

1 **Identify the core bacterial microbiome of hydrocarbon degradation and a**
2 **shift of dominant methanogenesis pathways in oil and aqueous phases of**
3 **petroleum reservoirs of different temperatures from China**

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17 **Abstract**

18 Microorganisms in petroleum reservoirs play significant roles in hydrocarbon degradation
19 and through the terminal electron-accepting process of methanogenesis, they also contribute
20 to microbial enhanced oil recovery (MEOR) worldwide with great economic and
21 environmental benefits. Here, a molecular investigation, using the 16S rRNA and *mcrA* gene
22 profiles based on MiSeq sequencing and clone library construction methods, was conducted
23 on oil and water (aqueous) phases of samples of high (82-88°C), moderate (45-63°C) and low
24 temperatures (21-32°C) from seven petroleum reservoirs in China. A core bacterial
25 microbiome with a small proportion of shared operational taxonomic unit (OTU), but a high
26 proportion of sequences among all reservoirs was discovered, including aerobic degraders,
27 sulfate/nitrate-reducing bacteria, fermentative bacteria and sulfur-oxidizing bacteria
28 distributed mainly in *Proteobacteria*, *Bacteroidetes*, *Deferribacteres*, *Deinococcus-Thermus*,
29 *Firmicutes*, *Spirochaetae*, and *Thermotogae*. Their prevalence in the previously reported
30 petroleum reservoirs and successive enrichment cultures suggests their common roles and
31 functions involved in aliphatic and aromatic hydrocarbon degradation. The methanogenic
32 process generally shifts from the dominant hydrogenotrophic pathway in aqueous phase to
33 acetoclastic pathway in oil phase in high-temperature reservoirs, but the opposite was true for
34 low-temperature reservoirs. No difference was detected between the two phases in moderate
35 temperature reservoirs. Physicochemical factors, including pH, temperature, phase
36 conditions, and nitrate, Mn^{2+} , and Mg^{2+} concentrations were the main ones correlated to the
37 microbial compositional and functional profiles significantly. LEfSe analysis shows
38 distribution differences of microbial groups towards pH, temperature, and oil/aqueous phases.
39 Tax4Fun functional profiling indicates major functional metabolism differences between the
40 two phases, including amino acids, hydrocarbons in oil phase, and carbohydrates in aqueous
41 phase.

42

43 **Keywords:** Oil reservoir; Core microbiome; Oil and aqueous phases; Temperature;
44 Methanogenesis; Hydrocarbon

45 **1 Introduction**

46 Microbial enhanced oil recovery (MEOR) is one of the most feasible and profitable
47 technologies for extracting residual oil from low-productivity reservoirs by utilizing
48 microorganisms as the major functional players for hydrocarbon transformation via
49 fermentation and methanogenesis in MEOR applications (Magot et al., 2000; Mbadinga et al.,
50 2011; Mesle et al., 2013). Implementation of MEOR could also attenuate side-effects by
51 injection of nitrate into oil wells to stimulate the propagation of nitrate-reducing bacteria
52 (NRB) and inhibit the growth of sulfate-reducing bacteria (SRB), subsequently reducing the
53 negative effects by sulfidogenic activity on corrosion and oil quality deterioration (Gao et al.,
54 2013). Currently, investigations on compositional and functional profiles of petroleum
55 reservoir microbiome have attracted attention from both scientists and engineers because the
56 information will improve the understanding on microbial diversity and function in oil
57 reservoir systems and implement MEOR through manipulating microbial community
58 composition and activities (Magot et al., 2000; Mesle et al., 2013; Youssef et al., 2009).

59

60 Petroleum reservoirs are often characterized as high temperature, and high-pressure
61 subterranean ecosystems, but low-temperature reservoirs also exist around the world
62 (Grabowski et al., 2005; Li et al., 2013; Voordouw et al., 1996; Li et al., 2014). The
63 physicochemical and geographic differences among petroleum reservoirs together with
64 exogenous injection of microorganisms and nutrients during MEOR operation could affect
65 the composition and functional groups of the indigenous microbial community. Many
66 previously discovered bacterial and archaeal groups have important functional roles, e.g.,

67 sulfate-reducing, nitrate/nitrite-reducing and fermentative bacteria, thermophilic archaeal
68 hydrocarbon degraders and methanogens as well as exogenous aerobic hydrocarbon
69 degraders in the microbial community (Gao et al., 2015a; Gao et al., 2015b; Lenchi et al.,
70 2013; Li et al., 2012a; Orphan et al., 2003; Wang et al., 2012; Grabowski et al., 2005; Pham
71 et al., 2009). The common and specific roles and functions of these microorganisms in
72 petroleum reservoirs are of great value to be comprehensively examined.

73

74 Methanogens cooperate with hydrocarbon degraders and fermentative bacteria synergistically
75 to make the overall degradation processes thermodynamically favorable under largely
76 anaerobic conditions (Mbadinga et al., 2011). Fermentative products from hydrocarbon
77 degradation provide the essential growth substrates for methanogens with two major
78 pathways, namely hydrogenotrophic and acetoclastic methanogenesis (Magot et al., 2000). In
79 many cases the syntrophic acetate oxidization pathway coupling with hydrogenotrophic
80 methanogenesis is prevalent in both *in situ* and enrichment incubation *ex situ* (Mesle et al.,
81 2013; Liang et al., 2015; Wang et al., 2012; Lee et al., 2015; Mayumi et al., 2011). However,
82 very little is known on the factors that shape the methanogenic pathways and the structure of
83 methanogenic communities.

84

85 Petroleum reservoir is a complex system consisting of porous sandstones with oil, water and
86 air. Microorganisms attached to the oil phase of petroleum fluids are largely neglected in
87 majority of the previous investigations due to technical difficulties in DNA extraction and
88 sequencing (Kryachko et al. 2012). Oil-attached microorganisms influence the oil-water
89 interface properties by the production of biosurfactants and metabolites or forming adsorbed
90 colloidal particles to enhance oil recovery performance (Kobayashi et al., 2012; Kryachko et
91 al., 2012; Wang et al., 2014). The distribution, function, and contribution of these
92 microorganisms to MEOR success could be very different from those of the aqueous phase

93 (Kobayashi et al. 2012). Oil-degrading microorganisms, including thermophilic hydrocarbon-
94 degrading archaea and methanogens, play different roles in the MEOR process (Mbadinga et
95 al., 2011). Their compositional patterns and functional profiles towards temperature and
96 oil/aqueous phase are of great value to understand the mechanism of MEOR. Knowledge on
97 the major microbial drivers, their potential functions, distribution characteristics and
98 changing patterns towards environmental parameters should be one of the research directions
99 for a better understanding of the MEOR process. In this study, the research objectives were to
100 reveal the compositional and functional differences of petroleum reservoir inhabiting
101 microorganisms under different temperatures (high: 82-88°C, moderate: 45-63°C and low:
102 21-32°C), the methanogenesis pathways in oil and aqueous phases of these samples, and the
103 influence of physicochemical factors on microbial community composition.

104 **2 Materials and Methods**

105 **2.1 Characterization of geographic properties of sampling reservoirs**

106 Petroleum production fluid samples were collected from seven oilfields across China,
107 covering oil wells of different geographical locations and temperatures. The reservoirs and
108 crude oil properties together with the aqueous phase chemical characteristics of this study are
109 listed in [Table 1](#). Detailed reservoir properties are described in the Supplementary Materials.

110 **2.2 Sample collection, ion concentration measuring, and DNA extraction**

111 Each sample containing a mixture of crude oil and water was collected on site into a sterilized
112 container after flushing each wellhead for at least 3-5 min. The containers were screw capped
113 to avoid air invasion and transported back to the nearby laboratory immediately for further
114 processing. The oil and water mixture was gently heated to 50°C to make it semi-fluid and
115 then separated into oil and aqueous phases in a separatory funnel. Heating was operated as

116 short as possible according to the solidification degree of individual petroleum fluid samples,
117 in order to reduce the lytic effect of microbial cells within them. Ion concentrations of the
118 aqueous phase samples were measured using Dionex 600 ion chromatography (Triad
119 Scientific, Inc., Manasquan, NJ, USA) following the manufacturer's instructions.

120

121 To obtain aqueous phase DNA, the aqueous phase of each sample after oil and water
122 separation was first filtered through a 0.22- μ m-pore-size polycarbonate membrane filter and
123 a portion of the membrane filter was used to extract DNA by AxyPrepTM Bacterial Genomic
124 DNA Miniprep Kit according to manufacturer's instruction (Axygen Biosciences, USA). For
125 the oil phase, three volumes of iso-octane (2, 2, 4-trimethylpentane) were used to dissolve the
126 crude oil, and then centrifuged at 5000 rpm for 30 min to concentrate non-dissolved
127 particulates and microbial cells for at least three times to obtain enough materials. Repetitive
128 DNA extractions were then conducted on these materials and combined to meet the quantity
129 requirement for the downstream quality control. Finally, all DNA samples from aqueous and
130 oil phases were measured by Nanodrop for concentration and checked for DNA integrity by
131 electrophoresis.

132 **2.3 Clone library construction and analysis**

133 The methyl-CoA enzyme-encoding gene (*mcrA*) primer pair, ME3MF&ME3MF-
134 e/ME2r'(Narihiro and Sekiguchi, 2011) and the 16S rRNA gene primer pair targeting
135 methanogens, MetA86f/ARC915r (Wright and Pimm, 2003; Narihiro and Sekiguchi, 2011;
136 Yu et al., 2008), were used to generate gene clone libraries for each sample (aqueous and oil
137 phases separately) according to the method established previously (Zhou et al., 2015) ([Table](#)
138 [2](#)).

139

140 FunGene pipeline chimera check was applied to check *mcrA* gene sequences using the
141 UCHIME *de novo* mode (Edgar et al., 2011). USEARCH software was used to check
142 chimeras of methanogenic 16S rRNA gene sequences using QIIME compatible SILVA 119
143 release SSURef database (“rdp_gold fasta”) file as a reference. Then, for *mcrA* gene
144 sequences, *de novo* OTU picking method was applied by QIIME at a cutoff value of 0.05
145 (Caporaso et al., 2010). Representative OTU sequences were aligned and inserted into the
146 *mcrA* gene ARB database by the maximum parsimony method without changing the initial
147 tree topology (Angel et al., 2012; Ludwig et al., 2004). The phylogenetic affiliation was
148 assigned, and taxonomic composition results were processed by QIIME accordingly (Figure
149 S1). For methanogenic 16S rRNA gene sequence clustering and diversity analysis, the same
150 method was applied via QIIME as described in the followings.

151 **2.4 MiSeq sequencing and QIIME based analysis**

152 The prokaryotic universal primer pair 515F/909R (Caporaso et al., 2012; Wang and Qian,
153 2009) and archaeal universal primer pair Arch347F/Arch806R (Takai and Horikoshi, 2000)
154 were used to amplify samples of this study (both with barcodes attached to the forward
155 primers, Table 2). Two independent PCR reactions were conducted and then combined to
156 yield enough PCR products to compromise variations between different batches. Then,
157 pooled PCR products of each sample with approximately 100 ng or 200 ng DNA were
158 subjected to one MiSeq run. Sequencing samples were prepared using TruSeq DNA Kit
159 according to the manufacturer’s instruction. The library was uploaded to an Illumina MiSeq
160 platform for sequencing with reagent kit v2 (2× 250 bp) or v3 (2× 300 bp) as described in the
161 manufacturer’s manual.

162

163 After merging paired-end reads from raw sequencing data by FLASH-1.2.8, fastx-toolkit was
164 applied to split merged reads from one run into individual samples according to the primer

165 barcodes (Magoc and Salzberg, 2011). Then, all sequences were split into each library with
166 the name of each sample attached according to the barcode map using QIIME command
167 “split_libraries” (Caporaso et al., 2010). The criterion for filtering out underqualified
168 sequences was “-s 15 -k -a 6 -r -l 150 -b 12 -M 5 -e 0”. Chimera checking was conducted by
169 USEARCH software using QIIME compatible SILVA 119 release SSURef database
170 (“rdp_gold” fasta) file as the reference (Edgar, 2010). Clustering, picking OTU, taxonomy
171 assignment, aligning, filtering alignments and phylogenetic tree construction, taxonomic
172 composition summarizing, alpha and beta diversity analyses were conducted step-by-step by
173 the QIIME pipeline with QIIME compatible SILVA 119 SSURef database as the reference
174 (Caporaso et al., 2010). In clustering, “pick_open_reference_otus.py” command was used to
175 conduct OTU dividing and BLAST method was used to assign taxonomy to input sequences.
176 This subsampled open reference OTU picking method was the performed-optimized and
177 optimal strategy suggested by the developers (Rideout et al., 2014). After OTU table “biom”
178 files were generated, in order to get the bacterial community composition information from
179 prokaryotic 16S rRNA gene primer amplified libraries, “filter_taxa_from_otu_table.py”
180 command was introduced to only retain bacterial OTUs in the “biom” file. Similarly,
181 exclusive archaeal and methanogenic OTU table files could also be processed from archaeal
182 16S rRNA gene primer amplified libraries. Summary information of each sample OTU
183 abundance could be calculated by “biom summarize-table” command, and then the lowest
184 number among all samples was chosen as the subsampling size to make each library acquire
185 an even size using “multiple_rarefaction_even_depth.py” command. The taxonomic
186 compositional table was drawn according to the subsampled biom file. Since there was no
187 lanemask file available in this SILVA compatible 119 release SSURef database, alignment
188 filtering method was performed independently with entropy threshold as 0.1 and gap filter
189 threshold as 0.9 after obtaining aligned sequences by PyNAST method. Diversity parameters
190 of each library could be generated by alpha diversity calling, with the rarefaction curve,

191 Good's coverage value, Shannon, Chao1, Simpson and PD whole tree indices calculated.
192 Beta diversity, which delineates the dissimilarity relationship among samples, was generated
193 and visualized through unweighted & weighted UniFrac matrix and non-phylogenetic Bray-
194 Curtis matrix method. The pairwise-shared OTU numbers were calculated from “biom” by
195 the command “shared_phylotypes.py”. Core microbiome (shared OTU table in a specific
196 sample category) was identified by “compute_core_microbiome.py” command.

197 **2.5 Diversity and statistical analysis**

198 The statistical significance of community composition of samples among different categories
199 was valued by anosim and adonis method, implemented in “compare_categories” command
200 in QIIME. Mantel Test was used to compare the distance matrix of physicochemical
201 parameters and UniFrac/Bray-Cutis distance matrix from beta diversity analysis by QIIME.
202 For aqueous phase samples, both of the *in situ* physicochemical parameters and ion
203 concentrations were used in the analysis; while only *in situ* physicochemical parameters were
204 included for oil phase samples. The compositional bar chart and bubble chart were modified
205 and illustrated from taxonomic summary results. Tax4Fun was used to predict the functional
206 capabilities based on abundance profiles of microbial 16S rRNA gene datasets (Asshauer et
207 al., 2015). LDA Effect Size (LEfSe) analysis was applied to illustrate the biomarker species
208 with high statistical significance in different sample categories and the functional profiles
209 statistically distributed in different sample categories (Segata et al., 2011).

210 **2.6 Quantitative PCR on *mcrA* gene abundance**

211 The quantitative PCR measurement was conducted using iTaq™ Universal SYBR® Green
212 Supermix Kit (BIO-RAD). The qPCR mixture contained in 15 µl: 7.5 µl of supermix, 16 µg
213 of BSA (Roche), and 1 µM final concentration of primer pair [ME3MF&ME3MF-e (250:1)
214 and ME2r']. Annealing temperature was set to be the same as the clone library PCR setting,

215 and the rest thermocycling settings were according to the manufacturer's instructions.
216 Randomly picked one pMD18-T plasmid with *mcrA* gene inserted was used to make the
217 standard curve. The DNA template concentration was adjusted to 0-40 ng/μl. Results, which
218 deviated significantly from values in the replicate groups, were omitted and undetermined
219 results (under the detection limit) were also deleted. The property of final adjusted standard
220 curve is $r^2 = 0.995$ and $\text{Eff}\% = 83.32$.

221 **2.7 Sequencing result deposition**

222 For clone library sequences, KT314862-KT315353, were assigned to methanogenic 16S
223 rRNA gene sequences; KT314340-KT314835 were assigned to *mcrA* gene sequences. For
224 high throughput sequencing data based on Illumina MiSeq platform, PRJEB9855 was
225 assigned to prokaryotic 16S rRNA gene sequencing data; PRJEB10996 was assigned to
226 archaeal 16S rRNA gene sequencing data.

227 **3 Results and Discussions**

228 **3.1 Common OTU among different categories and core bacterial microbiome**

229 Community composition results showed that 21 bacterial phyla were obtained with an
230 average abundance of more than 0.1% (Figure 1a), and three major archaeal phyla were
231 *Thaumarchaeota*, *Euryarchaeota*, and *Crenarchaeota* (Figure 1b). Pairwise-shared OTU
232 numbers of all samples indicated that, irrespective of combinations between aqueous and oil
233 phase samples, the average numbers of shared bacterial OTUs ranged from 199.9 to 292.4,
234 accounting for 26.6-36.2% of the total OTU numbers within individual samples; average
235 numbers of shared archaeal OTUs ranged from 1.8 to 11.9, accounting for 8.5-23.4% of total
236 OTU numbers within individual samples (Table S4). Core bacterial OTU numbers among
237 aqueous, oil and all samples were 73, 57 and 46, which accounted for 7.1-10.1%, 5.9-10.3%

238 and 4.5-8.3% of OTU numbers in individual samples. Core archaeal OTU number among
239 aqueous samples was only 3, accounting for 3.9-8.1% of OTU numbers in individual
240 samples, and no archaeal OTU was shared among oil phase samples.

241

242 However, by investigating taxonomic profiles of core bacterial OTUs, the shared OTUs were
243 49, 41 and 34 genera in aqueous, oil and all samples, corresponding to 65.5%, 59.9% and
244 58.8% of average sequences in the total bacterial community, respectively (Tables S5 and
245 S6). Most of the core bacterial OTUs belonged to the most abundant 36 genera, of which the
246 numbers of shared genera among aqueous, oil and all samples were 28, 23 and 23,
247 respectively (Figure 2, and Tables S5 and S6).

248

249 There was no significant difference of shared bacterial OTU numbers within and between
250 aqueous and oil phase samples, suggesting a core microbiome was shared among all
251 components. The core OTUs covered around 2/3 of the total bacterial sequences, even though
252 the percentages of core/total OTU number for individual samples were 4.5% to 10.3%. The
253 core microbiome shared among all petroleum reservoirs could be the key participants
254 mediating critical microbial processes, such as activation, degradation, fermentation, oil-
255 emulsification and methane generation (Yamane et al., 2008; Wang et al., 2014; Pham et al.,
256 2009; Orphan et al., 2000; Magot et al., 2000). This spectrum of core microbiome shares
257 common functional roles in facilitating MEOR and is modified by the *in situ*
258 physicochemical conditions of different reservoirs (Figure 2 and Table S5). It is important to
259 connect the major microbial players, including their community compositions and specific
260 functional capacities, to the interpretation of MEOR processes in the petroleum reservoirs.
261 Meanwhile, the core microbiome serves as a good basis for simplifying microbial participants
262 and their roles in the petroleum reservoirs primarily, and is useful for modeling and
263 monitoring the MEOR processes for petroleum reservoirs from different locations. Moreover,

264 substantial portions of the aerobic bacteria being discovered in the core bacterial microbiome
265 across different reservoirs imply that exogenous bacteria introduced into subsurface
266 reservoirs by water flushing can be also represented in the core composition and play
267 important roles in enhancing oil recovery (Gao et al., 2015b; Youssef et al., 2009).

268

269 Bacterial community distribution in aqueous phase showed correlation with temperature by
270 Mantel Test using unweighted UniFrac matrix method (Table S8). Additionally, temperature
271 also significantly affected the dissimilar distance matrix of PCoA coordinates of the bacterial
272 community by both adonis and anosim statistical analyses (Tables S3 and S7). However,
273 temperature changed the proportions of the taxa of the core microbiome significantly, but not
274 the presence/absence (Tables S3, S7 and S8). Consequently, the common genera discovered
275 from this study show that despite geographic and physicochemical differences, there is a core
276 microbiome with small OTU numbers but large sequence proportions in these petroleum
277 reservoirs across China which are possibly responsible for mediating hydrocarbon
278 degradation processes. Prevalent core genera discovered in this study also appear to be the
279 most abundant components as previously reported in petroleum reservoirs within and beyond
280 China (Gao et al., 2015a; Gao et al., 2015b; Li et al., 2012a; Orphan et al., 2003; Orphan et
281 al., 2000; Tang et al., 2012; Yamane et al., 2008; Dahle et al., 2008; Li et al., 2015; Liang et
282 al., 2015; Li et al., 2014) and enrichment cultures from petroleum reservoirs worldwide (Gray
283 et al., 2011; Li et al., 2012b; Liang et al., 2015; Wang et al., 2011).

284 **3.2 Shift of major methanogenesis pathways between oil and aqueous phases**

285 The methanogenic process generally shifts from the dominant hydrogenotrophic pathway in
286 aqueous phase to acetoclastic pathway in oil phase of high-temperature reservoirs. The
287 opposite is true for low-temperature samples, but no difference was detected between the two
288 phases in moderate temperature reservoirs (Figure 3). The different patterns of shifting of the

289 major methanogenesis mechanisms in aqueous and oil phases were evident in different
290 temperature petroleum reservoirs of this study. Quantitative measurements of *mcrA* gene in
291 both aqueous and oil phases of all samples are summarized in Supplementary Materials
292 (Figure S1).

293

294 The underlying methanogenesis mechanism could change substantially in reservoirs with
295 distinctive geochemical properties (Magot et al., 2000). The dominance of methylotrophic
296 methanogenesis is rarely observed in petroleum reservoirs. It is claimed that there is a very
297 low possibility that methyl-compounds could be generated during the degradation of kerogen
298 (Mesle et al., 2013). However, the dominance of obligate methylotrophic *Methanlobus* was
299 found in all water, rock and coal samples in a coalbed methane reservoir within Eastern
300 Ordos Basin of China (Guo et al., 2012). It could be deduced that the prevalence of
301 methylotrophic methanogenesis under certain conditions is directly fueled by the existence
302 and availability of methyl-containing substances. Since the relative abundances of bacteria in
303 petroleum reservoir samples are always higher than those of methanogenic community, the
304 methanogenesis process might not be the dominant process among all microbial processes.
305 For instance, in P1 and P5 aqueous samples, sulfate concentrations are considerably high
306 (Table 1), geochemical conditions are more favorable for sulfate reduction than
307 methanogenesis. A large proportion of Firmicutes and Deferribacteres in P5A were
308 potentially responsible for the activities of sulfate reduction according to their relative
309 abundance (Figure 1) and functional capacities (Figure 2). It is suggested the methylated
310 compounds could be produced by the degradation of glycine betaine (an osmoprotective
311 agent), that are accumulated or generated by halophiles in saline petroleum (Ollivier and
312 Alazard, 2010). Subsequently, it fuels the growth of methylotrophic methanogens in certain
313 petroleum environments. Although no molecular or chemical evidence was reported for this
314 process in petroleum environments, a pioneer study on hydraulic fracturing in shales has

315 detected glycine betaine as the major known osmoprotectant in the produced fluids, and the
316 combined metagenomic and metabolite findings suggest the similar glycine betaine
317 utilization pathways in fractured shales (Daly et al., 2016).

318

319 Numerous studies based on *in situ* or in enrichment incubation microcosms on aqueous phase
320 of reservoir fluids indicate that syntrophic acetate oxidation associated with hydrogenotrophic
321 methanogenesis is the major hydrocarbon degradation pathway (Wang et al., 2012; Mayumi
322 et al., 2011; Lee et al., 2015; Mbadanga et al., 2012; Gray et al., 2011). The dominance of
323 hydrogenotrophic methanogenesis in subsurface ecosystem could result from the external
324 hydrogen originated from maturation of organic matters and/or mineral hydrolysis (Head et
325 al., 2003), and the synergistic effect in association with acetate oxidizers, whereby acetate
326 was firstly oxidized to H₂ and CO₂, then utilized by methanogenesis (Liu and Whitman,
327 2008). Additionally, a stable isotope labeling experiment on oil-degrading microcosm
328 showed that despite coexistence of acetoclastic methanogenesis and acetate syntrophic
329 oxidization in the initial state, the latter process prevailed over the former one when
330 introducing a low acetate concentration initially (Gray et al., 2011). The above evidence
331 suggests that acetate syntrophic oxidization could exceed acetoclastic methanogenesis and
332 contribute substrates H₂ and CO₂ to potentially favor hydrogenotrophic methanogenesis
333 process.

334

335 It is still difficult to determine whether the temperature has directly or indirectly been
336 involved in the alteration of the methanogenesis pathways. Reservoir fluid constituents may
337 affect methanogenic degradation because crude oil and creosote inhibit acetoclastic
338 methanogenesis (Warren et al., 2004) and volatile hydrocarbons (*n*C₅-*n*C₁₀) inhibit
339 methanogenic degradation rate without changing the abundances of both hydrogenotrophic
340 and acetoclastic methanogens (Sherry et al., 2014). Since the most community data available

341 currently are based on the microbial assemblages within injection or production water, new
342 understanding on the local microbiome distribution and changes shall focus on oil or
343 hydrophobic fraction of the reservoir fluids (Kobayashi et al., 2012; Tang et al., 2012; Lenchi
344 et al., 2013). Consequently, the alteration pattern of major methanogenesis in the aqueous and
345 oil phases under different temperature conditions could be further delineated. A combination
346 of methods, including synthesis and quantification of degradation intermediate (Bian et al.,
347 2015), stable isotope labeling on tracing substrate transformation (Gray et al., 2011) and
348 molecular analysis of the metabolically active microorganisms can advance the information
349 on anaerobic degradation and methanogenesis processes in reservoir systems.

350 **3.3 Physicochemical influence and Taxa & Function profiles**

351 Temperature is an important physical factor shaping the community structure of bacterial
352 (anosim & adonis $p < 0.01$) and methanogenic communities (anosim $p < 0.05$ & adonis $p <$
353 0.01) of the samples in this study (Figure 4 and Table S7). Furthermore, a significant
354 difference of taxa abundance among the three temperature categories for both bacterial and
355 archaeal communities was evident by LEfSe analysis (Figure 5). For bacterial community,
356 sample group (aqueous or oil phase from the same sample group), temperature (anosim &
357 adonis $p < 0.01$), and pH (adonis $p < 0.05$) showed significant effects on separating samples
358 into different categories. For archaeal community, significant differences among sample
359 categories were detected with sample group and temperature (both adonis $p < 0.05$), pH and
360 phase (anosim & adonis $p < 0.05$). For methanogenic community, significant differences
361 among sample categories were detected with sample group and temperature (sample group:
362 anosim & adonis $p < 0.05$; temperature: anosim $p < 0.05$ & adonis $p < 0.01$) (Table S7).

363

364 Nitrate concentration dissimilarity matrix was significantly associated with all
365 unweighted/weighted UniFrac and Bray-Cutis matrices (all $p < 0.05$) for aqueous bacterial

366 community (Table S8). Meanwhile, methanogenic community in oil phase was significantly
367 affected by reservoir depth, temperature, pH and water flooding operation years based on
368 association analysis of weighted UniFrac matrix but not unweighted UniFrac matrix (all $p <$
369 0.05) (Table S8), indicating that it was the abundance difference of certain taxa affecting the
370 compositional pattern. More detailed relationships between physicochemical factors and
371 bacterial/archaeal/methanogenic communities are summarized (Tables S7 and S8).

372

373 Nitrate dissimilarity matrix was significantly correlated with dissimilarity matrices of all
374 aqueous phase bacterial communities using both unweight/weight UniFrac and Bray-Curtis
375 matrices (all $p < 0.05$) (Table S8). Nitrate is an important chemical used in injection water to
376 inhibit corrosion and maintain crude oil quality (Gao et al., 2013). Nitrate stimulates the
377 growth of nitrate-reducing bacteria and inhibits the growth of SRB (Nemati et al., 2001; Gao
378 et al., 2013). Consequently, nitrate injection shapes the microbial communities in petroleum
379 reservoirs. Mn^{2+} and Mg^{2+} were shown to be strongly associated with bacterial community of
380 aqueous phase based on Bray-Curtis matrix (both $p < 0.05$) (Table S8). Metal ions can be
381 electron acceptor for direct or indirect hydrocarbon degradation under anaerobic conditions
382 (Mbadinga et al., 2011). Metal reducers could also utilize electrons from syntrophic partners
383 to further facilitate direct aromatic hydrocarbon degradation (Kunapuli et al., 2007).

384

385 Differences in taxa and function profiles in oil and aqueous phases were analyzed based on
386 LEfSe and Tax4Fun (Figures S2 and S3, and Table S9). For bacterial communities, amino
387 acid metabolism and xenobiotics biodegradation and metabolism were distributed more in oil
388 phase, while carbohydrate metabolism was distributed more in aqueous phase. For archaeal
389 communities, ubiquinone and other terpenoid quinone biosynthesis and butanoate and
390 tryptophan metabolism together with other glycan degradation pathways were distributed
391 more in oil phase. Since FTU values of archaeal communities (fraction of taxa that could be

392 mapped to existing KEGG pathway) were unevenly distributed from sample to sample, the
393 reliability of these functional predictions is in question (Table S9). Meanwhile, the database
394 of Tax4Fun is far from completeness, due to the enormous amount of uncultured
395 microorganism and their unknown genomes. To date, there are still very limited studies on
396 investigating inhabiting preference of microbiome in petroleum reservoirs. Meanwhile, their
397 functional contributions to hydrocarbon degradation and methanogenesis in both aqueous and
398 oil phases remain elusive, which calls for further efforts on this point (Kryachko et al., 2012;
399 Wang et al., 2014; Kobayashi et al., 2012).

400 **4 Conclusions**

401 A core bacterial microbiome containing a small proportion of OTUs but a relatively large
402 proportion of sequences mediating hydrocarbon degradation and fermentation was revealed
403 by analysis of oil reservoirs of different temperatures. The core and common bacterial
404 microbiome of major biodegrading functions were shared among geographically and
405 physicochemically different reservoirs. The different and shifting patterns of the dominant
406 methanogenesis pathway in aqueous and oil phase within samples of different temperature
407 were evident. Factors of pH, temperature, phase conditions and nitrate, Mn^{2+} , and Mg^{2+}
408 concentrations shaped the microbial compositional and functional profiles significantly.
409 Moreover, biomarker groups of bacteria and archaea associated with different pH,
410 temperature and phase conditions indicate major differences in the biochemical function of
411 amino acid metabolism, xenobiotics metabolism enriched in the oil phase, and carbohydrate
412 metabolism enriched in the aqueous phase.

413

414 *Author contributions.* ZZ, BL, L-YW, B-ZM, HS, and J-DG conceived the project and
415 designed experiments. ZZ, BL, and L-YW conducted the sampling, chemical and molecular
416 experiments. J-FL, B-ZM, and J-DG managed sample collection, supervised on data

417 interpretation. ZZ performed the original data analysis and drafted the original manuscript.
418 All members contributed to refining the manuscript and approved the final version.

419

420 *Competing interests.* The authors declare no conflict of interest.

421

422 *Acknowledgments.* Dr. Xiangzhen Li's group at Chengdu Institute of Biology, Chinese
423 Academy of Sciences was thanked for MiSeq sequencing efforts and related technical
424 support. Ms. Kelly Lau was thanked for her supportive technician work. We thank the
425 support from local administrative and technical staff in Shengli, Daqing, Huabei, Xinjiang
426 Karamay, and Jiangsu Oilfields. This project was supported by NSFC/RGC (grant no. N
427 HKU718/11, J-DG) and NSFC (grant no. 41373070 and 41530318, B-ZM) and a
428 postgraduate scholarship from HKU (ZZ).

429

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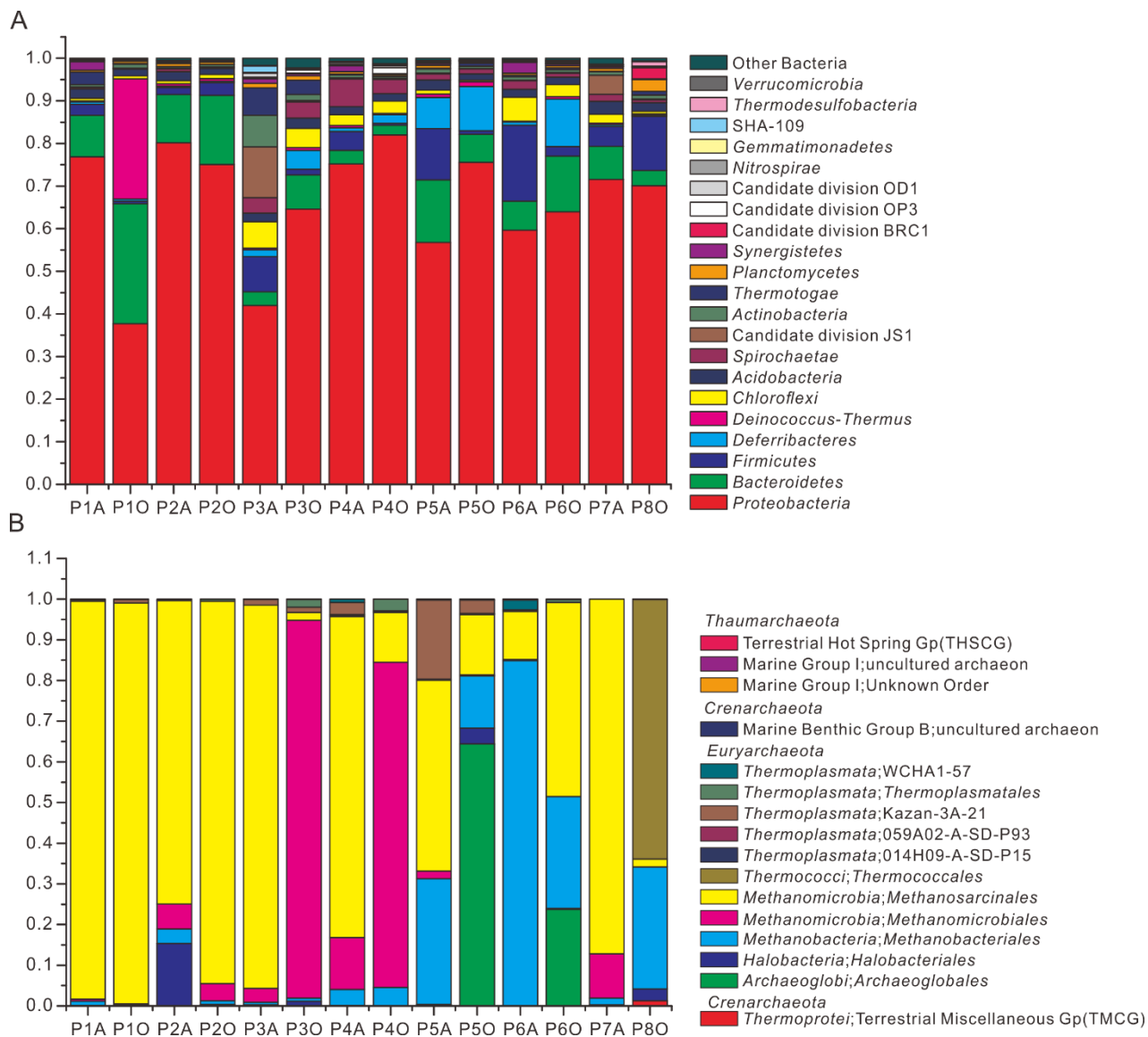


Figure 1. Relative abundance of bacteria (A) and archaea (B) from 14 aqueous and oil phase samples. Bacterial community was taxonomically assigned at the level of phylum. Those phyla with an average abundance of all samples below 0.1% were combined into “Other Bacteria” category. The archaeal community was taxonomically assigned at the level of class.

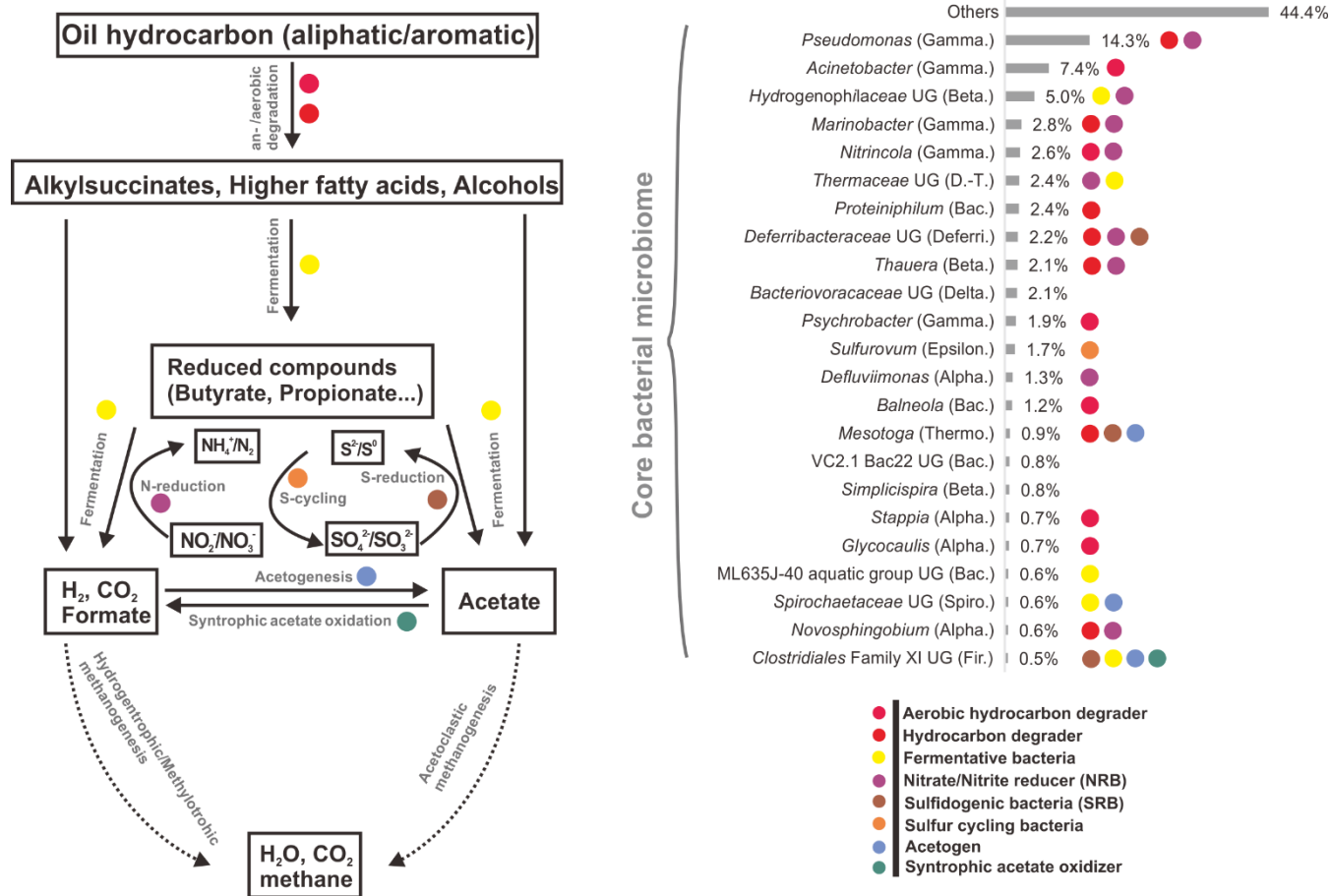


Figure 2. Average abundances and functional roles of core bacterial microbiome in all petroleum samples (including aqueous and oil phase samples). Core microbial OTUs (23 OTUs) were affiliated to the most abundant genera. The functional roles were assigned according to reference publications (Supplementary Material). Degradation pathway was modified from Mbadinga et al., 2011. Abbreviations: Alpha., Beta., Gamma., Delta., Epsilon (Alpha-, Beta-, Gamma-, Delta-, Epsilonproteobacteria), D.-T. (*Deinococcus-Thermus*), Bac. (*Bacteroidetes*), Deferri. (*Deferribacteres*), Delta. (*Deltaproteobacteria*), Thermo. (*Thermotogae*), Spiro. (*Spirochaetae*) and Fir. (*Firmicutes*). UG stands for uncultured genus.

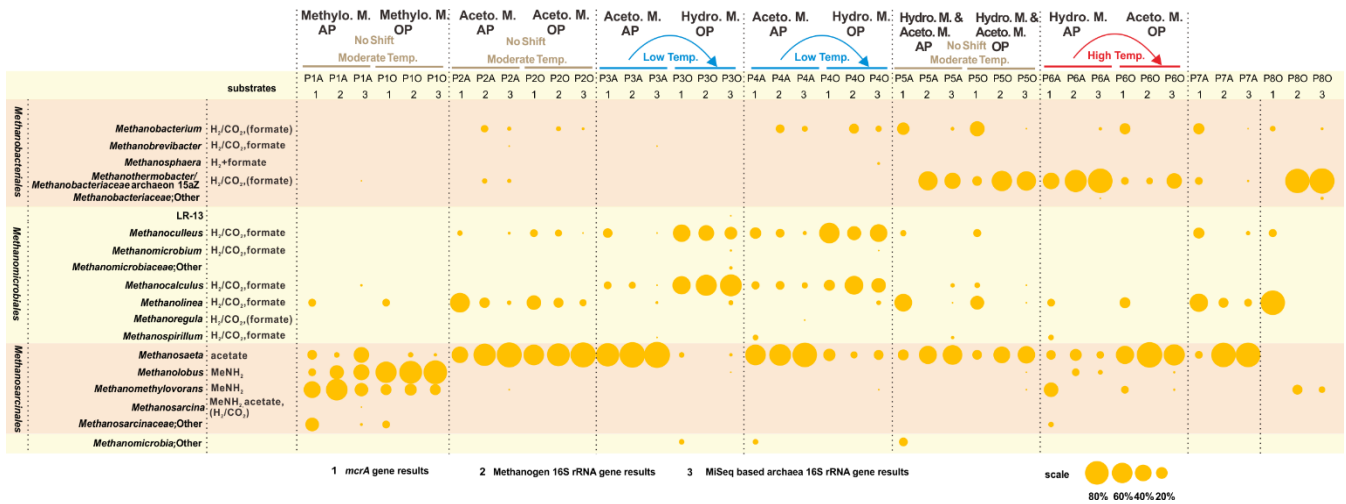


Figure 3. Bubble chart of the proportional composition of methanogens based on MiSeq sequencing results of archaeal 16S rRNA genes and clone sequencing results of *mcrA* gene and methanogenic 16S rRNA gene. The major substrate utilization properties were originated from Liu et al., 2008. $MeNH_2$ is methylamine and substrates in parentheses refer to being utilized by some but not all species. *Methanothermobacter thermautotrophicus* strain Delta H is the phylogenetically closest cultured clone to *Methanobacteriaceae* archaeon 15aZ. Thus, the combination of *Methanothermobacter* and *Methanobacteriaceae* archaeon 15aZ was used. Methanogenesis shift was based on the transition of major methanogenesis pathway. Abbreviations: Aceto. M., Hydro. M. and Methylo. M. (Acetoclastic methanogenesis, Hydrogenotrophic methanogenesis, and Methylo-trophic methanogenesis).

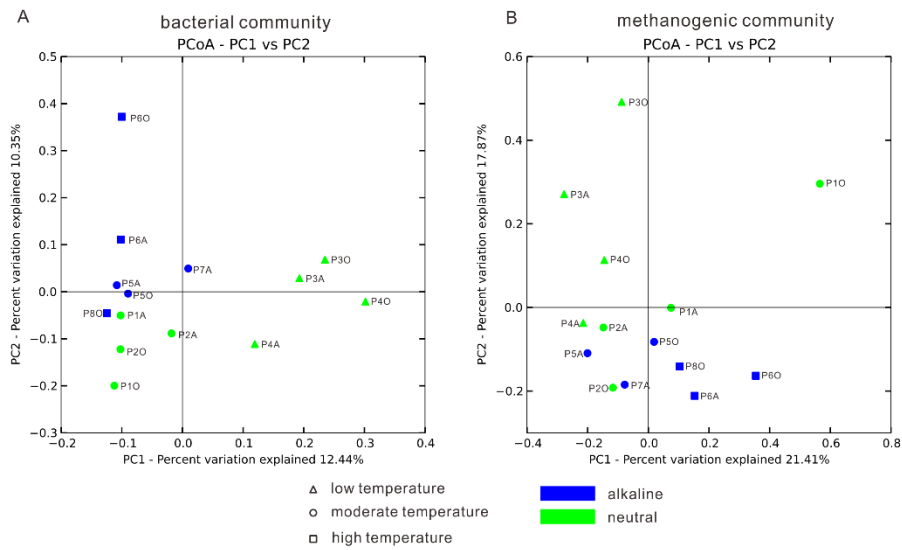


Figure 4. Principal Coordinate Analysis plot figures based on unweighted UniFrac matrices. Bacterial (A) and methanogenic (B) communities of 14 samples were separately analyzed to delineate the dissimilarity distances between each sample based on phylogenetic classification. The sample dots were categorized in terms of temperature (dot shape) and pH condition (color).

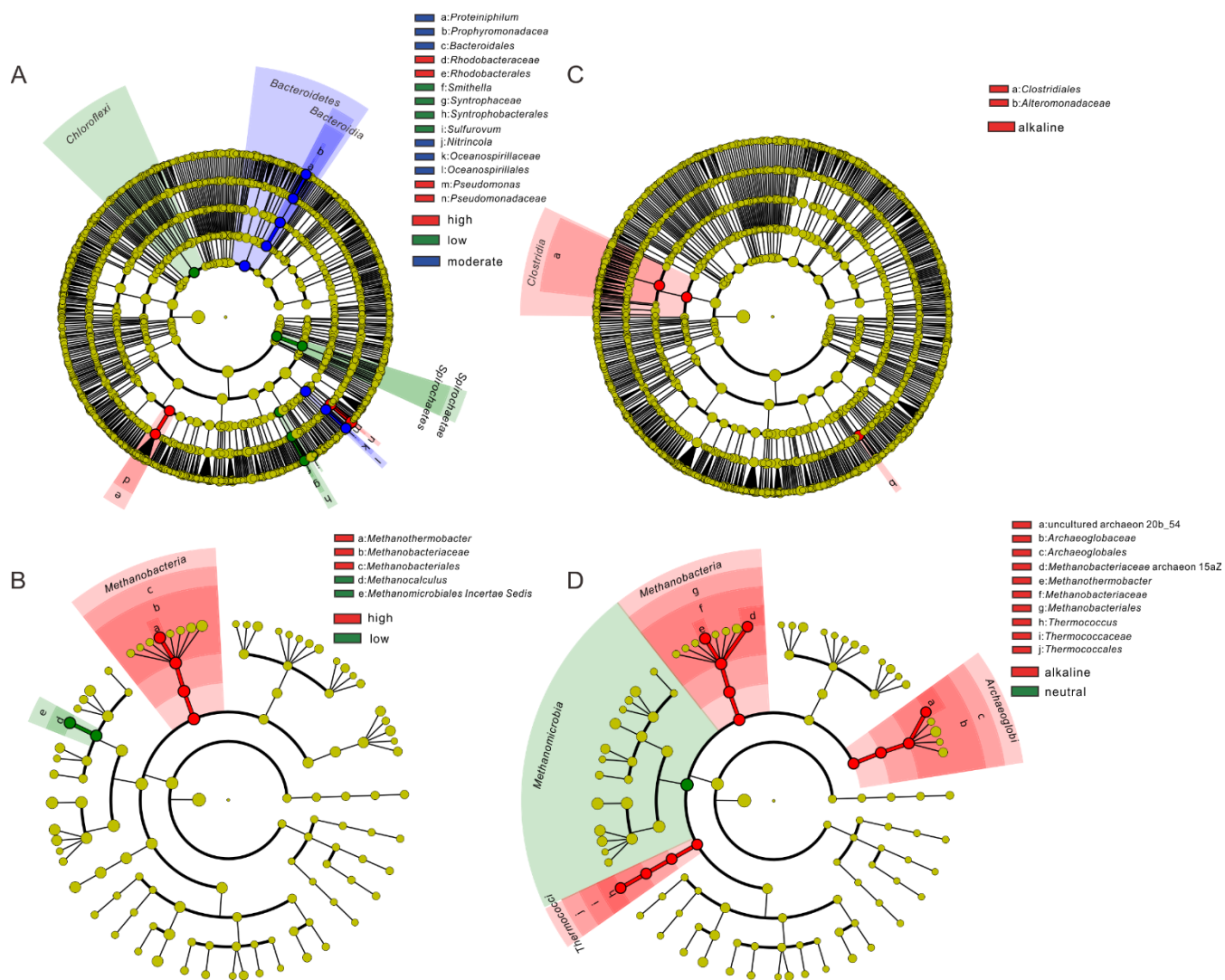


Figure 5. Cladogram based on LEfSe analysis results on bacteria (A, C) and archaea (B, D) in terms of temperature (A, B) and pH (C, D) categories. The taxonomic trees were generated from phylum to genus (inside to outside) in the hierarchical structure. Biomarker taxonomic levels were labeled in color which had at least logarithmic LDA score of 3.5. Pre-sample normalization was used to format the relative abundance data. All-against-all strategy was used in the multi-class analysis step.

Table 1. Reservoir geophysical properties and physicochemical properties of aqueous phase samples

Location name	Shengli-Zhan3-26	Huabei-B51	Xinjiang (Karamay)-Liuzhong	Xinjiang (Karamay)-Qixi	Jiangsu-Wei5	Jiangsu-Gao6	Daqing (Xingbei)-Xing4	Jiangsu-Qinying
Sample name	P1	P2	P3	P4	P5	P6	P7	P8
Depth (m)	~1300	~1101	~480	~1088	1018	1970	800-1200	2280-2524
Temp (°C)	63	45	21	32	53.5	82	49	88
pH	7.1	7.2	7.0	7.1	8.62	8.88	8.5	8.2
Effective porosity (%)	30	22.2	20.5	17.4	21.7	17.4	28.3	16.5
Average permeability ($\times 10^{-3} \mu\text{m}^2$)	800	12.6	466	274	91.8	94.5	481	22
Oil viscosity (mPa.s)	1720	402	417	44.8	292	8.4-83.9	13.86	98-188
Water flooding operation (years)	22	4	38	46	13	18	30	0
Cl ⁻ (mg l ⁻¹)	3850	819	3864	2000	7563.01	1155.95	659.31	-
SO ₄ ²⁻ (mg l ⁻¹)	2244	32.4	124.8	7.7	1921.06	156.81	333.29	-
PO ₄ ³⁻ (mg l ⁻¹)	0.1	Nd	Nd	Nd	139.77	6.29	Nd	-
NO ₃ ⁻ (mg l ⁻¹)	Nd	Nd	34.1	1.4	Nd	0.44	0.45	-
Na ⁺ (mg l ⁻¹)	3313	1064	4196	5399	7363.52	1593.55	1363.43	-
K ⁺ (mg l ⁻¹)	94.2	22.3	35.1	45.6	9.36	3.26	3.55	-
Ca ²⁺ (mg l ⁻¹)	195.6	53.0	103.3	128.2	17.42	7.24	41.94	-
Mg ²⁺ (mg l ⁻¹)	46.1	17.6	44.7	64.0	10.27	2.25	11.02	-
Mn ²⁺ (mg l ⁻¹)	0.3	0.1	0.3	0.4	Nd	Nd	Nd	-
Acetate (mg l ⁻¹)	32	57.9	344	6.97	44.07	285.66	Nd	-

“Nd” means not detected. No aqueous phase was obtained from Jiangsu-Qinying reservoir sample.

Table 2. PCR primers and PCR settings

	Primers	Sequences	Primer final concentration	Template DNA quantity	References
<i>mcrA</i> gene primer	ME3MF&ME3MF-e	ME3MF: ATGTCNNGGTGGHGTMGSTTYAC, ME3MF-e: ATGAGCGGTGGTGTCCGGTTTCAC; concentration ratio = 250:1	1 μ M	10-20 ng	(Narihiro and Sekiguchi, 2011)
	ME2r'	TCATBGCRTAGTTDGGRTAGT	1 μ M	10-20 ng	
PCR setting	95°C 2 min; 35 \times (95°C 30 s, 59.5°C 30 s and 72°C 40 s); 72°C 5 min and 4°C 2 min.				
methanogenic 16S rRNA gene primer	MetA86f	GCTCAGTAACACGTGG	300 nM	10-20 ng	(Wright and Pimm, 2003; Narihiro and Sekiguchi, 2011; Yu et al., 2008)
	ARC915r	GTGCTCCCCCGCCAATTCCT	300 nM	10-20 ng	
PCR setting	95°C 5 min; 35 \times (95°C for 30 s, 57°C for 30 s, and 72°C for 1 min); 72°C 10 min, 4°C 2 min.				
Prokaryotic universal primer	515F	GTGCCAGCMGCCGCGGTAA (12 nt barcode added)	1 μ M	10 ng	(Caporaso et al., 2012; Wang and Qian, 2009)
	909R	GGACTACHVGGGTWTCTAAT	1 μ M	10 ng	
PCR setting	94°C 3 min; 30 \times (94°C 40 s, 56°C 60 s and 72°C 60 s); 72°C for 10 min, and 4°C for 2 min				
Archaeal universal primer	Arch347F	GYGCASCAGKCGMGA AW (12 nt barcode added)	1 μ M	10 ng	(Takai and Horikoshi, 2000)
	Arch806R	GGACTACVSGGGTATCTAAT	1 μ M	10 ng	
PCR setting	95°C for 5 min; 32 \times (95°C 30 s, 50°C 30 s and 72°C 40 s); 72°C 10 min, and 4°C 2 min				