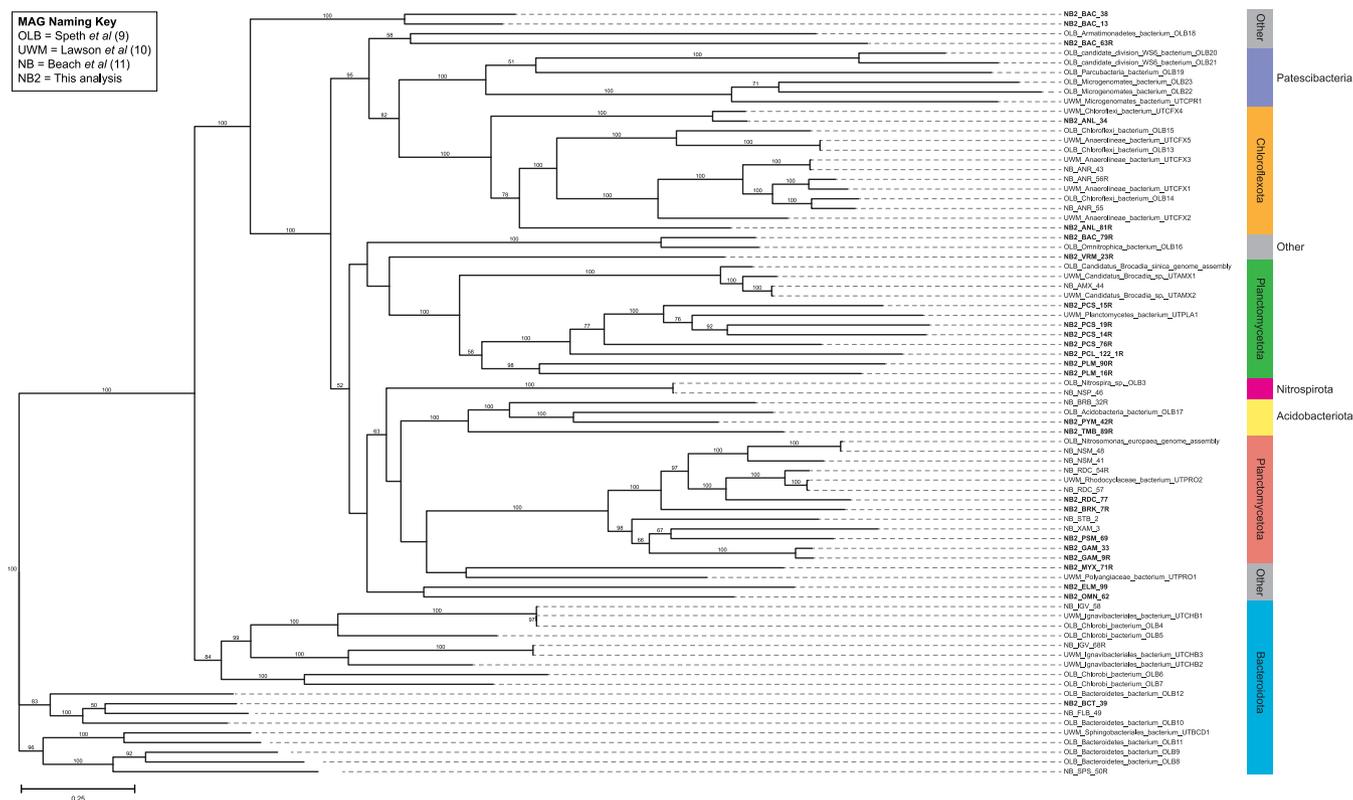


FIG 1 Phylogenetic tree of the 25 MAGs from this study (labeled NB2 and bolded), compared with the 16 most abundant MAGs from Beach et al. from the same microbiome (labeled NB) (11), as well as MAGs from Speth et al. (labeled OLB) (8) and Lawson et al. (labeled UWM) (9). The tree was constructed with RAxML-NG (unrooted) (25) with the 120 bacterial single-copy marker genes from GTDB-Tk (24) and visualized with TreeViewer (<https://treeviewer.org>). GTDB phylum taxonomy is listed on the right. Bootstrap values of more than 50 are shown. The scale bar indicates the number of nucleotide substitutions per sequence site. Strain codes for NB and NB2 strains are as follows: AMX, *Brocadia*; ANL, *Anaerolineae*; ANR, *Anaerolineales*; BAC, bacterium; BCT, *Bacteroidia*; BRB, *Bryobacteraceae*; BRK, *Burkholderiaceae*; ELM, *Elusimicrobia*; FLB, *Flavobacteriales*; GAM, *Gammaproteobacteria*; IGV, *Ignavibacteria*; MYX, *Myxococcota*; NSM, *Nitrosomonas*; NSP, *Nitrospira*; OMN, *Omnitrophota*; PCL, *Phycisphaerae*; PCS, *Phycisphaerae*; PLM, *Planctomycetota*; PSM, *Pseudomonadales*; PYM, *Pyrinomonadales*; RDC, *Rhodocyclaceae*; SPS, *Saprosiraceae*; STB, *Steroidbacteraceae*; TMB, *Thermoanaerobaculia*; VRM, *Verrucomicrobiae*; XAM, *Xanthomonadales*.

from the metagenomic samples were processed with default parameters as described previously (11), using SPAdes v3.3.0 (16) for assembly, Anvi'o v5.5.0 (17) for binning, BMap v38.22 (18) for mapping, Prodigal v2.6.3 (19), HMMER v3.2.1 (20), and NCBI Clusters of Orthologous Groups (COGs) (21) for annotation, and ProDeGe v2.2 (22) and tetranucleotide frequency comparisons for further cleaning of the MAGs. Genome statistics were determined with CheckM v1.0.3 (23), and taxonomy was assigned with GTDB-Tk v1.5.1 (database release 202) (24). The unrooted phylogenetic tree (Fig. 1) comparing these MAGs with MAGs from previous studies (8, 9) was generated using RAxML-NG v0.9.0 (25) and visualized with TreeViewer (<https://treeviewer.org>). These MAGs add to our understanding of PNA bioreactors and may aid in optimization of these systems once the function of these low-abundant microbes is elucidated.

Data availability. Raw metagenomic sequence data and MAGs are available in NCBI GenBank under BioProject accession number [PRJNA559529](https://www.ncbi.nlm.nih.gov/bioproject/PRJNA559529). All custom scripts are available at GitHub (https://github.com/GLBRC/metagenome_analysis).

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