

1 **Supplementary Materials for Magnuson et al. “Active lithoautotrophic and methane-**
2 **oxidizing microbial community in an anoxic, sub-zero, and hypersaline High Arctic**
3 **spring”.**

4
5 **Extended Materials and Methods**

6 i. Site description and sample collection

7 Lost Hammer (LH) spring discharges through a precipitated mineral salt tufa as described in
8 previous publications (1-5) (Figure S1). LH emits gases composed of methane (50%), nitrogen
9 (35%), carbon dioxide (10%), and trace hydrogen, helium, and short-chain alkanes (1). The
10 spring sediments and water contain high concentrations of sulfate (100 000 mg/kg) as well as
11 sulfide (<50 mg/kg), ammonia (2.55 mg/kg), nitrate/nitrite (2.87 mg/kg), and iron (13 000
12 mg/kg) (1). Physical and geochemical parameters in the outlet (Table S1) have remained highly
13 stable since 2005, allowing comparison among samples collected in different years.

14
15 For this study, sediment samples (top ~10 cm) were collected in July 2017 and July 2019 with an
16 ethanol-sterilized scoop, and stored in sterile Falcon tubes filled to maximum to avoid aerobic
17 headspace. Sediment from July 2017 was used for metagenomic sequencing, and sediment from
18 July 2019 was used for RNA and SAG sequencing. In parallel, sediment for RNA extractions
19 was mixed with Zymo Research DNA/RNA Shield (Irvine, CA, USA). Samples were kept at
20 <5°C during transportation to Montreal, after which they were stored at -20°C for DNA and
21 RNA extraction and at -5°C (unfrozen) for SAG sequencing. Physical and geochemical
22 parameters of the overlying water were measured *in situ* with a YSI Professional Plus
23 Multiparameter instrument (Yellow Springs, OH, USA) and a PyroScience Piccolo2 oxygen

24 meter (Aachen, Germany). Ortho-phosphate and ammonia were measured *in situ* with
25 CHEMetrics Inc. (Midland, VA, USA) test kits.
26
27 ii. DNA extraction, metagenome sequencing, and metagenome data analyses
28 DNA was extracted from two 5 g portions of sediment with a Qiagen DNeasy PowerMax Soil
29 Kit (Hilden, Germany). The resulting DNA from each sediment sample was concentrated with a
30 Thermo Fisher Scientific SpeedVac Vacuum Concentrator (Waltham, MA, USA) and sequenced
31 on a HiSeq2500 (2x126 base reads) (Illumina, San Diego, CA, USA) at The Centre for Applied
32 Genomics (Toronto, ON, Canada). Low-quality reads and bases were trimmed with
33 Trimmomatic (v0.38, settings LEADING:3 TRAILING:3 SLIDINGWINDOW:4:15) (6).
34 Remaining reads were classified with Kaiju (v1.7.3, default settings, nr_euk database) (7) and
35 phyloFlash (v3.4) (8). The reads from the two sediment samples were co-assembled with
36 Megahit (v1.1.3, setting meta-sensitive) (9) as well as assembled separately with metaSPAdes
37 (v3.13.0, default settings) (10). Reads were mapped to each assembly with BMAP (v38.26,
38 minid=0.95) and contigs longer than 5000 bp were binned with MetaBAT (v2.12.1) (11). Bin
39 completeness and contamination was estimated with CheckM (v1.0.12) (12). The Megahit co-
40 assembly and resulting bins were selected for downstream analysis based on sequencing statistics
41 as determined by metaQUAST (v5.0.1) (13) and number of high- and medium-quality bins.
42 Sequencing and assembly statistics can be found in Tables S2 and S3. The metagenome was
43 annotated with the Joint Genome Institute's IMG/M system using KEGG, COG, and pfam
44 databases (14, 15). Bins were classified with the Genome Taxonomy Database Toolkit (v1.3.0,
45 reference data R05-RS95) (16). Phylogenomic trees were constructed with Anvi'o (v6.2) using
46 the Bacteria_71 collection of single copy genes (17). Amino acid sequences for all genes were

47 concatenated, with a total alignment length of 24 451 bp, and approximately-maximum-
48 likelihood trees were constructed in FastTree with default settings within Anvi'o, with midpoint
49 rooting. Additional analyses were as follows: FeGenie was used to identify iron-related genes
50 (18); *amoA* and *pmoA* were distinguished by HMMer using FunGene Hidden Markov Models
51 (19); hydrogenases were classified with hydDB (20). Reductive and oxidative DsrAB were
52 classified as follows: DsrAB amino acid sequences were aligned against reference sequences
53 from Muller et al. (21) using MUSCLE with default settings (22). Maximum likelihood
54 phylogenetic trees were constructed in CLC Genomics Workbench (v. 12.0.3) using the WAG
55 protein substitution model and 1000 bootstraps (Figure S9). DsrAB were classified as reductive
56 or oxidative based on phylogenetic clustering and the presence of accessory proteins as in
57 Anantharaman et al. (23).

58

59 iii. Single cell sorting, genome amplification, sequencing, and data analyses

60 Sediment was shipped on ice to the Single Cell Genomics Center at the Bigelow Laboratory for
61 Ocean Sciences (East Boothbay, ME, USA) for SYTO-9 fluorescence-activated single-cell
62 sorting (FACS), DNA extraction, and genome amplification with WGA-X (as described in
63 Stepanauskas et al. (24)). The 16S rRNA gene was PCR amplified from the genomes (primers
64 27F (25) and 1492R (26)) and sequenced at the Centre de Recherche at Université Laval
65 (Quebec City, QC, Canada) on an Applied Biosystems 3730xl DNA Analyzer (Foster City, CA,
66 USA). Obtained 16S rRNA sequences were annotated by BLAST with the SILVA rRNA
67 database (release 138) (27). For whole genome sequencing, libraries were prepared with a
68 Nextera XT DNA Library Prep Kit and sequenced on a MiSeq (Illumina) with MiSeq Reagent
69 Kit v3 (600 cycles, 2x300 base reads). Low quality reads and bases were trimmed with BBDuk

70 (minimum Phred quality score 15, minimum length 30 bp) and contaminant human reads were
71 removed with DeconSeq (28). Genomes were assembled with SPAdes (v3.13.1, settings --sc --
72 careful) (29) and screened for contamination using JGI's Kmer Frequency Analysis tool and by
73 read classification with Kaiju (v1.7.3) (7). Average nucleotide identity of the SAGs against other
74 SAGs and metagenome bins was calculated with FastANI (v1.3) (30). Genome annotation was
75 done as described for the metagenome.

76

77 iv. mRNA extraction, sequencing, and analysis

78 RNA was extracted in triplicate with a Zymo Research ZymoBIOMICS DNA/RNA Miniprep
79 Kit from approximately 3 g sediment per extraction. Extracted samples were then treated with
80 the Invitrogen Turbo DNA-free kit (Carlsbad, CA, USA) to remove contaminating DNA. The
81 treated samples were then pooled and concentrated with a New England Biolabs Monarch RNA
82 Cleanup Kit (Ipswich, MA, USA). Ribosomal RNA was depleted with a New England BioLabs
83 NEBNext rRNA Depletion Kit (Bacteria) and a sequencing library was prepared with a New
84 England BioLabs Ultra II RNA Library Prep Kit. The generated cDNA library was sequenced at
85 The Center for Applied Genomics at the Hospital for Sick Children (Toronto, ON, Canada) on a
86 NovaSeq 6000 (Illumina) with an S Prime 100-cycle flow cell (2x100 base reads). Reads were
87 trimmed with Trimmomatic (settings LEADING:3 TRAILING:3 SLIDINGWINDOW:4:15
88 MINLEN 50) and rRNA reads were removed with SortMeRNA (v4.2.0) (31). Human
89 contaminant reads were removed with BBDMap using the BBTools RemoveHuman masked
90 reference genome. Reads were classified with Kaiju (v1.7.3), and reads classified within
91 common sequencing contaminant groups as in Sheik et al. (2018) (32) were removed (~81,500
92 reads removed representing 77 genera; the genera with the most reads removed were

93 *Pseudomonas* (~12,000 reads), *Streptococcus* (~7000 reads), and *Ralstonia* (~6000 reads)).
94 Remaining reads were aligned to metagenome contigs and SAG scaffolds with bowtie2
95 (v2.3.5.1, with setting -very-sensitive-local) (33). Reads aligned to CDS regions were counted
96 with HTSeq (v0.12.4) (34) and transcripts per million reads (tpm) was calculated to normalize
97 expression values for each gene. For comparison of stress response genes in SAGs to those in
98 related genomes, protein-coding genes from reference genomes were queried against LH SAG
99 protein-coding genes with BLASTp using a cut-off e-value of 1e-15.

100

101 v. Sequencing data availability

102 Sequencing reads, metagenome, MAGs, and SAGs were deposited in NCBI under BioProject
103 PRJNA699472. JGI annotations of the metagenome and SAGs are available under GOLD Study
104 ID Gs0135943. SAGs were also deposited individually in JGI (Table S4).

105

106 vi. Gibbs free energy calculations

107 Gibbs free energy values were calculated using the method reported in Jones et al. (2018) (35)
108 with the following parameters based on measurements reported in Table S1: 278.1° K, pH 6.4,
109 7.39 M ionic strength, 1.02×10^{-7} M CH₄, 3.81×10^{-4} M NH₃, 8.61×10^{-6} M H₂S, 1.57×10^{-5} M
110 O₂, 2.31×10^{-5} M NO₃⁻, 1.04 M SO₄²⁻, 4.11×10^{-5} M CO₂, 3.16×10^{-7} M H⁺, 3.12×10^{-5} M NO₂⁻,
111 and 0.23 M Fe³⁺.

112

113 **Supplementary Results and Discussion**

114 *Discrepancy between recovered MAGs and SAGs*

115 Using the criteria of average nucleotide identity >95%, only one SAG was found to correspond
116 to MAGs. Typically, more overlap between the two datasets would be expected, particularly for
117 those taxa abundant in the microbial community (36); conversely, the SAGs appear to be
118 enriched for taxa at low abundance in the metagenome. In the following sections we will discuss
119 possible reasons as to why this occurred.

120

121 Firstly, multiple selection steps occurred during generation of the SAGs that may have excluded
122 or did exclude genomes from taxa abundant in the metagenome and MAGs. While 365 events
123 (cells) were sorted during single cell sorting, only the 95 wells with the greatest estimated
124 genome amplification (based on fluorescence during the amplification reaction) were selected for
125 further analysis. This may have resulted in a biased selection of cells: for example, selecting for
126 those genomes most responsive to the MDA reaction or those most resistant to environmental
127 fluxes that may have occurred during transportation or sorting. Additionally, after an initial
128 round of 16S rRNA Sanger sequencing, some wells with highly similar (>98%) 16S rRNA
129 sequences were excluded from subsequent genomic sequencing to maximize sequencing
130 coverage on the remaining wells. This included several *Halomonas* and *Desulfobulbaceae*
131 genomes represented by MAGs, therefore potentially reducing the number of SAGs that may
132 have corresponded to MAGs. As a result of these filtering steps, SAGs corresponding to MAGs
133 may have been excluded and low-abundance taxa including archaeal genomes may have been
134 enriched.

135

136 Secondly, 20% of reads map to the high- and medium-quality MAGs, due in part to stringent
137 quality control during binning (i.e. only contigs >5000 bp were binned). As a result, while the

138 MAGs broadly represent the taxonomic groups present in the metagenome, they represent only a
139 portion of the sum diversity. Therefore, this restrictive binning process in combination with the
140 relatively high number of SAGs corresponding to taxa at low abundance in the metagenome may
141 also have contributed to the discrepancy.

142

143 Thirdly, the samples collected for metagenomic analysis and for SAG sequencing were collected
144 in different years (2017 vs. 2019, respectively). This was necessitated by the small amount of
145 sediment removed during sampling to avoid disturbing the site in combination with the low
146 biomass of the sediment, requiring relatively high amounts of sample for processing. While the
147 physical and geochemical parameters of the spring have remained stable for the ~15 years that
148 we have studied the spring, including the measurements taken in 2017 and 2019, we can't
149 exclude the possibility of minor changes in the spring that could have affected the microbial
150 community sampled. Additionally, due to the low biomass and high salinity of the sediment, it is
151 difficult to extract DNA. Biases introduced by the differing processing steps during MAG and
152 SAG sequencing (for example, DNA extraction vs. separation of cells from sediment) may
153 therefore have had outside effects.

154

155 To summarize, we suggest several factors potentially contributing to the discrepancy between
156 MAGs and SAGs: 1) Selection during generation of SAGs, 2) Stringent binning criteria, and 3)
157 Differences in input samples and biases introduced by differing experimental procedures. We
158 conclude by noting that although there is a discrepancy between the MAGs and SAGs, there is
159 overlap between the two samples additional to what was discussed in the manuscript. In addition
160 to the corresponding SAG and MAG noted in the manuscript, two additional SAGs had 16S

161 rRNA sequences >98% identical to unbinned 16S rRNA sequences, and nearly all taxonomic
162 groups represented by the SAGs were also present in the metagenome reads and 16S sequences
163 as classified by kaiju and PhyloFlash (with the sole exception of Iainarchaeota reads).
164 Additionally, comparison of 16S rRNA sequences in the SAGs to previous 16S rRNA
165 sequencing from over ten years ago (1) identified common sequences (>98%) between the two
166 datasets, including sequences for *Halomonas*, ANME-1, and *Iainarchaeota*, suggesting that the
167 discrepancy is more likely due to the potential technical factors discussed above rather than
168 significant changes in the microbial community. The relative abundances of the microbial
169 community represented in the metagenome are consistent with those observed in previous
170 CARD-FISH and 16S rRNA and metagenomic sequencing (1, 2). We therefore suggest that the
171 metagenome and MAGs more accurately represent the taxa abundant in the microbial
172 community, whereas the SAGs are disproportionately enriched in low-abundance bacteria and
173 archaea due to the potential factors discussed above.

174

175

176

177

178

179

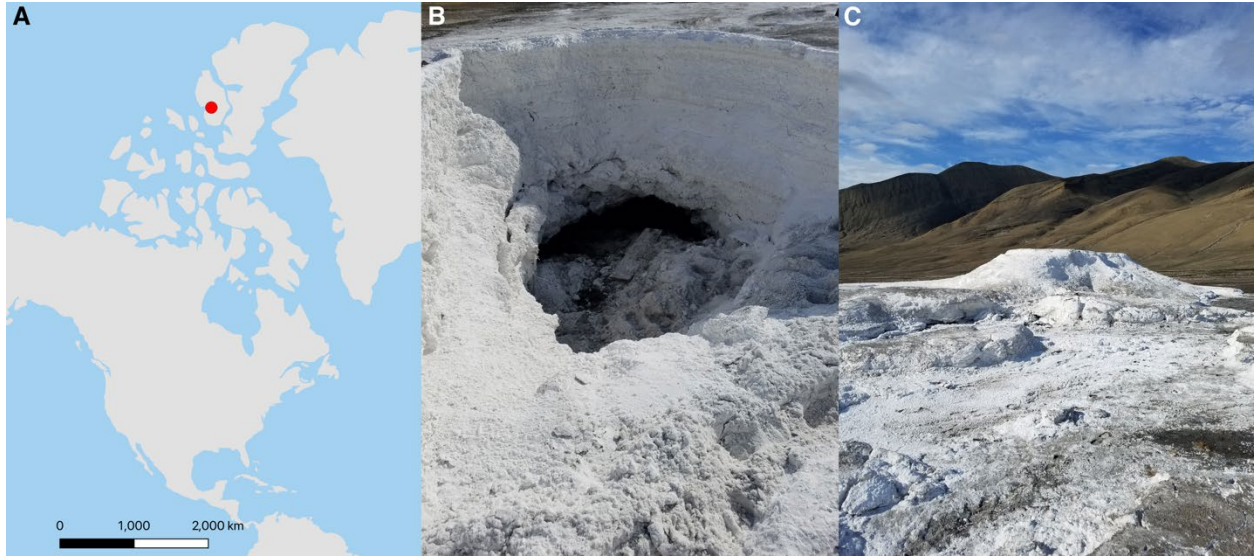
180

181

182

183

184 **Supplementary Figures and Tables**



185

186 **Figure S1.** A. Location of the LH spring on Axel Heiberg Island in the Canadian High Arctic.

187 Map generated in QGIS with the Natural Earth dataset. B. View looking into the LH salt tufa to

188 the spring source (July 2019). C. View of the LH tufa and outflow channels (July 2019).

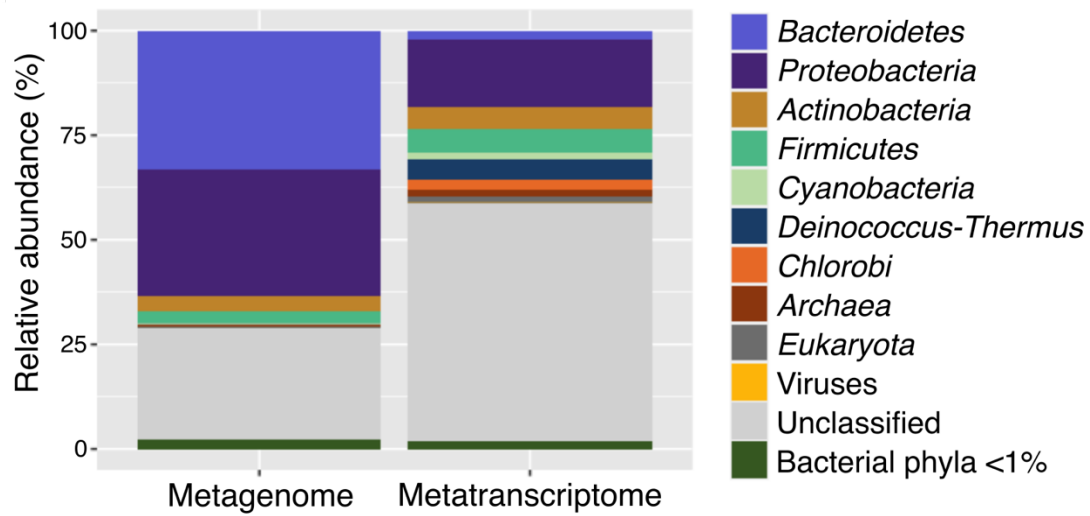
189

190

191

192

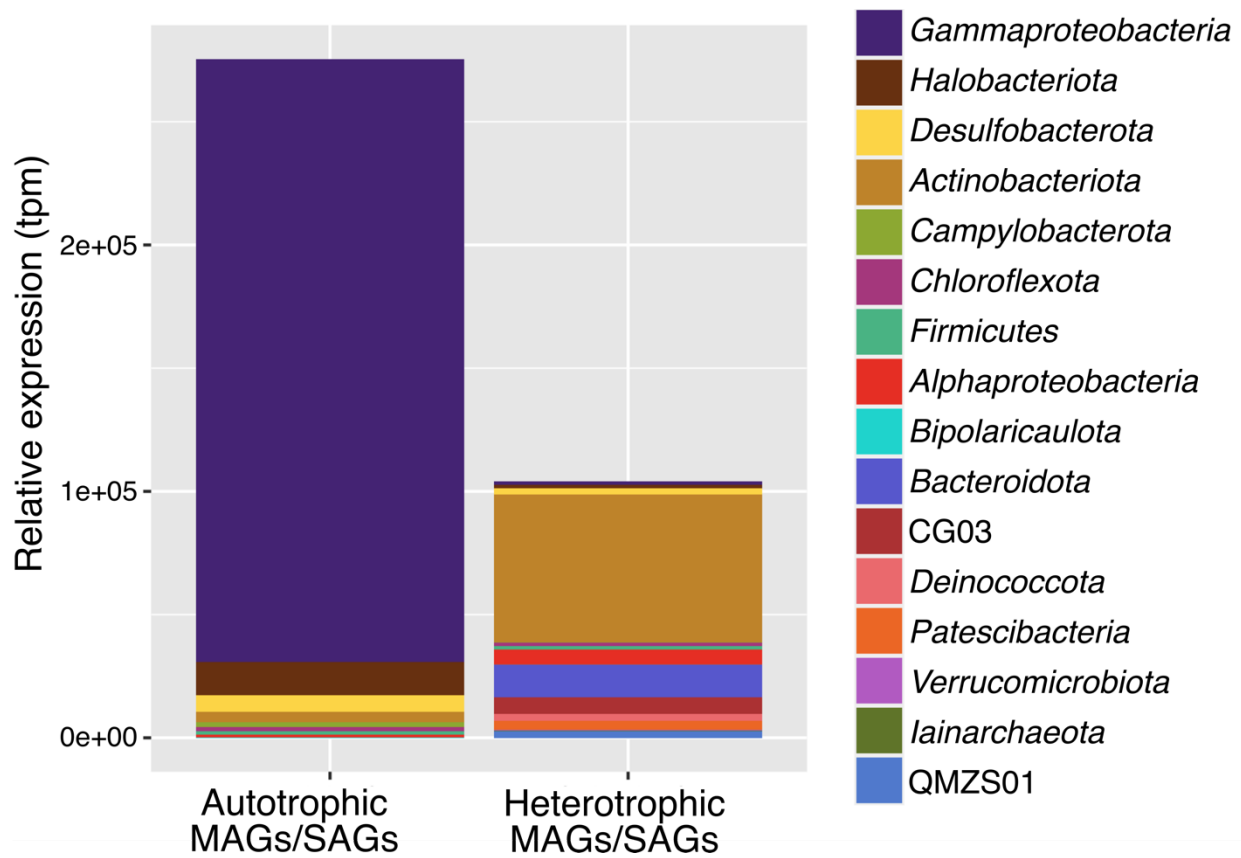
193



194

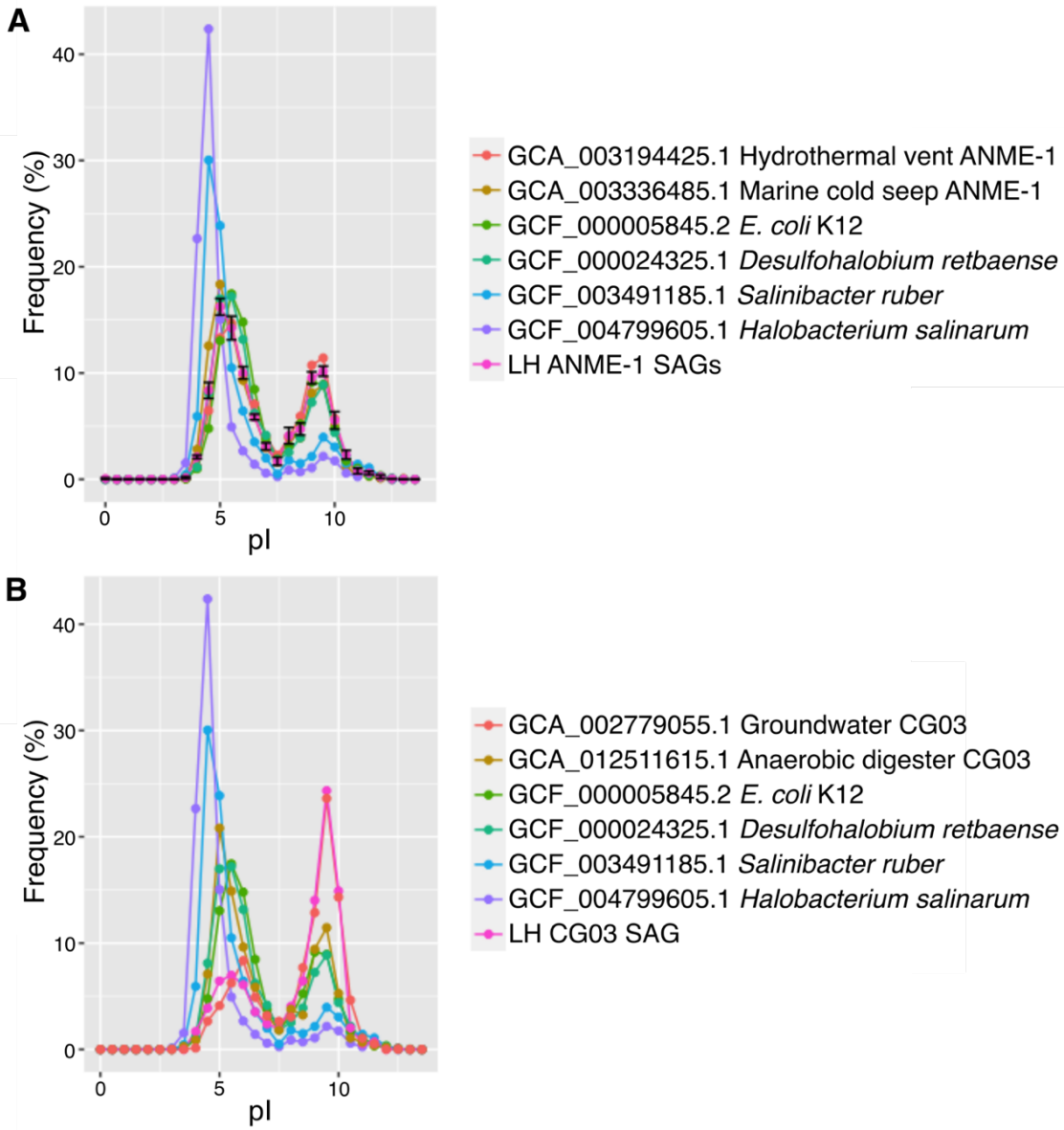
195 **Figure S2.** Taxonomic classification of metagenome and metatranscriptome reads. Reads were
 196 classified by Kaiju using the NCBI non-redundant database. Metagenome reads are an average of
 197 the relative abundance in duplicate samples.

198



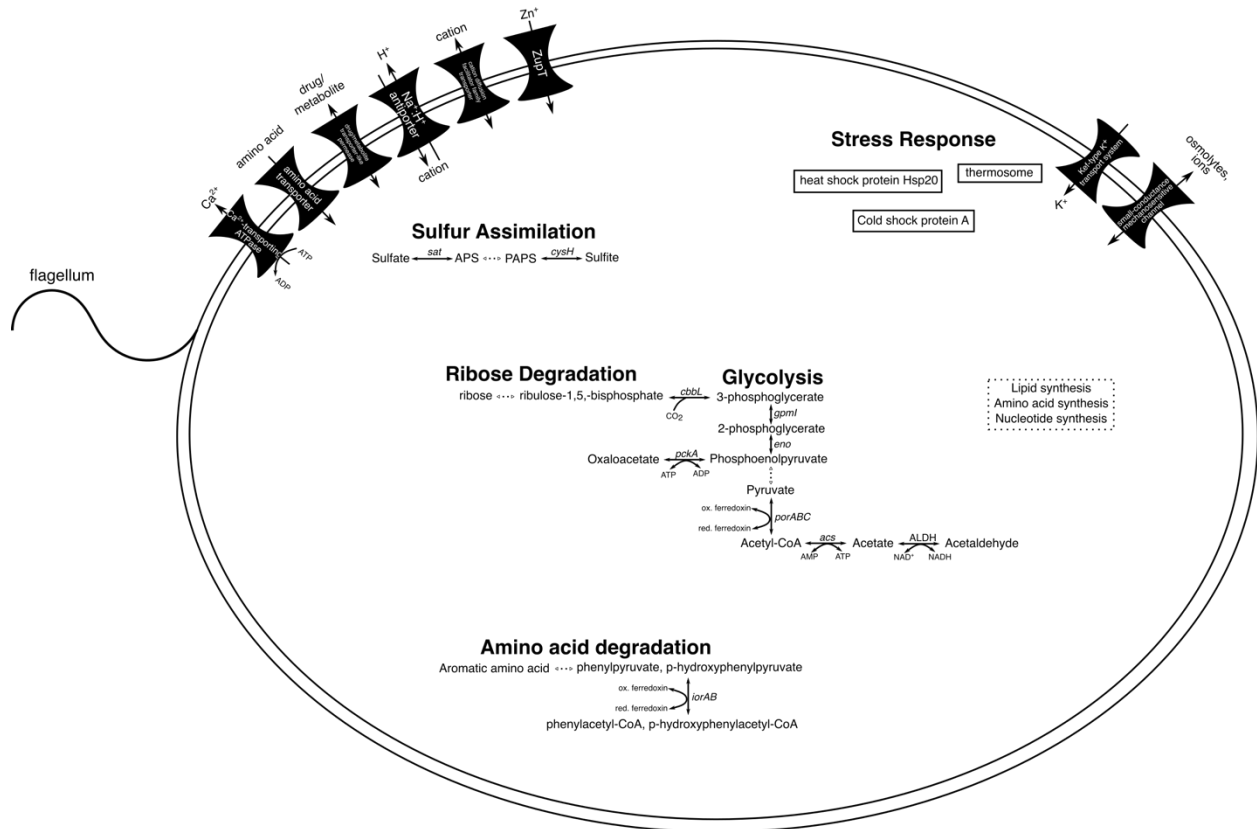
199

200 **Figure S3.** Relative expression (tpm) attributed to MAGs and SAGs containing CO₂-fixation
 201 genes (autotrophic MAGs/SAGs) and those without (heterotrophic MAGs/SAGs). Unbinned
 202 genes with mapped transcripts were omitted from this analysis.



203

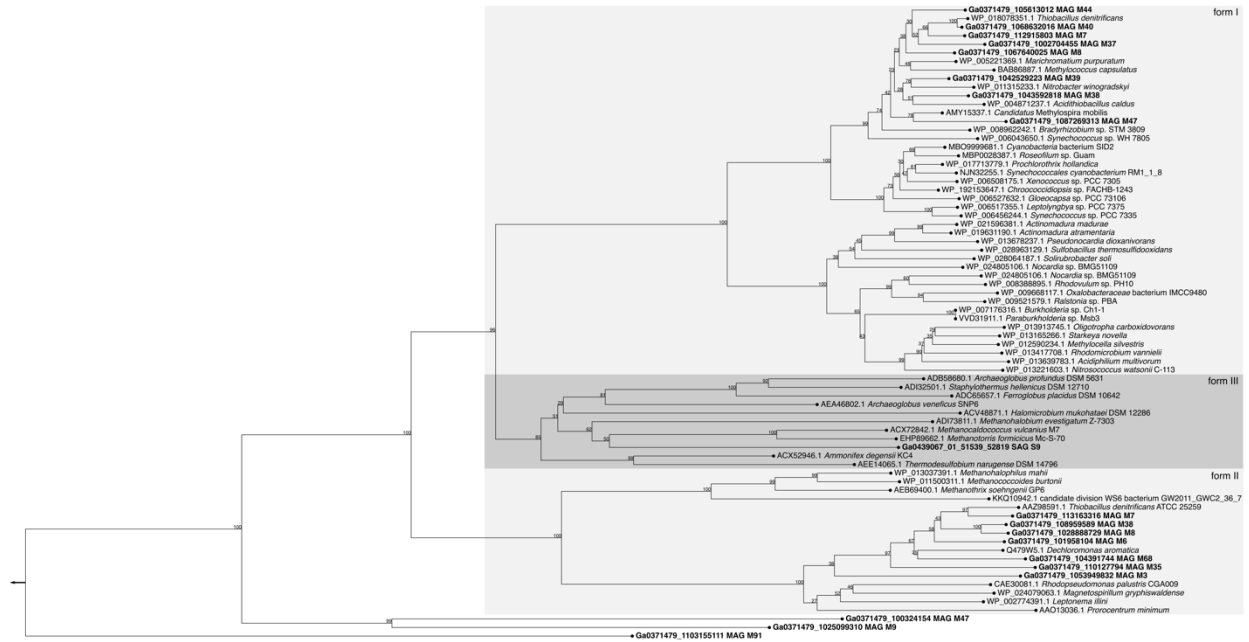
204 **Figure S4.** Comparison of predicted protein isoelectric points for A. ANME-1 SAGs S10-S14
 205 and B. CG03 SAG S2. Protein isoelectric points were calculated with the ExPASy Compute
 206 pI/MW online tool. Genomes for comparison (based on similar analysis in Nigro et al. (37))
 207 include microorganisms known to accumulate salts through the “salting-in” osmoregulation
 208 strategy (*Salinibacter ruber* and *Halobacterium salinarum*) and those that do not accumulate
 209 salts (*E. coli* K12 and *Desulfohalobium retbaense*).



210

211 **Figure S5.** Metabolic reconstruction of *Iainarchaeia* SAG S9 (670822 bp, 52.9% completeness,
 212 0% contamination). Solid lines represent genes present within the genome, and dashed lines
 213 indicate steps or pathways absent from the genome. Table S15 contains a complete list of
 214 annotated genes represented in this figure.

215

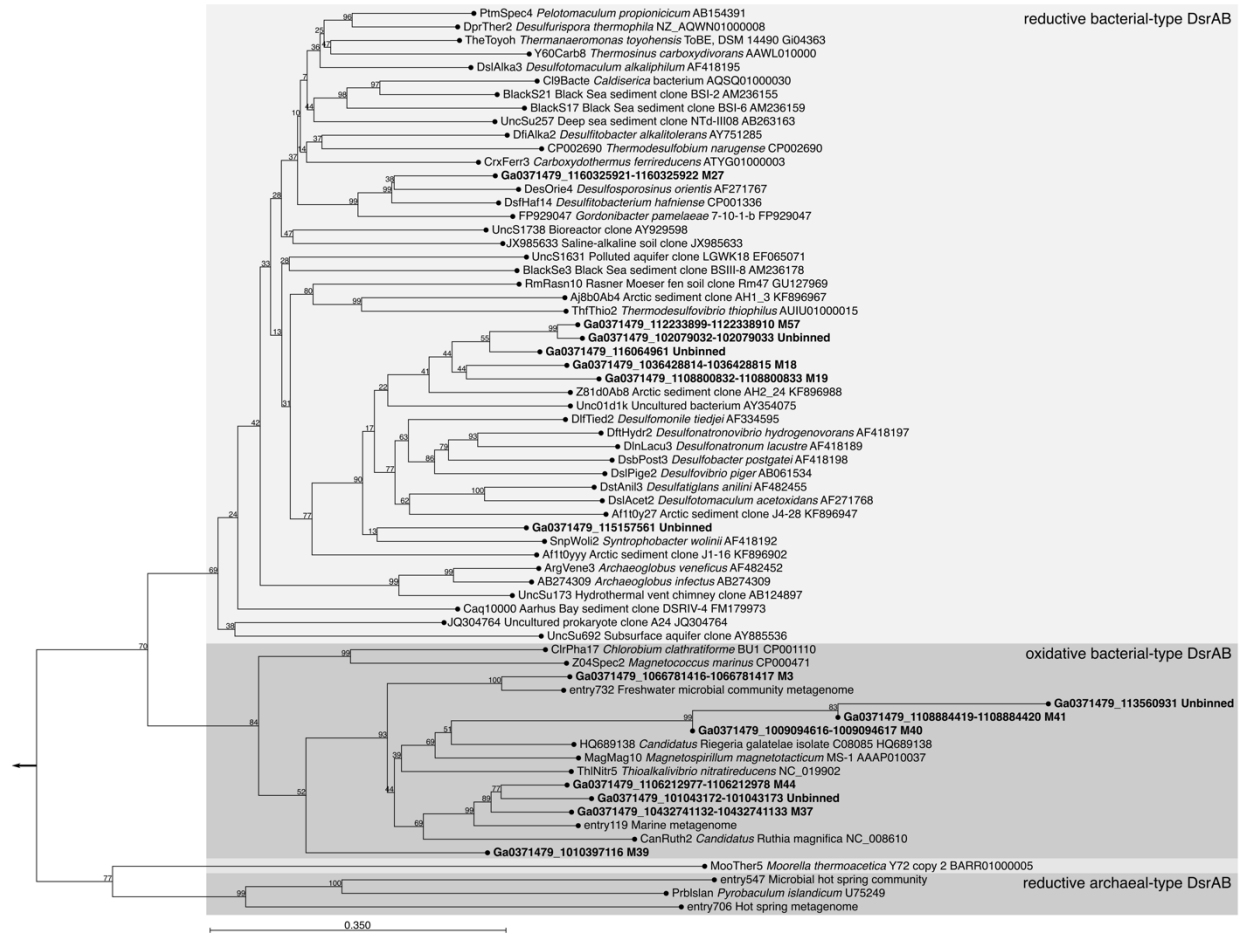


229

230 **Figure S7.** Maximum likelihood phylogenetic tree of RbcL/CbbL sequences. The tree was
 231 constructed in CLC Genomics Workbench with 1000 bootstraps and WAG substitution model
 232 using reference sequences from NCBI. A RubisCO-like protein sequence from
 233 *Rhodospseudomonas palustris* TIE-1 (ACF00976.1) was used as an outgroup (direction indicated
 234 by arrow).

235

236



244

245 **Figure S9.** Maximum likelihood phylogenetic tree of DsrAB. The tree was constructed in CLC

246 Genomics Workbench with 1000 bootstraps and WAG substitution model using reference

247 sequences from Muller et al. (2015). Sequence names include the gene IDs and MAG number

248 where applicable. A sequence from eggNOG group COG2221 (*Campylobacter ureolyticus*

249 JFJK01000015_gene476) was used as an outgroup (direction indicated by arrow).

250

251

252

253

254

255 **Table S1 (xlsx).** Physical and geochemical parameters of Lost Hammer water and sediment.

256 Located in Supplementary Tables xlsx file.

257

258 **Table S2.** Metagenome and metatranscriptome sequencing statistics.

DNA/RNA	Sample (NCBI ID)	Sequencing platform	Read length	Raw read pairs	Read pairs after quality control
DNA	SRR13628066	Illumina HiSeq2500	2x126	116844950	116746762
DNA	SRR13628065	Illumina HiSeq2500	2x126	107914225	107647735
RNA	SRR13633097	Illumina NovaSeq6000	2x100	9009952	4233149

259

260

261

262

263

264

265

266

267

268

269

270

271

272

273

274 **Table S3.** Metagenomic assembly statistics.

Assembler	Megahit v1.1.3
# contigs	1620755
# contigs >= 1000 bp	276578
# contigs >= 5000 bp	36050
# contigs >= 10000 bp	14841
# contigs >= 25000 bp	3946
# contigs >= 50000 bp	1251
Total length (bp)	1599970111
GC %	51.63
N50	2968
N75	1032
L50	67307
L75	264669
% reads mapping to contigs (average of two samples)	89.7
% reads mapping to contigs > 5000 bp	67.9
Average coverage of metatranscriptome against the metagenome (reads/base)	0.093 ± 59
Average coverage of metatranscriptome against CDS with mapped reads (reads/base)	13 ± 604

275
276
277
278
279
280
281
282
283

284 **Table S4 (xlsx).** SAG supplemental information. Located in Supplementary Tables xlsx file.

285

286 **Table S5 (xlsx).** MAG supplemental information. Located in Supplementary Tables xlsx file.

287

288 **Table S6 (xlsx).** Relative expression for KEGG, COG, and pfam IDs. Located in Supplementary
289 Tables xlsx file.

290

291 **Table S7 (xlsx).** Taxonomic classification of genes of interest with mapped transcripts. Located
292 in Supplementary Tables xlsx file.

293

294 **Table S8 (xlsx).** Gene presence and expression in medium-quality MAGs. Located in
295 Supplementary Tables xlsx file.

296

297 **Table S9 (xlsx).** Complete list of genes with mapped transcripts. Located in Supplementary
298 Tables xlsx file.

299

300

301

302

303

304

305

306

307 **Table S10.** Gibbs free energy of redox pairs present in LH. Values were calculated using the
 308 methodology in Jones et al. (2018) (35) (additional details in Supplementary Methods). These
 309 values should be considered preliminary estimates as some parameter concentrations are based
 310 on single-year measurements.

Redox pair	ΔG reaction (kJ/mol electron ⁻¹)
H ₂ /O ₂	-111.9
H ₂ /NO ₃ ⁻ (to NH ₃)	-61.3
N ₂ /NO ₃ ⁻ (to NO ₂ ⁻)	-68.1
H ₂ /SO ₄ ²⁻	-3.0
H ₂ /CO ₂	-9.1
H ₂ S/O ₂	-109.0
H ₂ S/NO ₃ ⁻	-93.8
Fe ²⁺ /O ₂	-16.0
NH ₃ /O ₂	-52.9
NH ₃ /NO ₃ ⁻	-90.8
NH ₃ /SO ₄ ²⁻	-56.5
CH ₄ /O ₂	-205.7
CH ₄ /NO ₃ ⁻	-32.4
CH ₄ /SO ₄ ²⁻	6.1

311
 312
 313
 314
 315
 316
 317
 318

319 **Table S11 (xlsx).** Stress response gene comparison to related genomes. Located in
320 Supplementary Tables xlsx file.
321
322 **Table S12 (xlsx).** Gene content of MAGs. Located in Supplementary Tables xlsx file.
323
324 **Table S13 (xlsx).** Gene content of SAGs. Located in Supplementary Tables xlsx file.
325
326 **Table S14 (xlsx).** ANME-1 composite genome gene content. Located in Supplementary Tables
327 xlsx file.
328
329 **Table S15 (xlsx).** *Iainarchaeia* sp. S9 gene content. Located in Supplementary Tables xlsx file.
330
331 **Table S16 (xlsx).** QMZS01 composite genome gene content. Located in Supplementary Tables
332 xlsx file.
333
334 **Table S17 (xlsx).** CG03 sp. S2 gene content. Located in Supplementary Tables xlsx file.
335
336 **Table S18 (xlsx).** Metagenome copy number and total relative expression of genes of interest.
337 Located in Supplementary Tables xlsx file.
338
339
340
341

342 **References**

- 343 1. Niederberger TD, Perreault NN, Tille S, Lollar BS, Lacrampe-Couloume G, Andersen D, et al.
344 Microbial characterization of a subzero, hypersaline methane seep in the Canadian High Arctic.
345 ISME J. 2010;4(10):1326-39.
- 346 2. Lay CY, Mykytczuk NC, Yergeau E, Lamarche-Gagnon G, Greer CW, Whyte LG. Defining
347 the functional potential and active community members of a sediment microbial community in a
348 high-Arctic hypersaline subzero spring. Appl Environ Microbiol. 2013;79(12):3637-48.
- 349 3. Lamarche-Gagnon G, Comery R, Greer CW, Whyte LG. Evidence of *in situ* microbial activity
350 and sulphidogenesis in perennially sub-0 degrees C and hypersaline sediments of a high Arctic
351 permafrost spring. Extremophiles. 2015;19(1):1-15.
- 352 4. Battler MM, Osinski GR, Banerjee NR. Mineralogy of saline perennial cold springs on Axel
353 Heiberg Island, Nunavut, Canada and implications for spring deposits on Mars. Icarus.
354 2013;224(2):364-81.
- 355 5. Ward MK, Pollard WH. A hydrohalite spring deposit in the Canadian high Arctic: a potential
356 Mars analogue. Earth Planet Sci Lett. 2018;504:126-38.
- 357 6. Bolger AM, Lohse M, Usadel B. Trimmomatic: a flexible trimmer for Illumina sequence data.
358 Bioinformatics. 2014;30(15):2114-20.
- 359 7. Menzel P, Ng KL, Krogh A. Fast and sensitive taxonomic classification for metagenomics
360 with Kaiju. Nat Commun. 2016;7(1):1-9.
- 361 8. Gruber-Vodicka HR, Seah BKB, Pruesse E. phyloFlash: rapid small-subunit rRNA profiling
362 and targeted assembly from metagenomes. mSystems. 2020;5(5):1-16.

363 9. Li D, Liu CM, Luo R, Sadakane K, Lam TW. MEGAHIT: an ultra-fast single-node solution
364 for large and complex metagenomics assembly via succinct de Bruijn graph. *Bioinformatics*.
365 2015;31(10):1674-6.

366 10. Nurk S, Meleshko D, Korobeynikov A, Pevzner PA. metaSPAdes: a new versatile
367 metagenomic assembler. *Genome Res*. 2017;27(5):824-34.

368 11. Kang DD, Froula J, Egan R, Wang Z. MetaBAT, an efficient tool for accurately
369 reconstructing single genomes from complex microbial communities. *PeerJ*. 2015;3:1-15.

370 12. Parks DH, Imelfort M, Skennerton CT, Hugenholtz P, Tyson GW. CheckM: assessing the
371 quality of microbial genomes recovered from isolates, single cells, and metagenomes. *Genome*
372 *Res*. 2015;25(7):1043-55.

373 13. Mikheenko A, Prjibelski A, Saveliev V, Antipov D, Gurevich A. Versatile genome assembly
374 evaluation with QUAST-LG. *Bioinformatics*. 2018;34(13):i142-i150.

375 14. Chen IA, Chu K, Palaniappan K, Ratner A, Huang J, Huntemann M, et al. The IMG/M data
376 management and analysis system v.6.0: new tools and advanced capabilities. *Nucleic Acids Res*.
377 2020;49(D1):D751–D763.

378 15. Mukherjee S, Stamatis D, Bertsch J, Ovchinnikova G, Sundaramurthi JC, Lee J, et al.
379 Genomes OnLine Database (GOLD) v.8: overview and updates. *Nucleic Acids Res*.
380 2020;49(D1):D723-D733.

381 16. Chaumeil PA, Mussig AJ, Hugenholtz P, Parks DH. GTDB-Tk: a toolkit to classify genomes
382 with the Genome Taxonomy Database. *Bioinformatics*. 2019;36(6):1925–7.

383 17. Eren AM, Esen OC, Quince C, Vineis JH, Morrison HG, Sogin ML, et al. Anvi'o: an
384 advanced analysis and visualization platform for 'omics data. *PeerJ*. 2015;3:1-29.

- 385 18. Garber AI, Nealson KH, Okamoto A, McAllister SM, Chan CS, Barco RA, et al. FeGenie: a
386 comprehensive tool for the identification of iron genes and iron gene neighborhoods in genome
387 and metagenome assemblies. *Front Microbiol.* 2020;11:1-23.
- 388 19. Fish JA, Chai B, Wang Q, Sun Y, Brown CT, Tiedje JM, et al. FunGene: the functional gene
389 pipeline and repository. *Front Microbiol.* 2013;4:1-14.
- 390 20. Sondergaard D, Pedersen CN, Greening C. HydDB: a web tool for hydrogenase classification
391 and analysis. *Sci Rep.* 2016;6:1-8.
- 392 21. Muller AL, Kjeldsen KU, Rattei T, Pester M, Loy A. Phylogenetic and environmental
393 diversity of DsrAB-type dissimilatory (bi)sulfite reductases. *ISME J.* 2015;9(5):1152-65.
- 394 22. Edgar RC. MUSCLE: multiple sequence alignment with high accuracy and high throughput.
395 *Nucleic Acids Res.* 2004;32(5):1792-7.
- 396 23. Anantharaman K, Hausmann B, Jungbluth SP, Kantor RS, Lavy A, Warren LA, et al.
397 Expanded diversity of microbial groups that shape the dissimilatory sulfur cycle. *ISME J.*
398 2018;12(7):1715-28.
- 399 24. Stepanauskas R, Fergusson EA, Brown J, Poulton NJ, Tupper B, Labonte JM, et al.
400 Improved genome recovery and integrated cell-size analyses of individual uncultured microbial
401 cells and viral particles. *Nat Commun.* 2017;8(1):1-10.
- 402 25. Kuske CR, Barns SM, Grow CC, Merrill L, Dunbar J. Environmental survey for four
403 pathogenic bacteria and closely related species using phylogenetic and functional genes. *J*
404 *Forensic Sci.* 2006;51(3):548-58.
- 405 26. Miller CS, Handley KM, Wrighton KC, Frischkorn KR, Thomas BC, Banfield JF. Short-read
406 assembly of full-length 16S amplicons reveals bacterial diversity in subsurface sediments. *PLoS*
407 *One.* 2013;8(2):1-11.

- 408 27. Quast C, Pruesse E, Yilmaz P, Gerken J, Schweer T, Yarza P, et al. The SILVA ribosomal
409 RNA gene database project: improved data processing and web-based tools. *Nucleic Acids Res.*
410 2013;41:D590-596.
- 411 28. Schmieder R, Edwards R. Fast identification and removal of sequence contamination from
412 genomic and metagenomic datasets. *PLoS One.* 2011;6(3):1-11.
- 413 29. Bankevich A, Nurk S, Antipov D, Gurevich AA, Dvorkin M, Kulikov AS, et al. SPAdes: a
414 new genome assembly algorithm and its applications to single-cell sequencing. *J Comput Biol.*
415 2012;19(5):455-77.
- 416 30. Jain C, Rodriguez RL, Phillippy AM, Konstantinidis KT, Aluru S. High throughput ANI
417 analysis of 90K prokaryotic genomes reveals clear species boundaries. *Nat Commun.*
418 2018;9(1):1-8.
- 419 31. Kopylova E, Noe L, Touzet H. SortMeRNA: fast and accurate filtering of ribosomal RNAs
420 in metatranscriptomic data. *Bioinformatics.* 2012;28(24):3211-7.
- 421 32. Sheik CS, Reese BK, Twing KI, Sylvan JB, Grim SL, Schrenk MO, et al. Identification and
422 removal of contaminant sequences from ribosomal gene databases: lessons from the Census of
423 Deep Life. *Front Microbiol.* 2018;9:1-14.
- 424 33. Langmead B, Salzberg SL. Fast gapped-read alignment with Bowtie 2. *Nat Methods.*
425 2012;9(4):357-9.
- 426 34. Anders S, Pyl PT, Huber W. HTSeq--a Python framework to work with high-throughput
427 sequencing data. *Bioinformatics.* 2015;31(2):166-9.
- 428 35. Jones RM, Goordial JM, Orcutt BN. Low energy subsurface environments as extraterrestrial
429 analogs. *Front Microbiol.* 2018;9:1-18.

- 430 36. Alneberg J, Karlsson CMG, Divne AM, Bergin C, Homa F, Lindh MV, et al. Genomes from
431 uncultivated prokaryotes: a comparison of metagenome-assembled and single-amplified
432 genomes. *Microbiome*. 2018;6(1):1-14.
- 433 37. Nigro LM, Hyde AS, MacGregor BJ, Teske A. Phylogeography, salinity adaptations and
434 metabolic potential of the candidate division KB1 bacteria based on a partial single cell genome.
435 *Front Microbiol*. 2016;7:1-9.