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Insights Gained into Activated Sludge Nitrification through Structural and Functional Profiling of Microbial Community Response to Starvation Stress

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2 Functional Profiling of Microbial Community Response to Starvation Stress

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5 Water Impact Statement:

- 6 An improved understanding of nitrifying microbial communities in wastewater treatment
- 7 is imperative for proper design and operation of biological nutrient removal systems.
- 8 Here we investigated nitrifying microbial community shifts in a starvation activated
- 9 sludge reactor at various temperatures. *Nitrospira* was the only nitrifying genus detected
- 10 consistently, and phylogenetic analysis showed close relation of the most abundant

11 *Nitrospira* to comammox and mixotrophic organisms.

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2	Functional Profiling of Microbial Community Response to Starvation Stress
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21 Abstract:

22 Although nitrification is widely applied for nitrogen removal in wastewater 23 treatment plants (WWTPs), more information about the microorganisms involved and 24 their corresponding capabilities and limitations is critical to refine kinetic parameters and 25 process design to optimize advanced nutrient removal. Here we carried out a series of 26 ammonia starvation stress experiments and applied a suite of assays to characterize the 27 microbial community response. Illumina sequencing was applied to both DNA and RNA-28 derived (i.e., cDNA) 16S rRNA amplicons to differentiate responses of functionally-29 active bacteria, with gene markers corresponding to known nitrifiers and compared to 30 those targeted via quantitative polymerase chain reaction (qPCR). As expected, total 31 bacterial DNA (i.e., 16S rRNA genes) and nitrifier activity potential decayed over the 32 course of the 18-day starvation period. *Nitrospira* was the only known nitrifying genera 33 consistently detected via 16S rRNA amplicon sequencing in all samples. Despite 34 relatively deep DNA sequencing (rarified to 38,000 sequences per sample), *Nitrosomonas* 35 was the only other known nitrifying genera detected (4 of 84 samples), although 36 *Nitrobacter* was detected via qPCR. Relative abundance of *Nitrospira* DNA and cDNA 37 remained relatively constant throughout the starvation experiments and did not vary with 38 temperature. Two *Nitrospira* OTUs were by far most dominant and were most closely 39 related to known Nitrospira capable of mixotrophy and comammox. Thus, observed 40 persistence of *Nitrospira* through starvation may be a function of diverse metabolic 41 capability. Recognition that bacteria of diverse metabolic capability can drive nitrogen 42 removal is critical to advance accurate modeling and design of advanced nutrient removal 43 processes.

45 Keywords: Nitrification, comammox, *Nitrospira*, nitrifying activated sludge, starvation,
46 stress

47 Introduction:

48 Variation from Canonical Model of Nitrification

49 Nitrification is a critical process for nitrogen removal in wastewater treatment 50 plants (WWTPs). Although widely applied, surprisingly little is known about the actual 51 microorganisms involved in nitrification. The canonical assumption has historically been 52 that nitrification is exclusively driven by a two-step process in which ammonia is 53 oxidized to nitrite by ammonia oxidizing bacteria (AOB) and subsequently nitrite is 54 oxidized to nitrate by nitrite oxidizing bacteria (NOB). However, challenges to this 55 paradigm have recently emerged.¹ For example, anaerobic ammonia oxidizing 56 (anammox) bacteria use nitrite as an electron acceptor and ammonia as an electron donor. Anammox bacteria were first discovered in 1995^{2,3} and now are already being 57 58 implemented for the first time in full-scale WWTPs.⁴ Nitrifiers also are not exclusive to 59 the bacterial domain, as illustrated by the discovery of ammonia oxidizing archaea 60 (AOA), which are now recognized to sometimes be the dominant nitrifiers in activated sludge.⁵ More recently, the comammox process has been described, in which a single 61 62 microorganism belonging to the genus *Nitrospira* completely oxidizes ammonia to nitrate.⁶⁻⁹ A recent metagenomic survey of 16 full scale biological nutrient removal 63 WWTPs found comammox bacteria in all systems.¹⁰ 64

65 While our understanding of the bacteria involved in the nitrogen cycle is rapidly 66 evolving, it remains unclear the extent to which various processes are the major drivers of

nitrification in real-world, typical nitrifying activated sludge systems or if they are
important only under specialized circumstances. Importantly, widely implemented
WWTP design parameters, such as growth and decay rate constants, were determined
long before these newer discoveries and do not take into account such deviations from the
canonical model of nitrification in activated sludge.¹¹

72 Nitrogen removal processes, in particular, tend to require precise manipulation of 73 conditions to achieve desired performance. This is especially true in the case of short-cut 74 denitrification, in which it is desired to stop nitrification at nitrite, which is subsequently 75 directly denitrified. While short-cut denitrification is highly attractive from the 76 standpoint of minimizing aeration and supplemental carbon costs, it is very challenging 77 to implement in practice. In an effort to limit nitrite oxidation, environmental factors such as DO, pH, ammonia and nitrite concentrations must be carefully controlled in order to 78 select for ideal nitrifying microbial communities that exclude NOB.¹²⁻¹⁴ However, any 79 80 deviation from the canonical model of nitrification presents a fundamental obstacle to 81 implementation. In particular, if the dominant organisms are capable of carrying 82 ammonia oxidation all the way through to nitrate, as is in the case of comammox, then it 83 will be impossible to out-select only the bacteria that oxidize nitrite.

84 Stress Response and Nitrifying Microbial Communities

Increased knowledge of the dominant nitrifiers and their behavior in wastewater treatment will inherently inform and benefit process design and optimization. To this end, stress tests can provide a useful tool not only for assessing the resilience of a microbial community in the face of reactor upset, but also for informing strategies for selecting a microbial community composition that is capable of performing the desired

90 process. Nitrification is known to be very sensitive to several types of stress, including variations in salinity,¹⁵ temperature,¹⁶ DO,¹⁷ and substrate concentration.¹⁸ Reactor upsets 91 92 due to influent variation and environmental conditions are a common and often expensive 93 nuisance for WWTP management.¹⁹ Nitrifying bacteria cope with stress using various 94 mechanisms. For example, prior studies of responses to low substrate concentration in 95 pure and mixed culture lab scale reactors identified increased intracellular ammonia concentrations,²⁰ dormant or resting states,^{21,22} and elevating ribosomal content.²³ In 96 97 response to light or acetylene exposure to inactivate ammonia monooxygenase (*amo*), the 98 enzyme responsible for ammonia oxidation, it was found that *Nitrosomonas europaea* 99 maintained multiple copies of *amo* with varying ammonia activity allowing recovery in various substrate concentrations.²⁴ Variations in stress response mechanisms can lead to 100 101 selective advantages for certain nitrifying microorganisms over others and thus influence 102 key process parameters, such as activity and decay rates. Indeed, varying operational 103 parameters have been shown to impact nitrifying communities, with *Nitrosomonas* and 104 Nitrobacter thriving in an alternating aerobic/anaerobic reactor and Nitrosospira and *Nitrospira* dominating in strictly aerobic reactors.¹² Also, some *Nitrospira* and 105 *Nitrobacter* species are able to oxidize formate as an alternative to nitrite,²⁵⁻²⁸ making 106 107 them more resilient to nitrite limitation. With improved understanding of the impacts of 108 stresses to nitrifying microbial communities, it may be possible to minimize nitrification 109 reactor upsets and improve recovery after reactor upset or dormancy by selecting for 110 nitrifying microorganisms best suited to handle stresses in a given wastewater. 111 Recently, we sought to refine estimates of AOB and NOB decay parameters 112 through modeling the decline in nitrate production rate (NPR) and NO_x production rate

113 (NO_xPR) during ammonia starvation of nitrifying activated sludge.¹¹ While apparent 114 AOB decay rates were similar to those reported in the literature, NOB decay rates were 115 much slower. This is one potential reason why out-selecting NOBs for processes such as 116 short-cut denitrification has been found to be challenging to implement in domestic wastewater treatment.²⁹ Interestingly, we further observed that the decay rates estimated 117 118 based on disappearance of gene markers (i.e., DNA) of known nitrifiers was very low 119 compared to the observed rates of activity decay. Such work highlights the importance of 120 understanding exactly who the dominant nitrifying microorganisms are in a given WWTP 121 and their various metabolic capabilities and constraints.

The purpose of this study was to gain insight into the structure and function of 122 123 typical nitrifying activated sludge communities through a series of ammonia starvation 124 experiments. Starvation is typically used to assess decay rates of nitrifying bacteria, and 125 therefore is an important stress response to investigate as it impacts key reactor design 126 parameters. We used the Washington D.C. Blue Plains Advanced WWTP, one of the 127 largest in operation in the world, as a source of stable and high-performing nitrifying 128 activated sludge. The sludge was assayed over a period of several months and multiple 129 seasons, with starvation trials conducted over a range of operating temperatures (14-30 130 °C). We applied a combination of nitrifier activity potential assays with molecular 131 profiling to identify shifts in nitrifying microorganisms with time and temperature within 132 the context of this real-world, complex activated sludge community. Illumina sequencing 133 was applied to both DNA and RNA-derived (i.e., cDNA) 16S rRNA amplicons to 134 differentiate responses of functionally-active bacteria, and gene markers corresponding to 135 known nitrifiers were compared to those targeted via qPCR. Improving our understanding

of the dominant nitrifiers active in activated sludge WWTPs; especially in terms of their metabolic capabilities and constraints, response to stress, resilience, and upset recovery patterns, is critical to future optimization of design and operation of advanced nutrient removal processes.

140 Materials and Methods:

141 Experimental Set-Up

142 Mixed liquor suspended solids (MLSS) was collected from nitrification basin 143 effluent at DC Water Blue Plains Advanced WWTP (Figure 1) and immediately 144 transported to the laboratory at Virginia Tech (Blacksburg, VA ~5 hours) with aeration 145 maintained using an aquarium-pump. The MLSS was transferred to a 25-L Starvation 146 Reactor and maintained in a constant temperature room with aeration controlled by a HACH SC100 sensor controller coupled with an LDOTM dissolved oxygen (DO) sensor 147 148 (HACH, Loveland, CO) to maintain DO between 2.5 and 4 mg/L. pH was manually 149 adjusted within a range of 7.2-7.8 using NaHCO₃ throughout the experiment. Samples 150 were collected from the Starvation Reactor for DNA and RNA extraction and 151 NPR/NO_xPR assays. Logistically, technical replicates of Starvation Reactors were not 152 possible so the experimental design instead emphasized biological repeatability with 153 different seed sludges, along with technical duplicates of the NPR/NoxPR assays.



Figure 1 Schematic overview of "Starvation" and "Recovery" Trials. First three Starvation trials (14 °C Trial 1, 20 °C Trial 1, and 30 °C Trial 1) included analysis of WWTP basin effluent MLSS along with samples collected during the starvation period, while Recovery trials (14 °C Trial 2, 20 °C Trial 2, and 27 °C Trial 1) all included WWTP basin influent MLSS, effluent MLSS, and Day 0 sampling, along with samples collected during the Recovery period.

162 Initial "Starvation" trials were performed at three temperatures typical of seasonal 163 cycles experienced in temperate regions (14 °C Trial 1, 20 °C Trial 1, and 30 °C Trial 1) 164 and were run for 16-18 days. To observe recovery of nitrifiers following starvation 165 experiments, follow-up "Recovery" trials were conducted at similar temperatures (14 °C 166 Trial 2, 20 °C Trial 2, and 27 °C Trial 1) in which ammonium was re-introduced into the 167 Starvation Reactors and maintained between 20-50 mg/L NH₃-N at the conclusion of the 168 Starvation trials. At the start of the starvation trials for these Recovery trials, "Influent" of 169 the DC Water nitrification basin was also sampled (Figure 1). These samples were taken 170 in order to compare the succession of microbial community composition in the laboratory 171 to that experienced during the conditions of the sequential nitrification aeration basins in 172 the field. This comparison of laboratory and field microbial community successions 173 would likely be impacted by the difference in scale of the reactors. The seed sludge was 174 also sampled directly from the basin effluent in the Recovery trials and is denoted as "Day 0" (Figure 1). While these Recovery trials were run at similar temperatures as the 175

176 former Starvation trials, it should be noted that these are not technical replicates, but are 177 biological replicates as a new seed sludge was used to start the Starvation reactor in every 178 trial. See Table S1 for sludge collection dates and details on sampling frequency for each 179 trial.

180 N

Nitrifier Activity Potential Assays

181 After a \approx 36 h acclimation time, 1,300 mL of MLSS was withdrawn from the main 182 Starvation Reactor and transferred to aerated batch reactors to conduct nitrifier activity 183 potential assays. (NH₄)₂SO₄ and NaNO₂ were dosed simultaneously to the batch reactors. 184 in duplicate, to reach an initial concentration of ≈ 50 and ≈ 15 mg-N/L, respectively. DO 185 was maintained at 2 to 4 mg/L during the batch tests by adjusting the air flow to aeration 186 stones using aquarium aerators with controlling valves. pH was also monitored and 187 maintained between 7.2 to 7.9 through addition of NaHCO₃, as needed. Aqueous samples 188 for analysis of ammonia, nitrite, and nitrate were collected every 20 min for 2 hours and 189 filtered through 0.45 µm VWR Polyethersulfone membrane filters (Arlington Heights, 190 IL). Nitrate was measured using a Dionex (Sunnyvale, CA) DX-120 ion chromatography system (IC) according to Standard Method 4110.³⁰ Ammonia and nitrite were measured 191 192 using the HACH Salicylate Method (TNTplus 832) and HACH Diazotiaztion method 193 (TNTplus 840), respectively. Nitrate concentration was plotted with respect to time and 194 the linear line of best fit was calculated – the slope of which was the NPR. Similarly, the 195 sum of nitrate and nitrite concentrations were plotted with respect to time, and the line of 196 best fit for this plot was the NO_xPR. The assays were repeated over seven time points 197 throughout the Starvation trials. In the Recovery trials, nitrifying microorganism activity 198 was monitored following the Starvation trials, after dosing (NH₄)₂SO₄ and NaNO₂ into

199	the Starvation Reactor at a concentration of \approx 50 and \approx 15 mg-N/L, respectively.
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- 200 Ammonia, nitrite and nitrate were measured every 20 min for 2 hours.
- 201 DNA and RNA Extraction

202 MLSS samples for DNA and RNA extraction were preserved with RNAlater®

203 solution (Life Technologies, Grand Island, NY) upon sampling, flash frozen using a dry

204 ice-ethanol bath, and stored at -80 °C prior to further processing. DNA was extracted in

205 duplicate from 175 µL of preserved sludge using a FastDNA® Spin kit for soil (MP

206 Biomedicals, Solon, OH) according to the manufacturer's protocol. RNA was also

207 extracted in duplicate from 175 μL of sample using a MagMAX® Total Nucleic Acid

208 Isolation Kit (Life technologies, Grand Island, NY), DNA was removed using a TURBO

209 DNA-freeTM Kit (Life Technologies Corporation, Grand Island, NY) and reverse

210 transcription performed using an iScriptTM cDNA synthesis kit (Bio-Rad, Hercules, CA)

211 on the same day according to the manufacturer's instructions.

212 Quantitative Polymerase Chain Reaction

AOB, as represented by the bacterial ammonia monooxygenase gene (*amoA*), along with *Nitrospira*, *Nitrobacter*, and total bacterial 16S rRNA genes were enumerated by qPCR using previously published methods.³¹ Reactions were performed in technical triplicate in 10 μ L volumes containing 1×SsoFast® Probes or Evagreen® supermix (Bio-Rad, Hercules, CA), 400 nM primers, and 1 μ L template. Samples were analyzed at 1:10 or 1:50 dilutions to minimize qPCR inhibition based on dilution curves.

219 16S rRNA Gene Amplicon Sequencing and Analysis

220	A cross-section of samples was selected from the experimental trials (see Table
221	S1 for sample selection) and subjected to 16S rRNA amplicon sequencing. The V4 region
222	of the 16S rRNA gene from DNA and cDNA samples was subject to PCR amplification
223	using universal bacterial/archaeal barcoded primers 515F/806R ³² following the Earth
224	microbiome protocol (http://www.earthmicrobiome.org). For QA/QC purposes, a DNA
225	extraction blank, cDNA extraction blank, and PCR blank were subjected to sample
226	processing in parallel and included in the sequencing pools. Two sets of pooled samples
227	were sequenced by the Biocomplexity Institute's Genomics Research Laboratory
228	(Blacksburg, Virginia) using a paired-end 250 Illumina MiSeq platform, one pool
229	included initial Starvation trials (14 °C Trial 1, 20 °C Trial 1, and 30 °C Trial 1) and the
230	second included follow-up Recovery trials (14 °C Trial 2, 20 °C Trial 2, and 27 °C Trial
231	1) as well as blank controls. Paired-end sequence reads were joined using PANDAseq ³³
232	and then processed using Quantitative Insights Into Microbial Ecology (QIIME) version
233	1.8.0. ³⁴ The <i>pick_de_novo_otus.py</i> script was used to perform <i>de novo</i> operational
234	taxonomic unit (OTU) picking using the uclust_ref method and taxonomy assignments of
235	OTUs were generated using Greengenes 13_8 reference database, aligned at 97%. The
236	resulting OTU table was filtered to remove chimeric sequences using the Chimera Slayer
237	method, ³⁵ and singletons and organelle sequences using the
238	<i>filter_taxa_from_otu_table.py</i> script. This procedure produced an OTU table consisting
239	of 12 million sequences from 84 samples and 3 blank controls. The DNA extraction
240	blank, cDNA extraction blank, and PCR blank had 9,135 sequences, 14,109 sequences,
241	and 24,136 sequences respectively. These sequence counts were much lower than the

242	samples in the second DNA pool sequences (sample with lowest sequence count of
243	71,924). Blank control sequences mainly consisted of OTUs identified as Proteobacteria
244	(76-94%) which is consistent with normal reagent contamination of extraction kits. ³⁶ The
245	lowest number of sequences in a sample was 38,561 and therefore the OTU table was
246	rarefied to 38,000 sequences per sample for all analyses, unless otherwise stated.
247	Similarity of microbial community compositions was quantified using weighted UniFrac
248	distances computed using the Greengenes reference tree within QIIME. ³⁷ A weighted
249	UniFrac distance matrix was exported to Primer 6 software (Primer-E, Plymouth, UK),
250	where multidimensional scaling (MDS) ordination was conducted. A phylogenetic tree of
251	Nitrospira OTUs was created using the maximum likelihood method, bootstrapped 100
252	times in MEGA7 software ³⁸ and plotted using FigTree software version v1.4.3
253	(tree.bio.ed.ac.uk/software/). Known Nitrospira 16S rRNA sequences were included as
254	reference sequences and the tree was rooted using Leptospirillum ferrodiazotrophum, a
255	species within the Nitrospiraceae family, but not in the Nitrospira genus (see Electronic
256	Supporting Information for sequences used).

257 **Results and Discussion**

258 Shifts in AOB and NOB Activity Potential with Starvation and Recovery

Nitrifier activity potential assays provided a means to measure shifts in microbial
 community function, in terms of observed ammonia oxidation and nitrite oxidation rates,

- 261 in response to stress during the Starvation and Recovery trials. Aliquots of MLSS were
- removed, spiked with ammonium sulfate and sodium nitrite, then NO₃⁻ and NO₂⁻
- 263 concentrations were measured with time to determine the NO_xPRs and NPRs,

264	respectively (Figure 2). In this manner, NO _x PR and NPR are gross indicators of AOB and
265	NOB activity potential respectively.
266	Nitrifier activity potential decayed with time during the Starvation trials, as
267	expected, reaching a stable baseline rate across all temperature conditions after about 10
268	days. Baseline NO _x PRs ranged from 13-58 mg-N/L-d while NPRs ranged from 64-134
269	mg-N/L-d by day 16-18. Nitrification rates are known to be temperature sensitive and are
270	generally thought to increase linearly over the range of 14 to 30 °C according to the
271	Arrhenius relationship. ³⁹ However, in the present study, NO _x PRs and NPRs did not trend
272	consistently with temperature under which the Starvation Reactors were incubated
273	(Figure 2). We note that we hypothesized that 30 °C was outside of the linear range
274	expected from the Arrhenius relationship due to the fact that the nitrification rate at 30 $^{\circ}$ C
275	was less than that at 20 °C. For this reason, we reduced the initial temperature to 27 °C
276	for the Recovery trial.

277 Response of Nitrifying Microbial Communities to Starvation and Recovery

During the 16-18 day Starvation period, the Starvation Reactor essentially functioned as an aerobic digester. Nitrifiers were intentionally starved by not providing soluble ammonia, although some low level of ammonia resulting from biomass decay theoretically would remain available. It is important to note that soluble organic carbon



Figure 2 Nitrifying activity potential assays carried out during ammonia Starvation and Recovery trials from 14 0 C – 30 0 C. a) NO_xPR is indicative of AOB activity potential and b) NPR is indicative of NOB activity potential. Error bars indicate standard deviation of duplicate assays, conducted on aliquots of MLSS from Starvation Reactors in batch mode. Replicates assays were not conducted during Recovery trials.

289 was also deprived and therefore shifts in heterotrophic organisms towards those capable 290 of hydrolysis would be expected. Microorganisms that were not adaptive to these 291 conditions might become inhibited, dormant, or die, thus influencing the observed 292 nitrifier activity potential discussed above. In the case of death, the microorganisms may exhibit variable decay rates, depending on amenability of cell envelope to digestion, and 293 294 therefore DNA from dead cells may persist at varying rates. Thus, molecular profiling of 295 microbial communities included both DNA- and RNA-derived profiling, with the latter 296 providing a more direct indicator of which microbes were active. 297 Overall, qPCR of 16S rRNA genes (i.e., DNA) indicated net decay of microbes 298 across Starvation trials, with a total loss of 2.1×10^8 - 5.6×10^9 gene copies/mL over the 16-299 18 days (Figure S1). Interestingly, the 14 °C Trial 1 consistently contained the greatest 300 density of 16S rRNA genes, as well as amoA and Nitrospira gene copies, whereas this 301 trial ranked in the middle in terms of nitrifier activity potential and rate of decline. The 302 20 °C Trial 1 displayed the sharpest decrease in 16S rRNA gene copies (Figure S1) and 14

303	correspondingly yielded the highest initial nitrifier activity potential and also the steepest
304	decline (Figure 2). Figure S2 summarizes the results of a Spearman correlation analysis
305	among NO _x PR and NPR rates and gene copies targeted by qPCR. Ammonia oxidation
306	rates (i.e., NO _x PR) correlated with <i>amoA</i> concentrations (p-value 0.00017), but did not
307	correlate with universal 16S rRNA concentrations (p-value 0.056). Nitrite oxidation rates
308	(i.e., NPR) correlated significantly with both Nitrospira 16S rRNA gene concentrations
309	(p-value 0.017) and universal 16S rRNA concentrations (p-value 0.0015), but did not
310	correlate with Nitrobacter 16S rRNA gene concentrations (p-value 0.27). The tendency
311	for nitrifier activity potential to correlate with gene copy numbers of total bacteria, as
312	well as those corresponding to known AOBs and NOBs suggests that there is a general
313	trend of death of nitrifying organisms through the Starvation trial.
314	
315	Microbial Community Structure and Activity During Starvation and Recovery
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326 3, increasing red shading). Importantly, the initial time points of the Starvation 327 experiments were distinct across different trials (Figure S3). Although technical variance 328 could not be assessed based on this study due to logistical constraints on operating 329 technical replicates of Starvation trials, distinct initial microbial community structure 330 across trials is consistent with distinct trends in terms of nitrifier activity potential as a 331 function of temperature (Figure 2). Even slight differences in microbial community 332 composition could influence the rates at which nitrifiers died, went dormant, or switched 333 metabolism during starvation. Differences between DNA- and RNA-derived analyses 334 indicate that a substantial portion of the DNA-based analysis represents either dead or dormant organisms. 335



336

Figure 3 Comparison of microbial community composition (DNA) and activity (RNA)
 changes with time during Starvation and Recovery trials as a function of temperature based
 on MDS analysis. MDS ordination of Unifrac distance matrix of OTU table were rarified

to 38,000 sequences per sample. a: 14 °C trials b: 20 °C trials c: 27 °C/ 30 °C trials.
Combined plot is available in SI (Figure S3). Intensity of red shading increases with time
in Starvation Reactor, purple points indicate MLSS samples collected from the Influent to
the WWTP nitrification basin. Blue points represent Recovery samples collected after
commencing ammonium feed to the Starvation Reactor.

345

346 Comparison of RNA- versus DNA-based community profiling provided a sense

347 of which microbes were stimulated by ammonia addition during the Recovery trials.

348 According to DNA profiling, the microbial community composition remained relatively

349 similar to the Day 18 condition after ammonia dosing, which is not unexpected during a

350 short 48 hour incubation. However, characterization based on RNA indicated much more

351 pronounced, though highly variable, response. This general trend was consistent across

352 temperatures. The contrast between the RNA during recovery and the RNA from the

353 Starvation period of the experiments was most apparent in 14 °C Trial 2 and 20 °C Trial

354 2 (Figure 2, panels a and b, open blue triangle and open blue circle, respectively).

355 However, in 27 °C Trial 1 experiment (Figure 2, panel c), the RNA shift after

356 commencing ammonia feeding returned to a state that resembled that of the microbial

357 community comprising the "influent" MLSS at the beginning of the WWTP aeration

358 basin (see open upside-down blue triangles).

359 Examining the relative abundances of the various phyla detected by amplicon

360 sequencing provided insight into potential underlying factors driving the trajectories of

361 microbial community shifts through time and the underlying causes of the bifurcation in

362 DNA and RNA-based analyses shown in Figure 3. The microbial communities

363 characterized by DNA were dominated by *Proteobacteria* and *Bacteroidetes*, as is

364 expected in activated sludge.⁴⁰⁻⁴² Notably, a trend of distinct relative abundances among

365 various phyla, as estimated by DNA versus RNA-based analysis, was observed across all

- 366 trials. Specifically, DNA-based measurements consistently indicated increased
- 367 Bacteroidetes and decreased Proteobacteria (Figure 4). RNA-based measurements, on
- 368 the other hand, indicated that *Planctomycetes* and *Cyanobacteria* were greater in relative



Figure 4 Relative abundances of the Phyla detected by 16S rRNA gene amplicon
sequencing in each trial with time. Symbols along x-axis correspond with those in Figure
where intensity of red shading indicates increase in time in Starvation trial, purple points
indicate samples collected from the "influent" of the DC Water nitrification basin, and blue
points represent Recovery samples collected after feeding of ammonium to reactor.

- abundance relative to the DNA dataset, especially at 20 °C and above. Increase in
- 377 Bacteroidetes relative abundance in the DNA dataset during this aerobic experiment was
- 378 surprising, given that most of these organisms are known to be obligate anaerobes.
- 379 Importantly; however, relative abundance of *Bacteroidetes* in the RNA dataset was much
- 380 lower and less variable, suggesting that these organisms were not very active.⁴³ Thus,

high abundance of *Bacteroidetes* probably relates to their dominance in fecal matter, with
perceived "increase" being an artifact of greater resilience to decay relative to other
phyla.

384 In order to identify OTUs driving the observed trends, the variance of the OTU 385 relative abundances within each trial and nucleic acid type (i.e., DNA or RNA) were 386 calculated and ranked (Figure S4). This revealed substantial differences in the ten most 387 variable OTUs based on DNA versus RNA datasets among the trials. For example, in the 388 14 °C trials, Bacteroidetes and Betaproteobacteria OTUs were among the most variable 389 OTUs compared to mainly *Deltaproteobacteria* OTUs being the most variable in the 390 RNA dataset. The trends observed in Figure 3 are supported by the 10 most variable 391 OTUs in each trial for DNA and RNA (Figure S4). In general, most of these OTUs 392 belonged to genera with species associated with organic carbon degradation. It has been 393 shown that influent COD composition and carbon source are strong drivers of activated sludge microbial community composition.⁴⁴ Therefore, the variation in microbial 394 395 community composition during the starvation trials may largely be due to the switch from 396 a soluble to a mostly particulate carbon source. 397 Of particular interest in this study, a Nitrospira OTU was among the top ten most 398 variable OTUs in 5 of the 6 trials for the DNA dataset (all but 14 °C Trial 2), and 4 of the 399 6 trials for the cDNA dataset (all but 20 °C Trial 2 and 27 °C Trial 1). Thus, Nitrospira, a

400 well-recognized genus of nitrifiers, were consistently identified as among the most

401 responsive microbes to the ammonia Starvation and Recovery conditions imposed.

402	Further Identification of Nitrifiers in 16S rRNA Amplicon Sequence Libraries
403	Following taxonomic assignment of OTUs using the Greengenes reference
404	database, the resulting OTU table was mined to identify known nitrifying
405	microorganisms. Genera known to contain AOBs searched included Nitrosomonas,
406	Nitrosospira, Nitrosococcus, Nitrosolobus, and Nitrosovibrio. ⁴⁶ Genera known to contain
407	NOBs searched included Nitrospira, Nitrobacter, Nitrococcus, Nitrospina, Nitrotoga,
408	Nitrolancea, and Nitromaritima. ¹ Since no OTUs were identified as Thaumarchaeota,
409	which is thought to be the only phylum containing AOA, ⁴⁷ it was concluded that archaea
410	did not play a significant role in nitrification in this system. Nitrosomonas were detected
411	in only four samples out of the rarefied dataset (84 samples), and were found at low
412	relative abundance (less than 10 ⁻³) (Figure 5). <i>Nitrospira</i> , however, were detected at high
413	relative abundance in all trials across temperature variations and remained relatively
414	stable throughout starvation experiments, ranging from 1.4% to 11% of the total DNA
415	sequences.

416





Figure 5 Relative abundance of nitrifying genera during Starvation experiments based on Illumina 16S rRNA gene amplicon sequencing. Relative abundance was calculated based on abundance in rarefied OTU table, normalized to total rarefied sequences (38,000 sequences). Influent WWTP nitrifying aeration basin samples were plotted and labeled as a reference, with ammonium dosing initiated at about day 18.5 of the experiment, as indicated on the plots.

426 Quantification of AOB and NOB Gene Markers by qPCR

427 AOB, *Nitrospira*, and *Nitrobacter* were also examined using qPCR to quantify 428 *amo*A as a functional gene marker for known AOB and 16S rRNA gene markers specific 429 to the known NOB, *Nitrospira* and *Nitrobacter* (Figure S1). *Nitrospira* 16S rRNA gene 430 copies were consistently much greater than *amo*A gene copies, averaging 2.0 log higher 431 in absolute abundance (ranging from 0.26-3.9 log). This suggests that canonical AOBs 432 are very low in relative abundance, which would explain absence of all canonical AOB 433 genera, except for four instances of *Nitrosomonas* detection, in the16S rRNA amplicon 434 sequence libraries. On the other hand, 16S rRNA genes corresponding to *Nitrobacter*,
435 traditionally considered to be an NOB, were consistently detected at sufficiently high
436 relative abundance (ranging from 0-23%) to be detected using 16S rRNA amplicon
437 sequencing, but they were not detected in any of the libraries. This suggests that the
438 "universal" 16S rRNA gene primers used for amplicon sequencing may be biased against
439 *Nitrobacter*, though they matched *Nitrobacter* in NCBI.

440 *Nitrospira* **OTU** Analysis

441 The *Nitrospira* genus comprised 159 unique OTUs, which is consistent with a 442 prior study that found 120 unique *Nitrospira* OTUs in an activated sludge WWTP.²⁵ 443 However, two were consistently more abundant (greater than 1 log) than the others (OTU 444 ID# 4460870 (henceforth OTU A) and OTU ID# 1491 (henceforth OTU B)) (data not 445 shown). OTU A was consistently more abundant than OTU B, with a few exceptions (See Figure 6). In the 14 °C trials, OTU A and OTU B trends were similar to each other. 446 447 However, the RNA-based relative abundance for OTU A decreased after 48 hrs of 448 feeding, while the RNA-based relative abundance for OTU B increased at the same time. 449 Similarly, in the 20 °C trials, both OTUs were stable in relative abundance throughout the 450 starvation period, but after 12 hrs and 48 hrs of feeding, RNA-based relative abundance 451 of OTU A sharply decreased while the RNA-based relative abundance of OTU B 452 increased. In the 27 °C and 30 °C trials Nitrospira abundance and activity throughout 453 Starvation seemed to be dominated by OTU A, with the exception of Day 7 of 27 °C 454 Trial 1 in which the relative abundance of OTU A DNA decreased and the relative 455 abundance of OTU B increased to control overall *Nitrospira* trends at that point.



456

Figure 6 Relative abundance of dominant *Nitrospira* OTUs based on Illumina 16S rRNA
amplicon sequencing. Relative abundance was calculated based on abundance in rarefied
OTU table, normalized to total rarefied sequences (38,000 sequences). Influent WWTP
nitrifying aeration basin samples were plotted and labeled as a reference, with ammonium
dosing initiated at about day 18.5 of the experiment, as indicated on the plots.

To better understand the potential roles of OTU A and OTU B, a phylogenetic

464 tree was constructed consisting of the *Nitrospira* species found across the entire dataset

- 465 relative to reference *Nitrospira* spp. (Figure 7). OTU A and OTU B were found to be
- 466 closely related to *Nitrospira defluvii* and *Nitrospira nitrosa*, respectively. Indeed, the
- 467 amplified 16S rRNA sequences from OTU A and OTU B (152 bp) matched the
- 468 corresponding sequence locations of *N. defluvii* and *N. nitrosa*, respectively (see
- 469 Electronic Supporting Information).

- 470 The discovery of *Nitrospira* as by far the most dominant and persistent known
- 471 nitrifier in this study provides a logical explanation for the resilience of nitrifying



473 Figure 7 Phylogenetic analysis of *Nitrospira* OTUs based on maximum likelihood analysis 474 of 152bp segment of V4 region of 16S rRNA genes captured by amplicon sequencing. 475 OTU A and OTU B branches are labeled and reference *Nitrospira* spp. are highlighted in 476 bold, unlabeled branches represent Nitrospira OTUs other than OTU A and OTU B. 159 477 total Nitrospira OTUs from this study were included in this analysis, many of which 478 truncated along the same branch of the tree, as indicated. Tree was rooted to Leptospirillum 479 ferrodiazotrophum. Reference sequences are available in the Electronic Supporting 480 Information.

481

482 activity, which was revivable during recovery as well as the fact that known AOBs were

483 scarcely detectable by amplicon sequencing or qPCR of *amo*A genes. Specifically, the

- 484 assumed interdependence between AOB and NOB has recently been challenged by the
- 485 discovery of *Nitrospira* species capable of ammonia oxidation^{6,8} and *Nitrospira* spp.
- 486 capable of ureolytic and heterotrophic activity using oxygen or nitrate as electron

487	acceptors have also been described. ²⁶ Given such newly discovered metabolic
488	capabilities, it is entirely possible for Nitrospira to be the dominant or sole nitrifying
489	genus in an activated sludge system. As it turns out, N. nitrosa, with which OTU B most
490	closely matched, is a known comammox bacterium. Also, the amoA gene in comammox
491	Nitrospira differs phylogenetically from those targeted by qPCR in this study and thus
492	would explain why ammonia oxidation readily occurred in the presence of such low
493	levels of <i>amoA</i> gene copies. ^{8,48} Similarly, Camejo et al. recently found comammox
494	Nitrospira derived amoA genes were more abundant than AOB derived amoA genes in
495	lab scale activated sludge sequencing batch reactors.49
496	N. defluvii, with which OTU A most closely matched, is a member of Nitrospira
497	sublineage I and has been found to be capable of heterotrophic metabolism via
498	catabolism and assimilation of acetate, pyruvate, and formate. ^{27,50} The metabolic
499	versatility of N. defluvii provides a good explanation for the stability of OTU A in the
500	absence of addition of traditional substrates (ammonia and nitrite), since it could switch
501	to heterotrophic metabolism.
502	While N. nitrosa is a known comammox organism, possessing genes responsible
503	for both ammonia oxidation and nitrite oxidation, ^{8,27} unlike <i>N. defluvii</i> , it is not known to
504	be a mixotroph. Thus, its survival and maintenance of stable levels through the Starvation
505	phase would likely have to be driven via other means. N. nitrosa have been found to

506 possess genes responsible for urea cleavage to ammonia and multiple copies of *amoA*

507 genes demonstrating their diverse metabolic capabilities,²⁷ possibly lending to their

stability in the present study. The relation of OTU B to *N. nitrosa* may also explain the

509 sharp increase of OTU B RNA-derived relative abundance during the Recovery period of

510	several trials, as the increase in ammonia would likely increase activity and may have
511	given a competitive advantage to the comammox-like organism.
512	Potential limitations of 16S rRNA amplicon sequencing are important to
513	acknowledge. For example, multiple, diverse AOBs may have been present, each
514	individually at a relative abundance below detection, but collectively sufficient to drive
515	ammonia oxidation. Limitations in the "universal" primers used in this study
516	(515F/806R) to amplify 16S rRNA genes corresponding to AOB are also possible. Such
517	limitations were apparent in the case of Nitrobacter, which was detected by qPCR within
518	1 log of Nitrospira (Figure S1), but were not detected in the 16S rRNA amplicon
519	sequencing dataset. Similarly, these primers have been found to have a bias that leads to
520	under-representation of certain phyla, including <i>Thaumarchaea</i> , ⁵¹ which contains all
521	known AOA. ⁴⁷ Also, there may be other nitrifying archaea or bacteria present in
522	adequate abundance to be sequenced, but were not identified due to limitations of
523	genomic data in the reference database. ⁵² Thus, it is important to bear in mind limitations
524	of amplicon sequencing and employ multiple methods for assaying the targets of interest,
525	as was carried out in the present study.

526 Operational Conditions Shaping the Microbial Community and Role of *Nitrospira*

As microbial community profiling is increasingly being applied to gain insight into the composition of nitrifying activated sludge communities, it is apparent that there is a vast range in composition.¹ As was in the case in this study, a lab-scale nitrifying reactor and a WWTP were similarly found to be dominated by *Nitrosomonas* and *Nitrospira* based on pyrosequencing analysis; in that particular study *Nitrosomonas* represented 15.54% of the DNA sequences in the reactor and 0.05% in the WWTP.⁴⁰ It

533	is important to consider how factors in this study, such as WWTP operational conditions
534	and those of the Starvation and Recovery trials, might have shaped the composition of the
535	microbial community. Operational conditions are known to shape nitrifying microbial
536	community composition, for example, with slower growing Nitrosospira and Nitrospira
537	found to be more dominant in a strictly aerobic reactor relative to sequential
538	anaerobic/aerobic reactors. ¹² Nitrospira has been found to be sensitive to changes in
539	temperature as well ¹⁷ ; however they were observed to be remarkably stable in this study.
540	16S rRNA gene amplicon sequencing of both DNA and RNA as well as qPCR targeting
541	of Nitrospira told a consistent story of their persistence at substantial levels (1.4% to 11%
542	of 16S rRNA sequences and 6.3 to 9.3 log genes/mL 16S genes via qPCR) through
543	multiple trials carried out over nearly 2 years of sampling and multiple seasons. They
544	also persisted through Starvation and were capable of regaining activity during the
545	Recovery phase.
546	Various Nitrospira OTUs responded distinctly to the introduction of ammonia
547	during the Recovery trials. Both the 14 °C Trial 2 (Figure 6, panel a) and 20 °C Trial 2
548	(Figure 6, panel b), OTU A DNA and OTU B DNA relative abundance responded

549 similarly to introduced ammonia. However, relative abundance of these OTUs derived

from RNA analysis had opposing reactions to introduced ammonia (Figure 6). While

551 OTU A was dominant during starvation, after feeding started the relative abundance of

552 OTU A RNA decreased, while OTU B RNA relative abundance increased and became

the dominant *Nitrospira* OTU. This effect was not observed at 27°C or 30°C, which

554 suggests different selective pressures between these two OTUs at elevated temperatures

and is consistent with the nitrifying potential assays not fitting the trend at these

556	temperatures. Stronger response of OTU B RNA may be explained by the fact that, being
557	more closely related to the comammox bacterium N. nitrosa, it may gain an advantage
558	when ammonia is reintroduced to the environment.
559	Conclusions:
560	Here we examined shifts in microbial community composition and activity of
561	nitrifying activated sludge representative of a large, stable advanced WWTP during an
562	imposed period of ammonia starvation followed by introduction of ammonia during a
563	recovery phase. Several key observations were made:
564	• Starvation experiments simulated aerobic sludge digestion, correspondingly with
565	net decline in total bacterial populations represented by 16S rRNA genes.
566	• Nitrification activity potential, in terms of NO _x and nitrate production rates,
567	declined as starvation ensued, but temperature was not the sole driver of rates of
568	decline, with sludge collection date also an important factor.
569	• Relative abundances of various phyla shifted distinctly, depending on whether
570	DNA or RNA was profiled by 16S rRNA gene amplicon sequencing, suggesting
571	that some bacteria are killed, with varying rates of decay, while others persist via
572	dormancy or versatile metabolism.
573	• Known AOB were not abundantly detected, either through qPCR of <i>amoA</i> genes
574	or 16S rRNA gene amplicon sequencing.
575	• The only known nitrifier consistently detected throughout the Starvation or
576	Recovery trials was Nitrospira, traditionally thought of as an NOB, but now
577	known to contain comammox and mixotrophic members. The most dominant
578	OTUs were most similar to known comammox and mixotrophic strains, providing 28

an explanation for the anomalous activity decay trends with starvation andtemperature.

581	Together these results suggest that major WWTPs employing traditional nitrifying
582	activated sludge nitrification can contain nitrifying bacterial populations that deviate from
583	the canonical model of nitrification. The dominance of likely comammox and
584	mixotrophic nitrifiers is a critical consideration for improving design and operation of
585	WWTPs, particularly in the context of advanced and alternative nutrient removal
586	strategies seeking to out-select nitrite oxidation. Effort is needed to improve the accuracy
587	of kinetic parameters employed for nutrient removal given such deviations from past
588	assumptions and better optimize process parameters, such as aeration, food to
589	microorganism ratio, and wasting rates. Such accuracy will be vital to robust design and
590	implementation of advanced nutrient removal processes. Further, this study can help
591	understand how nitrifying activated sludge communities respond to and recover from
592	upset such as starvation. Ultimately, deeper knowledge of microbial community
593	structure can serve to inform designs that are more efficient as well as more resilient to
594	upset.

595

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