### 1 Comparative metagenomics of coalbed methane microbial

### 2 communities reveals biogenic methane potential in the

### 3 Appalachian Basin

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23 Natural gas is a major source of global energy, and a large fraction is 24 generated in subsurface coalbed deposits. Microbial communities within 25 coalbed deposits impact methane production, and as a result contribute to 26 global carbon cycling. The process of biogenic coal-to-methane conversion 27 is not well understood. Here we demonstrate the first read- and assembly-28 based metagenome profiling of coal-associated formation waters, resulting 29 in the recovery of over 40 metagenome-assembled genomes (MAGs) from 30 eight individual coalbed methane wells in the Appalachian Basin. The 31 majority of samples contained hydrogenotrophic methanogens, which were 32 present in higher relative abundances than was previously reported for 33 other coalbed basins. The abundance of Archaea and salinity were 34 positively correlated, suggesting that salinity may be a controlling factor 35 for biogenic coalbed methane. Low-abundance coalbed microbial 36 populations were functionally diverse, while the most dominant organisms 37 exhibit a high degree of genomic and functional similarities. Basin-specific 38 pan-metagenome clustering suggests lower abundant and diverse bacterial 39 communities are shaped by local basin parameters. Our analyses show 40 Appalachian Basin coalbed microbial communities encode for the potential 41 to convert coal into methane, which may be used as an indicator of 42 potential biogenic methane production for future well performance and 43 increased well longevity.

45 Methane is an important fossil energy resource in the global energy landscape. 46 As conventional natural gas resources become depleted, unconventional gas 47 technologies will emerge with greater importance for global energy security. One 48 unconventional gas technology is coalbed methane (CBM), which relies on 49 underutilized natural methane repositories trapped in subsurface coalbeds. CBM 50 wells provide access to subsurface methane deposits at a reduced cost and 51 environmental impact relative to traditional mining practices <sup>1</sup>.

An estimated 40% of U.S.-based CBM is biogenic<sup>2,3</sup>, and recently, there 52 53 has been a growing interest in understanding microbial communities in coalbed 54 deposits, as these communities may be potential indicators of productive CBM wells or utilized to enhance *in situ* production of methane.<sup>4</sup> Research in this area 55 56 has offered insight into unique subsurface microbial pathways that affect 57 methane production and ultimately, the global energy supply and global carbon 58 Biological conversion of coal to methane is complex and requires cvclina. 59 multiple enzymatic steps, performed by a diverse set of microorganisms, methanogens<sup>5</sup>—unraveling 60 degraders and including hydrocarbon these pathways, and the associated diverse community of microorganisms, is 61 62 paramount to understanding biogenic methane production.

Much of what is known about coalbed microbial communities has been derived from 16S rRNA gene sequencing<sup>6–8</sup>, which provides important information about microbial community structure and diversity, but lacks functional data required for detailed metagenomic analyses. A limited number of studies have provided valuable insight into the metabolic potential of CBM

systems using metagenomics<sup>9,10</sup>. However, these studies are limited to the western (Alberta Basin, Powder River Basin, and the San Juan Basin) and interior basins (Illinois Basin), neglecting large coalbed basins in the eastern United States, specifically the Appalachian Basin. The Appalachian Basin is an energy rich expanse that stretches from New York to Alabama and according to the newest EIA report is one of the most productive energy regions in the nation (www.eia.gov).

75 To characterize the unknown microbial community in the eastern US coal 76 basin and gain insight in its potential for biogenic methane production, we 77 present the first metagenomic investigation of Appalachian Basin coalbeds and 78 the first coalbed-focused pan-genomic comparison across geographically distinct 79 regions. Findings from this study provide 1) a read-based and assembly-centric 80 taxonomic profiling of the previously uncharacterized microbial communities, 2) 81 the first Appalachian Basin coalbed metagenome-assembled genomes (MAGs), 82 functional potential characterization of coalbed microbial communities extant in 83 the Appalachian Basin, and 4) a taxonomic and functional comparison of 84 Appalachian Basin metagenomes to microbial communities from geographically 85 distinct coalbeds. Our results demonstrate the Appalachian Basin to be distinct 86 characterized specifically from previously basins, an abundance of 87 hydrogenotrophic methanogens from the Order *Methanomicrobiales*, suggesting 88 a novel subsurface environment that could be targeted for methane production 89 and contributes to unknown amounts of carbon cycling. We show abundant 90 populations were highly conserved across geographically distinct coalbed basins 91 while low-abundance microorganisms were functionally diverse within and across 92 coal basins. This work provides a framework for understanding subsurface 93 microbial metabolisms and the potential for coal to methane conversion, which 94 can be utilized as an indicator for the potential to recover methane from unmined 95 areas, and enable the development for microbial enhanced coalbed methane 96 strategies in the Appalachian Basin.

### 98 METHODS

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100 Sample collection and chemical characterization. Formation water samples 101 (*i.e.*, water from CBM wells produced after drilling) were collected from eight 102 separate coalbed methane wells from the Appalachian Basin. Samples were 103 extracted from CBM wells named Key8, Key7, K34, K35, P21, BB137, MC79, 104 and L32A, at a depth of 796 ft., 1,201 ft., 1,704 ft., 1,912 ft., 1,961 ft., 1,980 ft., 105 2,239 ft., and 2,578 ft., respectively. All samples were obtained from the 106 Pocahontas 3 coal seam in December 2015 (Buchanan County, VA; Tazewell 107 County, VA; McDowell County, WV) with the exception of Key8 (Middle 108 Kittanning) and Key7 (Upper Kittanning/Brookville "A"), which came from Indiana 109 County, PA and Westmoreland County, PA, respectively. Formation water 110 samples were collected in sterile 1L bottles, placed on ice (immediately following 111 extraction from the well), and frozen upon arrival at NETL. The coal rank 112 associated with production water ranged from low/medium volatile bituminous 113 (K34, K35, P21, BB137, MC79, and L32A) to medium/high volatile bituminous 114 (Key7 and Key8). Cumulative Appalachian Basin coalbed methane production ranged from 10 billion cubic feet (BCF) to 1 trillion cubic feet<sup>11</sup>. 115

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Major and trace cations and anions (Na, Ca, Mg, K, Fe, Sr, Ba, Li, Mn, and Cl)
were examined at the Pittsburgh Analytical Laboratory (National Energy
Technology Laboratory) following EPA methods 6010C and 300.1 using ICPOES (Perkin Elmer Optima 7300 DV) and IC (Dionex Ion Chromatograph).
Samples were run in duplicate, with average values being recorded
(Supplementary Table 1).

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Read-based microbial abundance correlation. Read-based taxonomy
 (Archaea and Bacteria relative abundance) was correlated to major and minor
 cation/anion concentrations with Spearman rank coefficients using vegan in
 R<sup>12,13</sup>.

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129 DNA extraction, metagenome sequencing, and assembly. Aliquots of 130 formation water (50 mL) were filtered through a 0.2 µm filter and DNA was 131 extracted from the filter using the MoBio PowerSoil or PowerWater DNA Isolation 132 Kits (Carlsbad, CA). For shotgun metagenome sequencing, DNA libraries were 133 prepared using the Nextera XT DNA Library Preparation Kit according to 134 manufacturer's protocol (Illumina, San Diego, CA). Paired-end sequencing reads 135 (2 x 300 bp) were generated on an Illumina MiSeq with the MiSeq v3 Reagent Kit 136 (600 cycles) (Illumina, San Diego, CA). Initial assessment of reads was 137 performed with Kaiju using default settings: nr reference database, no SEG low 138 complexity filter, greedy run mode, minimum match length = 11, minimum match

allowed mismatches =  $5^{14}$ : MG-RAST<sup>15</sup>: GraftM 139 score = 70, and 140 (https://github.com/geronimp/graftM); and Phylopythia (Supplementary Table 2)<sup>16</sup>. Paired end reads were assembled using metaSPAdes version 3.8.0<sup>17</sup> with 141 142 error correction and with k-mer values of 33, 55, and 77. Assembly quality metrics (e.g., GC-content, N50, number of contigs, longest contig) were 143 assessed using QUAST<sup>18</sup> (Supplementary Table 3; Supplementary Figure 1). 144 145 Percentage of reads mapping to metagenome contigs was performed with the 146 'map reads to contigs' tool in the CLC Genomics Workbench version 8.5.1 (CLC Bio, Aarhus, Denmark). Annotation of assembled metagenome contigs was 147 performed with PROKKA<sup>19</sup> and MG-RAST<sup>15</sup>. 148

149 Contigs generated from metagenome assembly were binned using reference-independent methods (VizBin<sup>20</sup>). Completeness of metagenome-150 assembled genomes (MAGs) was assessed with CheckM, using single-copy 151 marker genes<sup>21</sup>. The lineage\_wf was initially employed, and combined with 152 153 BLASTn and Phylopythia outputs to further interrogate MAG quality using the 154 user-defined taxonomy\_wf. For example, bins defined at the phylum level (e.g., 155 *Euryarchaeota*) using the lineage\_wf were analyzed further and based upon 156 sequence similarity to Methanocalculus or Methanoregula, order-level 157 taxonomy wf was used to assess completeness and contamination using the Methanomicrobiales marker gene set. MAGs generated from contig binning were 158 159 annotated using the RAST annotation server<sup>22</sup> and PROKKA<sup>19</sup>. Manual curation of MAGs was performed using output from CheckM and RAST. Quality 160 161 designation for metagenome assembled genomes (MAGs) or population genomes was based upon metagenome-assembled genome (MIMAG)<sup>23</sup> 162 163 standards developed by the Genomic Standards Consortium (GSC), with high-164 guality draft genomes having >90% completion, <5% contamination, with 23S, 165 16S, and 5S rRNA genes and >18 tRNAs. Medium-guality draft genomes were 166 characterized by having >50% completion, and <10% contamination, while low-167 guality draft genomes were <50% complete and <10% contamination.

DNA sequences for metagenome samples from the San Juan basin (CG7, 168 169 CG8, CG13, and CG19) and Alberta Basin (CO182, CO183, and Trident1560D) 170 were downloaded from the Hydrocarbon Metagenomics Project (HMP) database 171 (http://hmp.ucalgary.ca/HMP/). Full details of sample collection, DNA extraction, 172 and sequencing can be found online (http://hmp.ucalgary.ca/HMP/) and has been documented by An and coworkers<sup>9</sup>. Raw read statistics are shown in 173 174 Supplementary Table 4 and 5. Metagenomes can also be found at the 175 metagenomics analysis server (MG-RAST) under Hydrocarbon the Metagenomics Project (mgp2360). Produced water samples CG7, CG8, CG13, 176 177 and CG19 from the San Juan Basin, and Trident1560D from the Alberta Basin were sequenced using 454-pyrosequencing. Samples CO182 and CO183 fromthe Alberta Basin were sequenced with Illumina.

180 Average nucleotide identity (ANI) and average amino acid identity (AAI) 181 were calculated with the enveomics collection toolbox<sup>24,25</sup>.

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Pan-metagenome analysis. Pan-metagenomics was performed with the
 Bacterial Pan Genome Analysis Tool (BPGA) version 1.0.0<sup>26</sup>. Default BPGA
 parameters were employed, which included USEARCH clustering at 50%
 sequence identity cutoff<sup>27</sup>.

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188 **Data availability.** The datasets generated during and/or analyzed during the 189 current study are available on the Energy Data Exchange (EDX) page 190 (https://edx.netl.doe.gov/dataset/appalachian-basin-metagenome-sequencing-

- 191 <u>reads</u>).
- 192

### 193 **RESULTS**

# Read-based taxonomy and functional profiling of Appalachian Basin formation water metagenomes.

196 Formation water collected from eight separate coalbed methane wells was 197 utilized for metagenomic analysis and geochemical characterization. Taxonomic 198 profiling of metagenome reads revealed that Appalachian Basin metagenomes 199 contained varied proportions of Archaea and Bacteria, with the relative 200 abundance ranging from 1-81%, and 19-99%, respectively (Figure 1). The 201 relative abundance of Archaea was >48% for five of the eight metagenomes, 202 areater than what has been found in other coalbed basins<sup>6-8</sup>, but similar to Archaea abundances in crude oil reservoirs<sup>28</sup>. 203

204 Environmental factors, such as salinity, coal rank, and depth, are known to 205 play a key role in shaping the taxonomic distribution of subsurface microbial communities<sup>10,29–31</sup>. Coalbed-associated formation water geochemistry was 206 207 examined and varied from moderately to highly saline, with concentrations of 208 major ions (Na, Cl, Mg, and Ca) ranging from 7.29 g/L (Kev8) to 130.30 g/L 209 (K34), (Figure 1; Supplementary Table 2). There was a positive correlation 210 between Archaea abundances and salinity and a negative correlation between 211 Bacteria abundances and salinity (Figure 1, Supplemental Table 2) in the 212 Appalachian Basin samples. No observable trends were found between microbial 213 community composition and associated coal rank (*i.e.*, maturity) or depth.

Utilization of coal as a carbon and energy source requires breakdown of complex polymers into fermentable substrates, a poorly understood process and hypothesized to be the rate-limiting step in the conversion of coal to methane<sup>5</sup>. It is thought that hydrocarbon-activating organisms (*e.g.*, Deltaproteobacteria and Firmicutes) are responsible for the initial attack on coal<sup>32</sup>. Examination of metagenome reads revealed the presence of hydrocarbon degradation pathways in all samples, though samples containing a higher abundance of Bacteriaspecific sequences had a higher proportion of hydrocarbon degradation pathways (Figure 1; Supplementary File 1).

223 Samples containing an abundance of Bacteria-specific reads were 224 predominantly comprised of Proteobacteria and despite phylum-level similarities 225 across Appalachian Basin metagenomes, noteworthy distinctions were apparent 226 at the class level. For example, metagenome L32A contained mostly 227 Alphaproteobacteria (14%), metagenomes K35 and MC79 were comprised 228 predominantly of Gammaproteobacteria (52% and 46% Pseudomonas, 229 respectively), and Key8 was comprised predominantly of Deltaproteobacteria 230 (22% Desulfobacterium) (Figure 2). Within the Proteobacteria phylum, 231 Alphaproteobacteria and Deltaproteobacteria were most diverse across 232 Appalachian Basin samples, while Betaproteobacteria, Gammaproteobacteria, 233 and Epsilonproteobacteria were predominantly comprised of Burkholderiales, 234 Pseudomonadales or Alteromonadales, and Campylobacterales, respectively. 235 Four of the eight metagenomes containing the lowest relative abundances of 236 Bacteria (<40%) had a higher proportion of members from the Terrabacteria 237 group (Firmicutes, Chloroflexi, and Actinobacteria), FCB group (Bacteroidetes), 238 and Synergistales.

239 Closer examination of Archaea-specific reads revealed a predominance of 240 Methanomicrobiales in the majority of Appalachian Basin samples, with relative 241 abundances reaching 69% (Figure 2). Members of the order 242 Methanomicrobiales are hydrogenotrophic methanogens that utilize  $H_2/CO_2$  and 243 formate, and inhabit a variety of environments, including marine and freshwater 244 sediments<sup>33</sup>, and coalbed environments across the United States and 245 abroad<sup>6,7,34</sup>. Other less-abundant Archaea orders represented in metagenome 246 samples included Methanobacteriales and Methanosarcinales (Figure 2).

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# Reconstruction of Appalachian Basin metagenome-assembled genomes(MAGs)

250 Assembly-based metagenome profiling was performed to obtain genome-251 resolved metabolisms of eight coalbed-associated microbial communities. 252 Approximately 40 medium- to high-guality MAGs were recovered, with 77-94% of 253 reads mapping to the metagenome assembled contigs. From the most abundant Methanomicrobiales 254 taxa recovered ten (Methanocalculus we and 255 Methanoregula), four Gammaproteobacteria (Pseudomonas and Shewanella), 256 and two Deltaproteobacteria (Desulfobacteraceae and Desulfobulbaceae) 257 population genomes (Supplementary Table 7).

Methanomicrobiales. Of the ten Methanomicrobiales MAGs, six were most closely related to halotolerant and hydrogenotrophic methanogens from hydrocarbon resource environments (oil reservoirs)—specifically, *Methanocalculus* sp. 52\_23, *Methanomicrobiales* archaeon 53\_18, and 262 *Methanocalculus halotolerans* (Supplementary Figure 2-4). Four were most 263 closely related to *Methanoregula formicicum* (Supplementary Table 8; 264 Supplementary Figure 2-5).

265 Methanomicrobiales MAGs were highly similar, with ANI and AAI values 266 >95%, suggesting that five of the six *Methanocalculus* MAGs belong to the same 267 species, and three of the four *Methanoregula* MAGs belong to the same species 268 (Supplementary Table 8). While samples contain acetoclastic methanogens, the 269 predominance of Methanomicrobiales suggests hydrogenotrophic 270 methanogenesis is operative in Appalachian Basin (Figure 2). Furthermore, 271 acetoclastic methanogenesis may be inhibited by coal degradation products, 272 which are similar in composition to crude oil, and contain substances that may 273 interfere with this microbial metabolism<sup>35</sup>.

274 Following assembly and manual curation of *Methanomicrobiales* MAGs, 275 comparative genomics was performed on the six *Methanocalculus* population 276 genome bins (Supplementary Figure 5) to identify the functional similarities of the 277 predominant methanogens across the Appalachian Basin. The core genome 278 (predicted protein sequences common among all genomes examined) contained 279 1583 gene families (representing 44% of the total gene families) with 69% of the 280 representative sequences most closely related to Methanocalculus sp. 52\_23, 281 22% most closely related to *Methanomicrobiales*, 2.3% most closely related to 282 Methanoculleus, and 1.4% most closely related to Methanoregula. Analysis of 283 the Methanocalculus core genome revealed complete pathways for carbon 284 fixation (reductive pentose phosphate cycle, and incomplete reductive citrate 285 cycle), methane metabolism [methanogenesis (CO<sub>2</sub> to methane, and acetate to 286 methane), F420 biosynthesis, and acetyl-CoA pathway (CO<sub>2</sub> to acetyl-CoA)], 287 branched chain amino acid metabolism, ATP synthesis (V-type ATPase), and 288 transport systems (molybdate, tungstate, glycine betaine/proline, 289 osmoprotectant, phosphate, iron, zinc, cobalt/nickel, and lipooligosaccharide) 290 (Supplementary File 2 and 3).

**Gammaproteobacteria.** Four MAGs related to Gammaproteobacteria were recovered from metagenomes K35 and Key8 (*Pseudomonas stutzeri*), and MC79 (*Shewanella putrefaciens*, and *Pseudomonas libanensis*). *Pseudomonas* sp. K35 and *Pseudomonas* sp. Key8 are high-quality population genomes with a high degree of similarity to one another (ANI = 97.88%), and most closely related to *Pseudomonas stutzeri* CCUG29243<sup>36</sup>. Prior work has described a high abundance of *Pseudomonas* in coalbed systems<sup>9,10,37,38</sup>.

Formation water from wells Key8 and K35 were moderately saline (7-18 g/L). One mechanism of salinity tolerance employed by bacteria involves production of compatible solutes to maintain turgor pressure and water content within the cell<sup>39</sup>. *Pseudomonas* sp. K35 and *P.* sp. Key8 genomes contain the *ectABCD-ask* gene cluster, which encodes for biosynthesis of the compatible solute, hydroxyectoine<sup>40</sup>.

*Pseudomonas* may play an important role in coal solubilization as many pseudomonads have been reported to produce a variety of rhamnolipids surface-active agents (surfactants) with the ability to solubilize hydrophobic carbon sources<sup>41–43</sup>. Production of rhamnolipids is controlled by many factors,

including temperature, salinity, pH, and nutrient availability<sup>43,44</sup> and may be 308 309 induced by coal or coal substituents<sup>42</sup>. Both *Pseudomonas* sp. K35 and Key8 310 contain the rmIBDAC operon for I-rhamnose genomes biosynthesis 311 (Supplemental Text), suggesting a potential for rhamnolipid-mediated coal 312 solubilization in the Appalachian Basin. The solubilized coal constituents may be 313 utilized directly by Pseudomonas or may provide fermentative substrates for other community members<sup>5</sup>. 314

315 Polycyclic aromatic hydrocarbons (PAHs) are a major constituent in 316 coalbed production water and PAH-degrading microorganisms are prevalent in 317 coal reservoirs<sup>38,45</sup>. *Pseudomonas* sp. K35 encodes for the upper and lower naphthalene degradation pathways and has the potential to degrade PAHs<sup>36</sup>. 318 319 While we did not find naphthalene degradation pathways in *Pseudomonas* sp. 320 Key8, other community members in the Key8 metagenome could perform this 321 function, as the Key8 metagenome encodes for hydrocarbon degradation 322 pathways (Supplementary File 1).

323 The other abundant Gammaproteobacteria genome bin was most closely 324 related to Shewanella, with ~17% of metagenome reads mapping to the genome 325 bin. The Shewanella population genome was 96% complete, and based on 326 whole genome comparisons, was most closely related to Shewanella 327 putrefaciens (ANI = 98.57%) (Supplementary Text, Supplementary Table 7). The 328 genus Shewanella contains metabolically versatile facultative anaerobes best 329 known for the dissimilatory reduction of metals (e.g. iron, manganese, and elemental sulfur)<sup>46</sup> and Shewanella spp. have been isolated from coal 330

environments<sup>47</sup>. The Appalachian Basin coalbed *Shewanella* population genome bin encoded for genes involved in assimilatory sulfate reduction, *c*-type cytochrome biogenesis (required for heme proteins involved in dissimilatory iron and manganese reduction), and nitrate reduction (Supplemental Text).

335 Deltaproteobacteria Deltaproteobacteria. predominant in were 336 metagenome Key8, which contained two population genome bins related to 337 Desulfobacteraceae and Desulfobulbaceae. While population genome bins could 338 only be resolved to the family level, recovery of 16S rRNA genes from 339 metagenome assemblies revealed the presence of near-complete sequences 340 related to Desulfocapsa sulfexigens, Desulfobacula toluolea, and an Uncultured 341 Desulfobacterium. Both families (Desulfobacteraceae and Desulfobulbaceae) 342 contain known sulfate reducers, and annotation of the genome bins revealed the 343 metabolic potential for sulfate, sulfite, and sulfur reduction (Supplementary Text). 344 The presence of Deltaproteobacteria in metagenome Key8 combined with the 345 fact that produced water from Key8 was the only sample with significant amounts 346 of sulfate (1.45 g/L) (Supplementary Table 5), suggests that these 347 microorganisms may play a role in generating or removing sulfate.

Low abundance bins. Population genome bins from lower abundant taxa were recovered and include Alphaproteobacteria (*Roseovarius, Marinovum, Rhizobium,* and *Hyphomonas*), Clostridia (*Acetobacterium*), Methanomicrobia (*Methanosaeta* and *Methanospirillum*), Gammaproteobacteria (*Marinobacter*), and Epsilonproteobacteria (*Arcobacter* and *Sulfurospirillum*). In general, Alphaproteobacteria are capable of aromatic compound degradation and

fermentation, and may provide the necessary substrates for methanogenesis<sup>5,48</sup>; 354 355 The Roseovarius and Hyphomonas MAGs weren't well resolved but contained 356 fragments of aromatic hydrocarbon degradation pathways (Supplemental Text): Acetobacterium has previously been found in coalbed environments<sup>9,10</sup>, and is an 357 358 acetogen capable of CO<sub>2</sub> fixation using the Wood-Ljungdahl pathway (Supplemental Text)<sup>49</sup>; Methanosaeta spp. are acetoclastic methanogens that 359 360 utilize acetate for methanogenesis; and Marinobacter has been found in 361 hypersaline subsurface environments and degrade alkanes, BTEX-N, and aliphatic compounds (*e.g.*, hexadecane)<sup>50,51</sup>. 362

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## Intra- and inter-basin assembly-based functional profiling of Appalachian Basin metagenomes

366 Microbial pan-metagenomics was employed to examine the metabolic potential 367 of the Appalachian Basin metagenomes and the functional intra-basin 368 similarities. The functional relatedness of metagenomes (core metagenome) 369 describes the pathways and potential driving factors common in coal 370 while the functional dissimilarity (accessory and unique environments. 371 metagenome) provides insight into how microbial communities are shaped by 372 site-specific physicochemical factors. Analysis of protein sequences predicted 373 from metagenome-assembled DNA contigs across all eight coalbed methane 374 metagenomes, revealed a total of 429,800 predicted genes for the Appalachian 375 Basin pan-metagenome. Genes were grouped into functional protein families 376 based upon sequence identity (50% identity cutoff) yielding 329,564 total gene families, 52,470 accessory gene clusters, 276,720 unique gene clusters, and 374
core gene clusters (Supplementary Table 9).

379 The core metagenome of the Appalachian Basin represented 0.11% of the 380 pan-metagenome, suggesting high diversity amongst microbial communities 381 within the Appalachian coal basin. The diversity of microbial communities may be 382 a direct result of the inherent complexity and heterogeneity of coalbed systems. 383 Structurally coal seams are comprised of a cleat system connecting small pore 384 spaces (<50 nm in diameter) and depending on the coal rank, have low 385 permeability. Thus the physicochemical properties of coalbeds create micro-386 niches where microbial communities become physically separated and evolve 387 into functionally distinct entities.

388 Annotation of the core genome revealed sequences were predominantly 389 related to Methanomicrobiales (Methanocalculus, Methanoregula, and 390 *Methanoculleus*). Other core metagenome sequences were most closely related 391 to Methanobacteriales, Methanosarcinales, Desulfovibrionales and 392 Pseudomonadales (Supplementary Table 10). Specifically, conserved pathways 393 included hydrogenotrophic and acetoclastic methanogenesis, amino acid 394 biosynthesis, and oxidative stress. Therefore, while nutrient limitation and coal 395 heterogeneity may contribute to taxonomic and functional diversity of coal-396 degrading microorganisms, homogeneity of coal-degradation products (e.g., 397 intermediates for methanogenesis) may play a role in the conservation of the 398 aforementioned pathways (core metagenome).

399 The functional relatedness of Appalachian Basin metagenomes to other 400 coalbed basins was also examined using pan-metagenomics. Analyses were 401 aimed at determining how functionally related microbial communities from distinct 402 coalbed methane wells were, and how that might infer biogenic methane 403 production in coal from various basins. In total, fifteen metagenomes from three 404 geographically distinct coalbed basins (three from the Alberta Basin, eight from 405 the Appalachian Basin, and four from the San Juan Basin) were analyzed. A total 406 of 429,182 gene clusters (pan-metagenome) were identified—23 gene families 407 were common to all metagenomes (core metagenome), 353,745 were specific to 408 an individual metagenome (unique metagenome), and 75,414 were found in at 409 least one but not all metagenomes (accessory metagenome) (Supplementary 410 Table 9).

411 Pan-metagenome clustering (taking into account all predicted protein 412 sequences) of the Alberta, Appalachian, and San Juan Basin samples revealed 413 basin-specific grouping (Figure 2), suggesting the overall metabolic potential is 414 shaped by conditions unique to each coal region. The few exceptions were 415 metagenome CG7 from the San Juan Basin, metagenome Trident1560D from 416 the Alberta Basin, and metagenomes L32A and Key8 from the Appalachian Basin (Figure 2). Specifically, metagenome CG7 clustered with metagenome 417 418 Key8 and both contain similar taxonomic distributions of Pseudomonadales, 419 Clostridiales, and Bacteroidales, with little to no Archaea. The two remaining 420 metagenomes, Trident1560D and L32A clustered together, and compared to all 421 other coalbed methane samples contained lower abundances of 422 Betaproteobacteria. and higher abundances of Alphaproteobacteria 423 (Rhodobacterales and Rhizobiales). The clustering of the outliers appeared to be 424 driven by the presence or absence of shared taxonomic groups, which is similar 425 to trends observed by Lawson and coworkers, where habitat-specific taxonomic 426 distribution was shaped by coal rank and physicochemical conditions exhibiting 'patterns of endemism' in coalbed samples<sup>10</sup>. 427

428 Core metagenome clustering exhibited less basin-specific groupings, but 429 like the pan-genomic outlier clusters, aligned with relative abundances and 430 taxonomic distributions of Bacteria and Archaea (Supplemental Figure 10). For 431 example, formation water samples containing the highest amounts of Archaea 432 (Methanomicrobiales) grouped together (Key7, K34, BB137, and P21), as did 433 metagenomes comprised predominantly of Gammaproteobacteria 434 (Pseudomonas) (K35, CG7, CO182 and CO183). While Basin-specific clustering 435 in the pan-metagenome demonstrates a strong selecting force in conditions 436 unique to each coal region, the core taxonomy-specific metagenome clustering 437 reveals common pathways that may be taxonomically linked and evolve through 438 exposure to physicochemical properties common to all coal environments-for 439 coal-degradation products example, that feed into hydrogenotrophic 440 methanogenesis pathways.

The multi-basin core metagenome contained a high proportion of gene families having KEGG or COG hits to amino acid transport metabolism, nucleotide metabolism, or xenobiotics degradation and metabolism, suggesting these pathways may be important in all coalbed environments, especially considering the exposure to hydrocarbons associated with and generated from coal. Conversely, the unique and accessory gene families were highest in cell wall/membrane/envelope biogenesis, signal transduction mechanisms, and metabolism of cofactors and vitamins. Gene families in these functional categories were presumably abundant due to the predominating physicochemical properties (impacting environmental stresses and nutrient availability) in the different basins.

Based upon pan-metagenome analysis, the unique gene families were consistently high ranging from ~78% for the Alberta basin to ~89% for the San Juan Basin, which suggests coalbed metagenomes contain a high intra-basin metabolic diversity. These findings support previous metagenomic work in hydrocarbon resource environments (oil sands, oil fields, and coalbeds), where coalbed samples were the most diverse<sup>9</sup>.

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#### 459 **DISCUSSION**

460 Taxonomic characterization and pan-genome comparison of Appalachian Basin 461 coalbed formation water metagenomes, combined with genome-resolved 462 microbial metabolisms of subsurface coalbed systems, revealed functional 463 diversity within and across coalbed basins. Specifically, high-abundance 464 populations (Methanomicrobiales and Pseudomonadales) were highly similar 465 (low diversity), while low-abundance populations were much more diverse--a 466 pattern similar to other physicochemically distinct deep-sea hydrothermal vent microbial communities<sup>52</sup>. The high abundance of Methanomicrobiales suggests 467

468 ongoing biogenic methane production in Appalachian Basin coalbeds. 469 Furthermore, the co-occurrence of Methanomicrobiales with high salinity 470 suggests their adaptation to saline environments. Several samples encoded for 471 potential hydrocarbon degradation pathways, however the overall number of 472 genes related to these pathways was limited, especially in Archaea-dominated 473 samples. Results from this study provide a first insight of the subsurface 474 metabolisms of Appalachian Basin coalbed microbial communities, and reveal 475 valuable information on the potential for biogenic coal to methane conversion in 476 the Appalachian Basin. Additional analyses (e.g., metatranscriptomics, 477 metaproteomics) of coal cores with associated formation water are needed to 478 fully assess the real-time dynamics of microbial communities in coalbed systems. 479

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### 643 **FIGURES**.

Figure 1. Read-based taxonomy profiling, and geochemistry of eight
Appalachian Basin formation water samples. A) Class-level taxonomy of
coalbed methane metagenomes from Illumina sequencing reads. B) Major ions
found in formation water samples. C) Total number of reads mapping to
hydrocarbon degradation pathways. D) Depth profile of formation water samples.

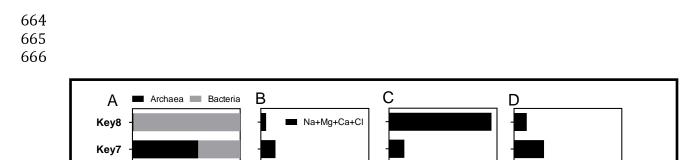
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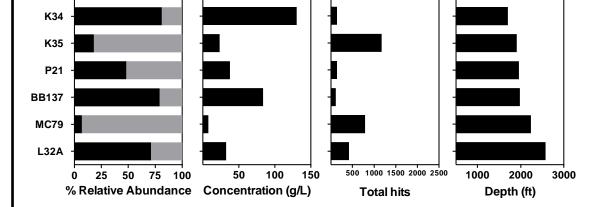
Figure 2. Assembly-based pan-metagenome analysis and read-based taxonomy of fifteen coalbed methane metagenomes from the Alberta Basin, Appalachian Basin, and San Juan Basin. Pan-metagenome clustering of coalbed methane metagenomes, with Bacteria and Archaea read distribution for Alberta Basin (yellow), Appalachian Basin (black), and San Juan Basin (red). Numbers on the pan-genome tree represent the number of gene families common across samples.

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Figure 3. Metabolic overview and potential functional coal-to-methane conversion in coalbed methane systems. Metagenome data suggests two coalbed regimes; one dominated by hydrocarbon degraders, the other dominated by methanogens.

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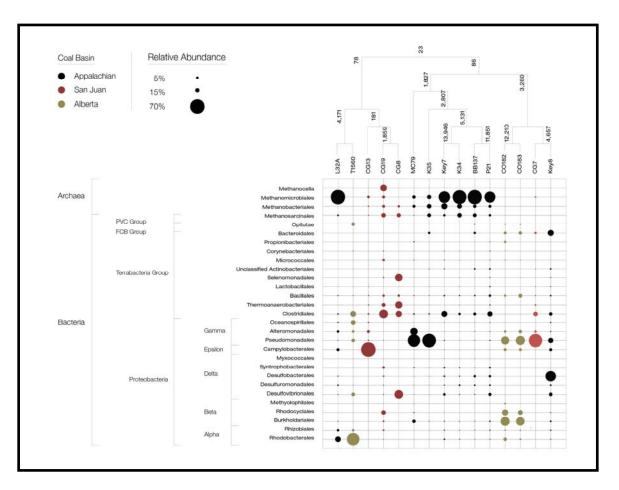




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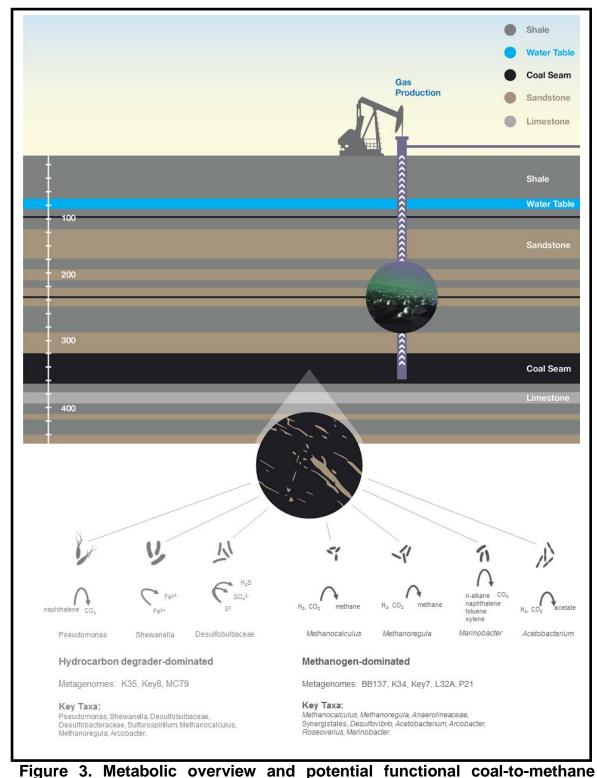
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