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Genome-Resolved Meta-Omics Ties Microbial Dynamics to Process Performance in Biotechnology for Thiocyanate Degradation

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23 ABSTRACT

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25 Remediation of industrial wastewater is important for preventing environmental 26 contamination and enabling water reuse. Biological treatment for one industrial 27 contaminant, thiocyanate (SCN), relies upon microbial hydrolysis, but this process is 28 sensitive to high loadings. To examine the activity and stability of a microbial community 29 over increasing SCN⁻ loadings, we established and operated a continuous-flow bioreactor 30 fed increasing loadings of SCN⁻. A second reactor was fed ammonium sulfate to mimic 31 breakdown products of SCN. Biomass was sampled from both reactors for 32 metagenomics and metaproteomics, yielding a set of genomes for 144 bacteria and one 33 rotifer that constituted the abundant community in both reactors. We analyzed the 34 metabolic potential and temporal dynamics of these organisms across the increasing 35 loadings. In the SCN⁻ reactor, *Thiobacillus* strains capable of SCN⁻ degradation were 36 highly abundant, whereas the ammonium sulfate reactor contained nitrifiers and 37 heterotrophs capable of nitrate reduction. Key organisms in the SCN⁻ reactor expressed 38 proteins involved in SCN⁻ degradation, sulfur oxidation, carbon fixation, and nitrogen 39 removal. Lower performance at higher loadings was linked to changes in microbial 40 community composition. This work provides an example of how meta-omics can increase 41 our understanding of industrial wastewater treatment and inform iterative process design 42 and development.

43

44 INTRODUCTION

45

46 Microbial communities in biotechnology have historically been treated as black boxes, 47 but as molecular methods have improved, our knowledge of these systems has deepened. 48 Increasingly, 'meta-omics' methods are being used to investigate critical processes and potential weak points in biotechnology, such as nitrogen and phosphorus removal or 49 bulking in wastewater treatment.¹⁻³ In particular, specialized treatment of industrial 50 wastewater has benefited from a genome-resolved meta-omics approach⁴⁻⁷ using high-51 52 throughput sequencing of community genomic DNA (metagenomics) or RNA 53 (metatranscriptomics), or spectral characterization of proteins (metaproteomics). These 54 data can be used to identify the key species involved in processes of interest. Improved 55 understanding of the activities and abundances of these organisms under varying 56 conditions could inform design and operation of these systems.

57 Thiocyanate (SCN) is a widespread industrial contaminant found at especially high concentrations (up to 4000 mg/L) in gold mining effluents. If not remediated, it can 58 affect human health and aquatic organisms.⁸⁻¹⁰ Notably, SCN⁻ is inhibitory toward iron-59 60 and sulfur-oxidizing microorganisms used in bio-oxidation processes at some gold mines 61 (such as BIOX®) and therefore must be removed before wastewater can be recycled 62 within a mining site or discharged into the environment. SCN⁻ can be biodegraded by chemolithoautotrophic bacteria as a source of energy,¹¹⁻¹⁵ by heterotrophic organisms as a 63 sole source of nitrogen,^{16,17} and by complex microbial consortia.¹⁸ The initial degradation 64 products are ammonium, carbon dioxide, and reduced sulfur compounds. An industrial-65 scale process known as Activated Sludge Tailings Effluent Remediation (ASTERTM, 66 Outotec, South Africa) successfully treats SCN-containing wastewater at several gold 67 mines.¹⁹ 68

Inoculated with sludge from the ASTERTM process, a long-running laboratory-69 scale SCN-fed bioreactor (known as the "SCN⁻ stock reactor") at the University of Cape 70 Town contains a characterized, diverse microbial community.²⁰⁻²² Previous work on this 71 72 community implicated several abundant *Thiobacillus* spp. in SCN⁻ degradation due to the 73 presence of an SCN⁻ operon in the genomes of these autotrophic bacteria. Results also 74 suggested the potential for nitrogen removal by Thiobacillus spp. and other community members, and the presence of heterotrophs.²¹ Questions remained regarding community 75 76 stability at different SCN⁻ loadings, expression of the observed metabolic potential, and 77 the importance of inter-organism interactions, especially for nitrogen removal. The SCN⁻ 78 stock reactor provided the inoculum for the bioreactors established in this study.

79 We used time-series genome-resolved metagenomics, in combination with 80 metaproteomic analyses of the samples from final time point, to track changes in the 81 microbial community of a newly-inoculated SCN⁻ bioreactor operated with increasing 82 loadings. To enrich for organisms that can use and remove the nitrogen produced by 83 SCN⁻ degradation, a second reactor with the same inoculum was fed ammonium sulfate 84 $(NH_4(SO_4)_{1/2})$ and molasses but no SCN. We describe the microbial community 85 structure, protein expression, and replication rates in both reactors during the experiment. 86 Our analysis linked shifts in community membership to changes in reactor function, 87 highlighted organisms and metabolic pathways active under high-SCN⁻ conditions, and 88 supported the importance of biofilm in this system.

89 90

MATERIALS AND METHODS

91

92 Reactor set-up, inoculation, and operation: Two continuous stirred tank reactors 93 (CSTRs) were inoculated with homogenized biofilm and planktonic samples from the 94 long-running SCN⁻ stock reactor at the University of Cape Town. Reactors were stirred 95 with a pitched-blade impeller at 270 rpm and sparged with filtered air at 900 mL/min. 96 One reactor was fed KSCN while the other was fed $NH_4(SO_4)_{1/2}$ at equivalent nitrogen 97 loadings in order to mimic the end-products of thiocyanate degradation. Both reactors 98 were also fed molasses (150 mg/L) and KH_2PO_4 (0.28 mM) to provide supplemental 99 nutrients. Feed contained increasing amounts of KOH to modulate reactor pH as 100 necessary (Figure 1) and small amounts of 5 N KOH were added directly to reactors if 101 observed pH was \leq 6.5. Bicarbonate (4 g/L) was added to the feed to buffer the system 102 from day 112 to day 136.

103 The reactors were run in batch-fed mode until SCN⁻ degradation was stably 104 observed in the SCN⁻ reactor, at which time both reactors were switched to continuous 105 feeding at a residence time of 42 hours (day 5). The hydraulic retention time (HRT) of 106 both reactors was lowered from 42 hours to 12 hours (days 5-68) and then maintained at 107 12 hours while the feed concentration of SCN⁻ or equivalent NH₄(SO₄)_{1/2} was increased 108 stepwise. The reactor was allowed to stabilize between each step to reach steady state 109 (**Figure S1**).

110

Sampling: Samples of biomass from each reactor were taken for metagenomic sequencing just before increases in feed concentration (Figure 1 and Table S1). Approximately 0.5 g (wet-weight) of biofilm was scraped from the wall of each reactor with sterile spatula and stored at -60 °C. Paired samples of planktonic biomass were

115 collected by filtering 300 mL of the liquid phase from each reactor onto a sterile 0.22 μ m 116 filter. Biomass was gently washed off the filter with sterile water, pelleted, and stored at -117 60 °C until further analysis. Filtered media was returned to the reactor to maintain 118 chemical continuity.

119

Chemical analysis: Bulk liquid was sampled daily for chemical analysis, filtered through
 a 0.22 μm filter, pH analyzed, and frozen at -20 °C until further analysis. SCN⁻ was
 measured using High Performance Liquid Chromatography as described previously.²¹ Ion
 chromatography was performed to quantify nitrate and sulfate (Supporting Information).

124

125 DNA extraction and sequencing: DNA was extracted using a NucleoSpin[®] soil 126 genomic DNA extraction kit (Machery-Nagel, Germany) with the inclusion of a repeated 127 extraction step, according to the manufacturer's instructions. Paired-end Illumina TruSeq 128 libraries with either tight insert fragment sizes of 800 bp or regular insert sizes of 500 bp, 129 depending on the sample, were prepared at the Joint Genome Institute (Walnut Creek, 130 CA) (Table S1). Libraries were sequenced on an Illumina HiSeq-2500 in rapid run mode 131 to yield 250 bp paired-end reads.

132

133 Metagenomic assembly, binning, and annotation: Reads from each sample were 134 trimmed based on quality scores using sickle (https://github.com/najoshi/sickle) and then assembled independently with idba ud²³. Binning of the assembled scaffolds was 135 performed using ggKbase (ggkbase.berkeley.edu) based on scaffold taxonomy, percent 136 137 GC, and sequencing coverage. Within each assembly, bins were refined and added using 138 differential abundance data visualized in emergent self-organizing maps (ESOMs) as in 139 Sharon *et al.*²⁴ (see Supplemental Information). Each ESOM was trained and visualized 140 using databionic ESOM tools (Figure **S2**) (http://databionicesom.sourceforge.net/index.html).²⁵ In two samples (planktonic inoculum and SCN-141 reactor T2 biofilm), subassemblies using $1/60^{\text{th}}$ or $1/50^{\text{th}}$ of the reads, respectively, were 142 143 performed to improve assembly of the most abundant organism as previously described by Hug *et al*.²⁶ 144

145 Many bins were redundant given the recurrence of organisms across the time series experiment. Nucmer²⁷ was used to align sequences and identify sets of bacterial 146 147 genomes sharing \geq 98% nucleotide identity across > 50% of the sequence. The best bin 148 was chosen based on genome completeness and length for inclusion in a de-replicated 149 dataset. Genomes without replicates were also included, except for two known 150 contaminant genomes ('Candidatus Altiarchaeum hamiconexum' and an 151 Epsilonproteobacterium) from another sequencing run on the same lane. Bins were 152 excluded from the final de-replicated dataset if they contained < 36 of 51 single copy 153 genes (SCG) or > 8 multi-copy SCG prior to curation (see Table S2A). One recurring 154 eukaryotic genome bin, one chloroplast, several mitochondria, phages, eukaryotic 155 viruses, and plasmid bins were included in the de-replicated dataset. De-replicated 156 bacterial genomes were curated using ra2.py, an automated curation method that uses 157 coverage and paired-end read information to find and reassemble or mask regions with 158 mis-assemblies

(https://github.com/christophertbrown/fix_assembly_errors/releases/tag/2.00).²⁸ Curation
 used the reads of the sample from which the genome originated.

Annotation of genome bins used reciprocal ublast²⁹ searches against KEGG³⁰ and 161 UniRef100,³¹ as well as single-direction searches against UniProt.³² Functional genes and 162 163 marker genes were identified by annotations and using hmmsearch (HMMER 3.1b2; 164 http://hmmer.org/) with Hidden Markov Models (HMMs) from TIGRFAM (v15.0), PFAM,³³ and with custom HMMs³⁴ (accessible at https://github.com/banfieldlab). 165

166

167 **Community composition:** Bowtie2 was used with default settings to map reads from 168 each sample to the de-replicated dataset (Figure S3). The resulting files were filtered 169 using mapped.py (https://github.com/christophertbrown/mapped) to remove reads that 170 mapped with > 3 mismatches. Coverage for each genome in each sample was calculated 171 and values $\leq 1x$ were converted to zero. Genome coverage values were then normalized 172 by dividing by the number of reads for each sample and then multiplying by the number 173 of reads in the largest sample. Normalized coverage was used as a proxy for the relative 174 abundances of organisms across samples (Figure S4). A concatenated ribosomal protein 175 tree containing references and sequences from SCN⁻ bioreactors was constructed as 176 previously described³⁵ (Figure S5). Datasets for each of 16 ribosomal proteins were 177 aligned independently with MUSCLE, alignments were trimmed and columns with 99% 178 gaps were removed in Geneious, and trimmed alignments were concatenated. RAxML 179 was used to generate a maximum likelihood phylogeny under the LG + gamma model.

- 180
- 181

Variant analysis and Replication rate calculations: see Supporting Information.

182 183 Protein extraction and proteomic data analysis: Proteins were extracted as previously described³⁶ and ~ 1 mg of protein was subjected to trichloroacetic acid precipitation and 184 185 subsequent digestion with trypsin. Proteolytic peptides were analyzed via an online nano 186 2D LC-MS/MS system interfaced with hybrid LTO-Orbitrap-Velos MS (ThermoFisher 187 Scientific). Subsequent processing of the collected spectra was done using Myrimatch³⁷, 188 with the de-replicated set of genomes as the database (Supporting Information). Peptide 189 identifications were quality-filtered to < 1% false discovery rate. Analysis of proteins 190 involved in key metabolic pathways considered spectral counts for unique peptides and 191 total spectral counts for each protein from two technical replicates.

192

193 Data availability: Raw read data is accessible at NCBI under accession number 194 SRP056932 (http://www.ncbi.nlm.nih.gov/sra/SRP056932) and genome accession 195 numbers may be found in Table S2A. Genome bins and sequences for scaffolds, genes, 196 and proteins can be viewed and downloaded at http://ggkbase.berkeley.edu/scnpilot-197 dereplicated/organisms. Proteomics data is available at 198 https://massive.ucsd.edu/ProteoSAFe/datasets.jsp (MassIVE ID MSV000080104).

199

200 **RESULTS AND DISCUSSION**

201

202 **Reactor chemistry and efficiency:** In the newly-inoculated SCN-fed and $NH_4(SO_4)_{1/2}$ 203 reactors, the loading rate was increased across 238 days. Samples were taken for 204 metagenomic analysis after HRT reached 12 h. In the SCN⁻ reactor, the SCN⁻ removal 205 rate consistently increased to match the increasing loading rate to a maximum of 1.07 mmol. h^{-1} (Figures 1A and S1A). On further increase to 1.43 mmol. h^{-1} , the SCN⁻ removal 206

rate decreased and efficiency dropped to near 50%. On average, the stoichiometry 207 208 between the SCN⁻ removal rate and sulfate output was 1.05:1, near the 1:1 ratio expected 209 based on known SCN⁻ degradation mechanisms coupled to complete oxidation of the 210 sulfide released.

211 As loading increased, the thickness of the biofilm that formed on all surfaces 212 within the reactor increased. During one period early in the experiment, nitrate (NO_3) 213 output reached up to 30% of nitrogen input as SCN⁻ (days 86-107) and fluctuated 214 thereafter, reaching a maximum of 64% of nitrogen input. After day 200, nitrate output 215 remained consistently low and sulfate accumulated. Base was added periodically to 216 counter acidification as loadings increased. The SCN⁻ removal rate decreased during one 217 period of high pH that resulted from over-correction of the feed pH (Figure 1A).

218

In the $NH_4(SO_4)_{1/2}$ reactor, the rate of sulfate leaving the reactor rose steadily 219 throughout the experiment, matching the sulfate loading rate and indicating that little 220 sulfate was retained in biomass or converted to other forms (Figure 1B and S1B). The 221 nitrate output rate increased with decreasing HRT and then increased more slowly as 222 biofilm established and thickened. Overall, higher nitrate effluent concentrations were 223 measured in the $NH_4(SO_4)_{1/2}$ reactor than in the SCN⁻ reactor (Figure S1B).

224

225 Genome recovery over the sample series: Biofilm from the two reactors was sampled 226 at four time points (T1-T4) during the experiment, and concurrent samples of planktonic 227 biomass were collected at T2 and T3 in the SCN⁻ reactor and T2-T4 in the NH₄(SO₄)_{1/2} 228 reactor. The inoculum for these reactors (T0), biofilm and planktonic biomass taken 229 from the SCN⁻ stock reactor, was also sampled (Figure 1 and Table S1). Independent 230 metagenomic assemblies were performed for each sample and 789 bacterial genome bins 231 were reconstructed (Figure S2). Two genomes represented highly abundant organisms, 232 and subassemblies substantially improved these genomes (see Methods). Dereplication 233 across the time-series yielded a non-redundant set of 144 draft-quality genomes (Tables 234 **1**, **S2A**). Eukarvotic, mitochondrial, chloroplast, phage and plasmid genomes were also recovered and de-replicated (Tables 1, S2B). No Archaea were detected in this system, 235 consistent with previous studies²⁰⁻²². Mapping reads from each assembly demonstrated 236 237 that this non-redundant genome set accounted for between 72.5 and 93.2% of the data 238 (Figure S3). This level of genome recovery approaches that reported for much simpler communities such as those from the infant gut^{38} . The de-replicated metagenomic dataset 239 240 was used as a database for proteomic searches and accounted for 34, 32, and 15% of 241 high-quality peptides from the SCN⁻ reactor biofilm, $NH_4(SO_4)_{1/2}$ reactor biofilm, and 242 $NH_4(SO_4)_{1/2}$ reactor planktonic samples, respectively, at T4. This level of identification is 243 comparable to that seen with the same type of analysis on infant gut metaproteomes 244 paired to metagenomic databases³⁹.

245

246 **Community structure and metabolic potential:** Hierarchical clustering of samples 247 based on their community compositions grouped the samples first by the reactor, then by 248 type of biomass and time point from which the samples were taken (Figure S4). 249 Clustering organisms by abundance delineated several distinct groups: a small subset of 250 organisms was present in both reactors, while other subsets were found at high-251 abundance in SCN⁻ community or the NH₄(SO₄)_{1/2} community. Still other organisms 252 were abundant primarily in the inoculum (Figure S4). The taxonomic identity of genomes was determined via phylogenetic reconstruction using ribosomal proteins (**Figure S5**). In order to identify key organisms in the bioreactor communities, we characterized the metabolic potential encoded and expressed by each genome with respect to the key processes of SCN⁻ degradation, sulfur, ammonium, and nitrite oxidation, denitrification, and carbon fixation (**Figure S4**, **Table S2A**).

258

259 **SCN⁻ removal and sulfur cycling:** Four genomes, Thiobacillus 1, Thiobacillus 3, Thiobacillus_4 and Afipia_1, contain one of two known types of SCN⁻ hydrolases^{11,40}. 260 261 The corresponding organisms were abundant only in the SCN^{-} reactor (Figure 2), and proteomics data support activity of these organisms in SCN⁻ degradation, sulfur oxidation 262 263 and carbon fixation (Figure 3). Peptides detected by proteomics matched to predicted proteins in the recently described SCN⁻ operon from *Thiobacillus* spp.²¹ Specifically, the 264 265 genes in the operon were detected by proteomics with spectral hits in at least one of the 266 three *Thiobacillus* genomes that contained this operon. Interestingly, an SQR-like protein 267 in this operon had the highest count of unique spectral hits for all three genomes (**Table** 268 **S3**). The cbiM-like protein hypothesized to be involved in cobalt metabolism was the 269 only protein from the operon not detected in proteomics.

Consistent with the genome of *Thiobacillus denitrificans*,⁴¹ all six recovered *Thiobacillus* genomes encode the potential for autotrophic growth on sulfur compounds via numerous sulfur-related genes from multiple pathways (**Table S2A**). In a transcriptomics-based study of *Thiobacillus denitrificans*, some of these genes were constitutively expressed (*sox*, *rDsr*, *apr*, *atps*) whereas others are upregulated under denitrifying conditions (e.g., sulfide quinone reductase).⁴² We identified all sulfur oxidation genes in proteomics for several of the Thiobacilli described here (**Figure 3**).

277 In addition to the four genomes encoding thiocyanate-degradation, twenty-two 278 other genomes possess the Sox pathway (including at least 4 of soxX, Y, Z, A, and B, with 279 or without soxCD; Table S2A). Of these, the genomes with the highest normalized 280 coverages in the SCN⁻ reactor were Burkholderiales_6, Thiobacillus_2 and Rhizobiales_3 281 (Figure 2). Sulfur oxidation may proceed from sulfide to elemental sulfur or sulfate, 282 likely determined by the availability of electron acceptors, as discussed below. Sulfur globules may be produced by SCN-degrading Thiobacillus spp., which use the reverse 283 dissimilatory sulfite reductase (rDsr) pathway instead of soxCD⁴³. In turn, other sulfur 284 285 oxidizers may use this elemental sulfur and any excess sulfide produced by SCN-286 degradation. Since chemical data showed that SCN⁻ was completely converted to sulfate, 287 and proteomic data showed expression of Sox proteins from genomes lacking known 288 thiocyanate hydrolases (Figure 3), we suspect that sulfur species were passed from SCN-289 degraders to the rest of community as "metabolic handoffs". Additionally, some of these 290 organisms may use as-yet-unidentified pathways for SCN⁻ degradation.

291

292 **Community dynamics and SCN⁻ removal across increased loadings:** As SCN⁻ 293 loadings and SCN⁻ degradation rate increased (**Figure 1**), the relative abundance of 294 *Thiobacillus* spp. whose genomes encode SCN-degradation also increased (**Figure 2**), 295 with Thiobacillus_1 alone accounting for 38% of all reads at T2. During operation at the 296 two final loading rates (T3 and T4), the Thiobacillus_1 population decreased in relative 297 abundance, concordant with a decrease in SCN⁻ degradation rate and reactor efficiency. Given this observation, we looked for other changes associated with loss of reactorefficiency.

300 First, a read mapping-based sequence variance analysis of the Thiobacillus 1 301 population in each sample showed that it was largely clonal throughout the time series 302 but two distinct strains were present at the last time point, where the relative strain 303 proportions were ~60 and 40% (Figures S6A and S6B). The genome for the second 304 strain was not recovered, but a scaffold corresponding to the SCN⁻ operon was identified 305 among the un-binned metagenomic data from this time point. We noted a few differences 306 in the protein sequences of genes contained in this operon, which could in principle affect 307 the efficiency of SCN^{-} degradation relative to the dominant strain (**Figure S7**).

308 A second change in the community was the increase in relative abundance of 309 Burkholderiales_6, which became dominant in T3 and T4 (Figure S8). No known genes 310 for SCN⁻ degradation were found in the Burkholderiales 6 genome, but previous studies 311 have isolated and characterized a strain of Burkholderia phytofirmans capable of thiocyanate degradation with acetate as a carbon source.⁴⁴ The genome of 312 313 Burkholderiales_6 contains genes encoding the sox pathway, and the corresponding 314 proteins were detected in proteomics (Figure 3). Hence we infer that the 315 Burkholderiales_6 organism likely used reduced sulfur species for mixotrophic growth, 316 and may have increased in relative abundance concordant with the accumulation of 317 organic matter in the reactor (in the form of biomass), perhaps outcompeting 318 Thiobacillus_1 for SCN⁻ as a source of nitrogen and energy. Overall, the results suggest a 319 transition from autotrophic to mixotrophic / heterotrophic thiocyanate degradation at high 320 thiocyanate loadings and after long periods of reactor operation.

A third change in the reactor community that could have affected SCNdegradation was substantial algal growth at the last time point. This was visually observed in the planktonic portion of the SCN⁻ reactor and could have affected microbial population dynamics and reactor efficiency. Lastly, the increase in residual SCN⁻ concentrations in the reactor may have led to toxicity effects including lower bacterial replication rates (see below). This in turn could have reduced the SCN⁻ degradation rate, creating a negative feedback effect on reactor performance.

328 Overall, we speculate that the decline in reactor efficiency at high loading rates 329 occurred when the capacity for SCN⁻ degradation was exceeded. The abundance of 330 Thiobacillus cells may have been insufficient to meet the demand for SCN⁻ degradation 331 owing to a maximum specific SCN⁻ degradation rate. Alternatively, degradation may 332 have been inhibited or the per-cell rate of degradation may have decreased. Since 333 metagenomic data provide relative abundance information, the apparent decrease in 334 Thiobacillus_1 relative abundance may have been due to increases in abundance of other 335 organisms, such as Burkholderiales 6. Further studies are needed that apply 336 measurements of absolute, species-specific biomass and metabolic rates.

337

Nitrogen removal and dynamics over time: Since SCN⁻ degradation releases nitrogen in the form of ammonium, we looked for possible mechanisms of nitrogen cycling and removal to N₂. No anamox genes were detected in any genome and, based on identified genes and genomes, the conversion of ammonium to nitrate occurred aerobically. A single genome in the dataset, Nitrosospira_1, encoded the potential for ammonium oxidation (**Figure S4**). The Nitrosospira_1 organism became enriched at early time points 344 in the NH₄(SO₄)_{1/2} reactor and later in the SCN⁻ reactor (Figure 2) and was active at the 345 final time point, based on proteomic data (Figure 3). Two predicted nitrite oxidizers, 346 Nitrobacter 1 and Nitrobacter 2, were present at low abundances in the $NH_4(SO_4)_{1/2}$ 347 reactor (Figure 2) and were so low in abundance in the SCN⁻ reactor that their genomes 348 did not assemble. However, proteins for nitrite oxidation corresponding to one of these 349 genomes were detected in samples from both reactors (Figure 3). Low abundance but 350 high activity has been observed previously for other nitrite oxidizing bacteria, and some 351 have hypothesized that high nitrite oxidation activity may be a requirement for growth, given the low energy yield of this metabolism.⁴⁵ Despite the low relative abundance of 352 353 both predicted ammonium and nitrite oxidizers, nitrate was detected in the effluent of 354 both reactors during the initial ramping phases (Figure 1).

355 Searching for mechanisms of nitrate removal, we identified 92 genomes that 356 contained at least one gene involved in denitrification (including nar, nap, nirS, nirK, 357 norB/norZ, and nosZ; Table S2A). Fourteen of these genomes encoded the capacity for 358 complete nitrate reduction to N₂, whereas 49 had only one gene in the pathway. Genomes 359 of complete denitrifiers corresponded to three of the predicted autotrophs implicated in 360 SCN⁻ removal: Thiobacillus 1, Thiobacillus 3, and Afipia 1 (Figure 2). For the 361 Thiobacilli, several denitrification-related complexes were detected with proteomics 362 (Figure 3). SCN⁻ hydrolysis and concomitant sulfide oxidation coupled to denitrification may be possible in these organisms, as was described for *Thioalkalivibrio*.⁴⁶ Based on 363 364 proteomic evidence, the Burkholderiales_6 organism, which became abundant in T3 and 365 T4 in the SCN⁻ reactor, also likely contributed to denitrification (Figure 2 and 3). All 366 denitrification genes except *norB* were identified in proteomics from the SCN⁻ reactor biofilm (Figure 3). The limited detection of NorB may be an extraction bias artifact due 367 to numerous transmembrane domains in these proteins.^{47,48} 368

In the NH₄(SO₄)_{1/2} reactor, three of most abundant bacteria, Rhodanobacter_1, Xanthomonadales_1, and Novosphingobium_1 may have roles in denitrification (**Figure** and **Table S2A**). The potential for dissimilatory nitrate reduction to ammonia (via *nrfA*) was detected in genomes from several members of the Bacteroidetes and one *Aeromonas* sp. but most of these were abundant only in the inoculum (**Table S2A**) and NrfA was not detected in the proteomic data.

375

376 Changes in bacterial replication rates across the time series: We used a recently established approach to investigate bacterial replication rates from metagenomics⁴⁹ with a 377 new implementation that reports rates as index of replication (iRep) values.⁵⁰ In the SCN 378 379 reactor biofilm, iRep values for most genomes increased between T0 and T1, suggesting 380 replication proceeded more quickly in newly-forming biofilm than in inoculum biofilm 381 taken from the long-running SCN⁻ stock reactor (Figure S7). Over the remainder of the 382 experiment (T2-T4), iRep values decreased for most genomes, especially toward the end 383 of the experiment. This may have been due to toxicity of residual SCN⁻ in the reactor 384 media or to spatial limitations for growth within the thickening biofilm.

In contrast, iRep values in the $NH_4(SO_4)_{1/2}$ reactor biofilm were initially low, but increased from T1 to T3. This is suggestive of a period of adaptation, as organisms adjusted to the new conditions relative to the SCN⁻ stock reactor (**Figure S9**). 389 Biofilm and planktonic communities: The planktonic and biofilm portions of the 390 reactor were sampled separately in order to understand whether these represented distinct 391 communities. At T1, the planktonic fraction of each reactor was very dilute, yielding 392 inadequate amounts of DNA for sequencing, and at T4, the planktonic portion of the 393 SCN⁻ reactor was overgrown with algae. At T2, metagenomes for planktonic samples 394 from both reactors were highly enriched in a rotifer genome, which accounted for 45 and 395 25 % of the sequence data in the $NH_4(SO_4)_{1/2}$ and SCN^2 reactors, respectively. With 396 microscopy, rotifers were observed grazing on biofilm (Figure S10). Other eukaryotes 397 were observed in both reactors, and many of these organisms were at higher relative 398 abundance in the planktonic samples compared to biofilm (Table S2). Recurring 399 mitochondrial sequences recovered in the metagenomes corresponded to relatives of 400 Vermamoeba vermiformis, Acanthamoeba spp., Naegleria fowleri, and Chlorella sp., 401 identified based on the phylogenetic profile of their proteins and searches against NCBI-402 nr (Table S2B).

Eleven bacterial genomes in the dataset derived from predicted symbionts, as indicated by their phylogenetic affiliations and/or reduced genome sizes (**Figure S4** and **Table S2A**). These included two complete genomes for organisms belonging to the phylum Saccharibacteria^{51,52} (formerly TM7). One organism, TM7_2, was the only putative symbiont found at high abundances in planktonic fractions of both reactors (**Figure 2**).

409 Overall, we estimate that biofilm constituted the majority of the biomass in the 410 reactors, and sloughing may have contributed to the planktonic community. While SCN⁻ degradation itself does not rely on biofilm,^{53,54} the formation of biofilm effectively 411 412 uncoupled the HRT from bacterial growth rates, preventing wash-out as the HRT was decreased. This may have allowed Thiobacillus spp., and nitrifiers to reach higher 413 414 population sizes than would otherwise have been possible, thereby converting higher 415 loadings of SCN⁻ to nitrate. The biofilm likely also provided anoxic conditions promoting 416 denitrification, as suggested by proteomic data (see above).

417

418 Long-term community stability and phage susceptibility: We compared the 114 419 bacterial genomes in this study to 86 genomes reconstructed in a prior study of the SCN 420 stock reactor and a daughter reactor fed cyanide and SCN⁻ (CN-SCN reactor) conducted two years earlier.²¹ Thirty-one genomes were matched, overlapping by at least a total of 1 421 422 Mbp at 98% nucleotide identity (Table S2A). These included close relatives of the three 423 SCN-degrading *Thiobacillus* spp. and Burkholderiales 6 enriched in the SCN-treated 424 reactor studied here. Given the importance of these three populations, we looked for 425 evidence of recent phage infections, based on changes to CRISPR loci over time. 426 Thiobacillus 1 has no CRISPR locus, perhaps making it more susceptible to acquisition 427 of mobile elements and phage attack. The CRISPR locus in Thiobacillus_3 was identical 428 in all versions of this genome recovered from the current study, but was not recovered in 429 the previous study. The recovered Thiobacillus_4 genomes belonged to two distinct 430 CRISPR sub-types that differed from one another in 12 spacers at the 3' end of the array: 431 sequences from biofilm and planktonic inoculum samples comprised one version, and 432 sequences from later in the time series (and those recovered previously in the SCN⁻ stock 433 reactor and the CN-SCN reactor) comprised the second version. Importantly, no spacers 434 from Thiobacillus_3 and Thiobacillus_4 targeted any sequence in the metagenomes (or previous metagenomes from the SCN⁻ stock or CN-SCN reactors), suggesting little recent
 phage interaction.

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438 System overview: Relatives of known SCN-degrading chemolithoautotrophs (several 439 Thiobacillus spp. and one Afipia sp.) are predicted to oxidize the SCN-sulfur as their sole 440 energy source under both aerobic and anaerobic conditions in the SCN⁻ reactor (Figure 441 **4A**). Sulfide oxidation may stop at elemental sulfur when parts of the reactor become 442 anaerobic⁵⁵ (Figure 4B), and proteomic data suggest that several *Thiobacillus* spp., as 443 well as Burkholderiales 6, coupled sulfur oxidation to denitrification (Figure 3). In fact, 444 based on their abundance, and proteomic evidence, these organisms may be the key 445 denitrifiers in the system. Other sulfur oxidizing autotrophs and mixotrophs may use 446 reduced sulfur compounds produced during SCN⁻ degradation, including elemental 447 sulfur. The combination of sulfide oxidation by SCN-degraders and non-degraders may explain the complete conversion of sulfur from SCN⁻ to sulfate (Figure 4B). The 448 449 breakdown of SCN⁻ produces ammonium that can be converted to nitrate by autotrophic 450 ammonium and nitrite oxidizers. Overall, some heterotrophs in the system likely 451 contributed to sulfur oxidation and denitrification, and likely also to biofilm formation 452 and biofilm integrity (perhaps via filamentous morphology, see Figure S10). 453 Heterotrophs may also break down SCN⁻ as a source of nitrogen (via an unknown pathway).⁴⁴ However, this SCN degradation may be inhibited if there is abundant 454 nitrogen available as ammonium, as observed in some alkaliphiles.¹⁵ Lastly, eukarvotes 455 456 such as rotifers and amoeba are predators and thus contribute to carbon turnover in the 457 system.

458

459 Engineering of SCN⁻ degradation by a microbial community: The consortium described here can completely hydrolyze SCN⁻ and oxidize sulfide under a range of SCN⁻ 460 461 loadings but reduced performance occurred at higher loadings. Smaller increases in 462 concentration to reach higher loadings may lead to sustained reactor performance by 463 allowing microbial cell numbers and associated volumetric degradation rates to keep pace 464 with input SCN⁻. For treating such high SCN⁻ concentrations at low retention times (>12.9 mM or 750 ppm in 12 hours), an attached growth design may have better SCN⁻ 465 466 and nitrogen removal.

467 Studies on such biofilm-based systems for SCN⁻ treatment have demonstrated the effectiveness of rotating biological contact (RBC)^{56,57} and fixed bed and fluidized-carrier 468 type reactors.^{58,59} Potential weaknesses of these systems include biofilm overgrowth 469 leading to poor mixing and aeration. The ASTERTM process, modeled here with 470 laboratory-scale CSTRs, was designed to be easy to establish at remote locations and to 471 accommodate suspended solid tailings when necessary.^{53,54} The results of our work on a 472 473 laboratory-scale CSTR suggest that biofilm can play an important role when this reactor 474 design is operated at high loadings.

Our work highlights the applicability of bioinformatics tools to gain a mechanistic understanding of contaminant degradation by a microbial community, to assess community stability, and ultimately, to inform engineering design choices. Others have called for broader use of metagenomics to advance biotechnology, including in wastewater treatment,^{60,61} and this study represents a step toward the use of such techniques in the field. The level of resolution achieved using metagenomics combined with metaproteomics allowed us to access not only phylogenetic classifications and
diversity of community members, but also which members express key proteins involved
in the degradation process. The dataset and analysis provide valuable information that
can be used to generate primers or probes for on-site measurements.

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499 SUPPORTING INFORMATION AVAILABLE

- 500 This information is available free of charge at http://pubs.acs.org.
- 501

502

503 AUTHOR CONTRIBUTIONS

RSK, RJH, STLH, and JFB designed the study; RSK and JFB wrote the paper; all coauthors read and contributed to revisions of the paper; RJH operated and sampled the
bioreactors and conducted analysis for metadata, with contributions from STLH; SGT
provided DNA sequencing; RI and RLH collected and processed proteomic data; RSK

508 analyzed metagenomic and proteomic data with contributions from BCT, CTB, and KA.

509 FIGURES AND TABLES

Table 1. Counts and completeness estimated by single copy gene (SCG) inventories for

512 de-replicated bins resulting from 15 metagenome assemblies.

Bacterial genomes (144)	Count
Genomes \geq 96% complete (49/51 SCG)	111
Genomes in \leq 30 scaffolds and 49/51 SCG Genomes with \geq 90% of sequence in scaffolds > 10 kb	37 117
Non-bacterial bins (90)	Count
Plasmids and mobile elements	45
Phage	25
Eukaryotes	1
Mitochondria	15
Chloroplasts	1
Viruses	3



Genome-based dynamics

- 515 Genome-based **TOC/Abstract art**



Figure 1. Chemical parameters of operation for the SCN⁻ reactor (**A**), and NH₄(SO₄)_{1/2} reactor (**B**). Sampling time points are indicated above plots. Gray shading indicates the different loading regimes.



Figure 2. Metabolic potential and relative abundances of key organisms of interest over time in biofilm and planktonic samples from both reactors. Gray shading corresponds to increasing loading rates of SCN⁻ or $NH_4(SO_4)_{1/2}$ as in Figure 1. SCN⁻ degraders (orange) and several key sulfur oxidizers (yellow and purple) were found at higher relative abundance in the SCN⁻ reactor while ammonium and nitrite oxidizers (green) were at higher relative abundances in the $NH_4(SO_4)_{1/2}$ reactor. Several highly abundant heterotrophs (red and blue) and one possible sulfur oxidizing mixotroph (purple) were present in both reactors. Note different y-axis scales. Hexagons indicate carboxysomes where annotations support this prediction.



- 542
- 543 **Figure 3.** Metaproteomics at end point (T4) in SCN⁻ reactor shows expression of
- 544 genes involved in SCN⁻ degradation and byproduct breakdown in key organisms.
- 545 Each arrow indicates that the average of unique spectral counts across two technical
- 546 replicates was ≥ 2 for at least one subunit or component of the enzyme complex 547 involved in the chemical transformation.
- 548
- 549



Figure 4. SCN⁻ removal and sulfur, nitrogen, and carbon cycling in the reactor

554 system depicted based on metagenomic analysis.

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