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1 **How to use molecular biology tools for the study of the anaerobic digestion process?**

2

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23

24

25 **Abstract**

26 Anaerobic Digestion is used with success for the treatment of solid waste, urban and industrial
27 effluents with a concomitant energy production. The process is robust and stable, but the
28 complexity of the microbial community involved in the process is not yet fully comprehensive.
29 Nowadays, the study of this complex ecosystem is facilitated by the availability of different
30 molecular tools, but it is very important to choose the adequate tool to answer specific
31 questions. The aim of this review is to describe different molecular techniques, indicate the

32 questions that can be addressed by each technique, enumerate their limitations and give practical
33 advices for their use. Examples of how the molecular tools have been used to address various
34 questions in anaerobic digestion are presented. The key point now is to apply all this
35 information to improve anaerobic digestion. The integration of concepts of microbial-ecology,
36 environmental-engineering, modeling and bioinformatics is currently necessary.

37

38 **Key words**

39 Anaerobic digestion, bioreactor, molecular biology, microbial ecology, function, microbial
40 resource management.

41

42 **List of abbreviations**

43	AD	anaerobic digestion
44	ANOSIM	analysis of similarity
45	ANOVA	analysis of variance
46	ARDRA	amplified ribosomal DNA restriction analysis
47	BES	2-bromoethanesulphonic acid
48	BONCAT	biorthogonal non-canonical amino acid tagging
49	CE-SSCP	capillary electrophoresis Single Strand Conformation Polymorphism
50	CCA	canonical correspondence analysis
51	CLSM	confocal laser scanning microscopy
52	CNV	copy number variation
53	DIET	direct interspecies electron transfer
54	DGGE	denaturing gradient gel electrophoresis
55	DNA	desoxyribo nucleic acid
56	EGSB	Expanded granular sludge blanket
57	FISH	fluorescent in situ hybridization
58	HIT	Hydrogen interspecies transfer
59	MAR-FISH	micro auto radiographic fluorescent in situ hybridization
60	NanoSIMS	nanoscale secondary ion mass spectrometry

61	NGS	next generation sequencing
62	NMDS	nonmetric multidimensional scaling
63	NPMANOVA	non parametric multivariate ANOVA
64	OLR	organic loading rate
65	OTU	operational taxonomic unit
66	PCA	principal component analysis
67	PCoA	principal coordinate analysis
68	PCR	polymerase chain reaction
69	qPCR	quantitative polymerase chain reaction
70	RISA	ribosomal intergenic spacer analysis
71	RNA	ribo nucleic acid
72	rRNA	ribosomal RNA
73	SAO	syntrophic acetate oxidizers
74	SIMSISH	secondary ion mass spectrometry combined with FISH
75	SIP	stable isotope probing
76	SSCP	Single Strand Conformation Polymorphism
77	ssDNA	single strand DNA
78	T-RFLP	Terminal-Restriction Fragment Length Polymorphism
79	T-RF	Terminal-Restriction Fragment
80	UASB	Upflow anaerobic sludge blanket
81	VSS	volatile suspended solids

82

83

84 **1-Introduction**

85

86 **1.1 Anaerobic Digestion, a robust and complex process**

87 Anaerobic Digestion (AD) of waste and effluents is a robust process that is nowadays used with
88 success in full scale systems for the treatment of solid waste and urban and industrial
89 wastewaters. Worldwide, there are thousands of high rate sludge bed reactors for industrial

90 wastewater treatment, and millions of domestic biogas plants generating energy from waste
91 (Ren 2013; Noyola et al. 2012). Many engineering enterprises are involved in the design and
92 installation of these type of biological reactors. Moreover, new technologies are being tested at
93 lab-scale using different reactor configurations (e.g. UASB, EGSB, anaerobic membrane
94 reactors), integrating different treatments (N-removal, S-removal, micropollutants removal,
95 etc.), or systems to recover nutrients (Batstone and Virdis 2014).

96 Anaerobic biological process is driven by a complex network of microorganisms belonging to
97 *Bacteria* and *Archaea* domains, working together for a complete degradation of the organic
98 compounds into CH₄ and CO₂. In the complete anaerobic chain, four main metabolic steps are
99 involved: 1-hydrolysis, 2-fermentation, 3- acetogenesis, 4- methanogenesis (Zinder 1984)
100 (Figure 1). These four steps involve different microbial guilds who are specialized in different
101 processes. A diverse number of hydrolytic and fermentative bacteria take part in the first two
102 steps, whereas the oxidation of intermediate fermentation products to acetate is performed by
103 either hydrogen- or formate- producing acetogens (Stams and Plugge 2009). Ultimately,
104 methane is exclusively generated from acetate and hydrogen/CO₂ by methanogenic archaea.

105

106 In general, the microbial composition of anaerobic digestion sludge is highly diverse and shows
107 high redundancy. This means that several microorganisms are metabolically flexible and
108 capable of performing the same work. It has been postulated that this characteristic is one of the
109 reasons for the robustness of anaerobic digestion processes (Fernández et al. 1999; Zumstein et
110 al. 2000).

111 To study these very complex ecosystems there are nowadays different molecular tools available.
112 It is very important to choose the adequate tool to answer the questions formulated in each
113 experiment.

114

115 **1.2 Questions to answer**

116 Regarding the microbial composition and function of the communities in methanogenic
117 bioreactors, four main questions are usually formulated:

118 1-Who is there?, 2- How does the community change over time?, 3-How many microorganisms
119 from the different groups are present?, 4-What are the specific functions of the microorganisms
120 within the community and their relation to each other?

121 Molecular ecology techniques have been evolving over the past years to give answers to these
122 questions and an array of options are available and will be discussed in this review (Figure 2).

123

124 ***1.2.1 Microbial diversity - Who is there?***

125 Information on the diversity and identity of the microorganisms involved in the anaerobic
126 digestion process is important to understand bioreactor functioning, especially when concerning
127 new metabolic processes. The discovery of microorganisms involved in the anaerobic oxidation
128 of ammonium (Anammox process) is a nice example (Jetten et al. 1999; Ni and Zhang 2013).
129 Other important key process are syntrophic oxidation of organic acids (McInerney et al. 2008),
130 degradation of recalcitrant compounds such as detergents (Okada et al. 2014), and anaerobic
131 oxidation of methane (Fernández et al. 2008).

132 To assign the identity of the microorganisms involved in a microbial community the most
133 frequently used technique is based on the analysis of the 16S rRNA gene. This gene was
134 proposed as a housekeeping genetic marker to study bacterial and archaeal phylogeny and
135 taxonomy for several reasons: (i) is present in almost all bacteria and archaea; (ii) the function
136 over time has not changed, suggesting that random sequence changes are a more accurate
137 measure of time (evolution); and (iii) the 16S rRNA gene (1,500 bp) is large enough for
138 informatics purposes (Patel et al. 2001). Nowadays, it is possible to determine the genus and
139 species of a bacteria or archaea by sequencing the 16S rRNA gene and compare the sequence
140 with available databases. This technique was adapted to study the microbial composition of a
141 sample as will be explained below.

142

143 ***1.2.2 Microbial dynamics - How does the community change over time?***

144 Different reactor operation parameters, such as applied organic loading rate, hydraulic retention
145 time or operating temperature, are frequently tested to determine the optimal operation
146 conditions for the anaerobic digestion process. In order to explain how these operation
147 conditions can change reactor performances it is necessary to monitor the microbial community
148 composition during operation. When the objective of the research is to test different reactor
149 configurations, different sources of inocula, or different substrates, comparing the microbiology
150 of different systems is needed. In all these cases, numerous community structures could be
151 obtained and the application of a fingerprinting technique such as DGGE, T-RFLP or SSCP is a
152 suitable choice.

153

154 ***1.2.3 Microbial quantification -How many microorganisms from different groups are
155 present?***

156 In AD process it is important to quantify some key groups, in particular the density and
157 proportion of methanogens because of their relevance for ensuring an efficient methanogenic

158 process. For this, techniques that quantify different groups of microorganisms present in a
159 complex community such as Fluorescent *in situ* Hybridization (FISH) or quantitative PCR (q-
160 PCR) are adequate.

161

162 ***1.2.4 Microbial function –What are the roles of different microbial groups in anaerobic*** 163 ***microbial communities?***

164 Discovering specific microbial roles in anaerobic microbial communities is currently one of the
165 most challenging issues for microbiologists and molecular ecologists. Metabolic pathways and
166 interspecies relations involved in the anaerobic process are frequently unknown, in particular
167 when a difficult process is going on, such as the degradation of recalcitrant compounds.
168 Knowing the identity of the microorganisms involved in the process (section 1.2.1) can give a
169 hint on the metabolic potential of abundant players, but it is generally not enough to assign a
170 function to those microorganisms. It is also possible that a single microorganism plays a role at
171 different steps of the metabolic pathways, e.g. *Clostridium* sp. can encode both hydrolytic and
172 fermentative enzymatic machineries. In these cases more sophisticated techniques, such as
173 proteomics, metagenomics, metabolomics and techniques that link the identity with function,
174 such as Stable Isotope Probing (SIP) or micro-autoradiography linked to FISH (MAR-FISH),
175 are the most relevant choice.

176

177 **2-Sampling, storage of samples and environmental data collection**

178 **2.1 Sampling**

179 Sampling is one of the most important steps in microbial ecology analysis. A good sampling
180 strategy is necessary to ensure the success of the study. It has to be taken into account that the
181 use of statistical tools to compare microbial communities might not be correctly applied if a
182 representative sampling is not performed from the beginning of the study.

183 Choice of sample amount is not trivial and has to be considered from the beginning of the study,
184 both to ensure enough biomass concentration for molecular analysis and a representative
185 selection of the microorganisms in the reactor sludge. The majority of the techniques and kits
186 used to extract DNA or RNA from reactor's biomass are designed for soil samples (Griffiths et
187 al. 2000), e.g. PowerSoil DNA isolation (Mo Bio Laboratories Inc. Carlsbad, CA) and Nucleo-
188 Spin Soil (Macherey-Nagel, Düren, Germany). Considering these example kits, manufacturers
189 protocols suggest the use of 0.5 g to 1.0 g of soil. Reactor's sludges are in general suspended in
190 the liquid mixture and present fewer amounts of inorganic compounds and higher amount of
191 cells than soil. According to that, the volume of the sample has to be recalculated. The amount

192 and frequency of sampling might also have an impact in the biological reactor itself; volume of
193 sampling needs to be considered to avoid volume changes in the reactor, especially when
194 working with small lab-scale reactors. To guarantee bioreactor sample representativeness, it is
195 recommended to take samples from different parts of the reactor and pool them, especially in
196 the case of full scale reactors. The sampling of the biomass from solid waste digesters may
197 present additional difficulties due to the heterogeneity of the system. If statistical analysis is
198 going to be applied it is recommended to plan the sampling with a specialist. When the
199 microbial community is analyzed in time series, it is important to adapt the sampling frequency
200 with the objective that samples represent the studied period. Several samples taken during a
201 phase of operation are recommended to be able to compare the different phases. Last but not
202 least, sampling directly in the vial that will be used for further analysis spares some
203 manipulations and limits the risk of sample loss.

204 Once the sample is taken it must be stored correctly to avoid microbial growth during storage.
205 The recommended temperature of storage is -20°C for further DNA extraction and -80°C if it is
206 planned to work with RNA. If a freezer is not available close to the place of sampling, it is
207 recommended to store the samples on ice until you reach the laboratory. If the samples are
208 going to be analyzed by FISH, they should be fixed with formaldehyde or ethanol before storing
209 at -20°C .

210 A fact to consider before DNA/RNA extraction is how to handle the sample. Depending on the
211 kind of reactor, the biomass can be suspended, attached to an inert support, aggregated in
212 granules or forming flocs. In some cases, before nucleic acids extraction, it will be necessary to
213 detach the biomass from the support or disrupt the aggregates. Several protocols are available in
214 literature using ultrasound treatment (Perna et al. 2011) or physical disruption (Bergmann et al.
215 2010).

216

217 **2.2 Environmental data collection**

218 In order to be able to link reactor performances with the microbial community structure,
219 collecting reactor operation data at sampling time is absolutely necessary. In general, parameters
220 such as temperature and pH are recorded daily, but other parameters that need further
221 physicochemical analysis might be determined weekly or monthly depending on the duration of
222 reactor operation. Information about reactor design and wastewater composition is also very
223 important for interpreting the results. There is no general rule but the collection of a complete
224 set of operating data is preferred.

225

226 **3- Tools for studying microbial diversity (Who is there?)**

227

228 **3.1 Cloning and Sanger sequencing**

229 The identity of the microorganisms present in a sample can be determined through PCR
230 amplification and analysis of conserved marker genes. The 16S rRNA gene is the most widely
231 used marker gene and has the most extensive reference database (Su et al. 2012). During the last
232 decades cloning in a plasmid vector followed by Sanger sequencing has been widely used. The
233 technique involves PCR amplification of the 16S rRNA genes using primers directed to either
234 *Bacteria* or *Archaea* domains. In the anaerobic digestion research area, *Bacteria* and *Archaea*
235 are both important, and both corresponding clone libraries can be made separately by choosing
236 domain-specific PCR primer sets during the initial PCR amplification step (Vanwonterghem et
237 al. 2014).

238 Several *Bacteria* and *Archaea* specific primers have been designed (Lane 1991; Grosskopf et al.
239 1998; Leclerc et al. 2004), e.g. the *Bacteria* specific PCR primer sets 27f and 1492r (Lane 1991)
240 and the *Archaea*-specific primers A349f (Takai and Horikoshi 2000) and A915r (Stahl and
241 Amann 1991).

242 The sequences obtained are then clustered into arbitrary Operational Taxonomic Units (OTUs),
243 generally defined as sequences that shared more than 97% identity. This value has been chosen
244 considering that, in general, two strains from the same species share more than 97% homology
245 in their 16S rRNA gene sequences (Yarza et al. 2014). The taxonomic position of the
246 representative sequences is determined by comparing the sequences with databases using
247 BLAST at National Center for Biotechnology Information (NCBI)
248 (<http://www.ncbi.nlm.nih.gov/>) (Altschul et al. 1990), the Classifier at the Ribosomal Database
249 Project (RDP) (<https://rdp.cme.msu.edu/>) (Wang et al. 2007) or the European database
250 (<http://www.ebi.ac.uk/ena>).

251 Several ecological questions of the AD process have been answered by using 16S rRNA gene
252 sequencing techniques. At the beginning, as the cost by sequence was high, the clones were
253 previously analyzed by other techniques (Amplified Ribosomal DNA Restriction Analysis
254 (ARDRA), Single Strand Conformation Polymorphism (SSCP)) and only representative clones
255 were then sequenced. Studies were focused on describing the composition of the complex
256 communities developed in lab-scale methanogenic reactors (Godon et al. 1997) or full scale
257 systems (Roest et al. 2005). Other works focused on the stability of these systems (Fernández et
258 al. 1999, Zumstein et al. 2000) or comparing communities developed under different
259 temperatures (Sekiguchi et al. 1998) or feeding conditions (Goberna et al. 2009).

260 With the development of the automatic Sanger sequencing technology, higher number of clones
261 could be sequenced. In a very extensive sequencing effort, Riviére et al. (2009) analyzed the

262 microbial composition of seven anaerobic sludge digesters. A total of 9,890 16S rRNA gene
263 sequences were analyzed. From the comparison of all the retrieved information, the authors
264 defined three component models within the bacterial communities: 1-a core group of phylotypes
265 common to most of the digesters, 2-phylotypes shared among few digesters and 3-specific
266 phylotypes (Table 1).

267 To give a global view of the microbial diversity involved in the AD process in general, Nelson
268 et al. (2011), compiled the information produced by the cloning/sequencing approach. The
269 authors performed a meta-analysis based on all publicly available 16S rRNA gene sequences
270 generated by Sanger sequencing from various anaerobic digesters up to may 2010. A total of
271 19,388 sequences (16,519 bacterial and 2,869 archaeal) were analyzed, representing 28 known
272 bacterial phyla (e.g. *Proteobacteria*, *Firmicutes*, *Bacteroidetes*, *Chloroflexi*) (Table 1). Archaeal
273 sequences were assigned to 296 OTUs, primarily *Methanosaeta* and the uncharacterized WSA2
274 group.

275 The cloning/sequencing approach is also very valuable to get sequence information for the
276 design of signature oligonucleotides that are complementary to target groups, at the taxonomic
277 level of family, genus, species, or strain. These newly designed oligonucleotides may serve in
278 FISH studies and in the development of real-time PCR assays for quantification (Ariesyady et
279 al. 2007).

280 The methodology might also be applied to study specific functional genes. As an example, one
281 of the key genes involved in the methanogenic pathway was used to study the specific
282 composition of this population (Zhu et al. 2011). A different strategy could be applied to study a
283 specific population using 16S rRNA gene primers specifically designed to target that microbial
284 group (Chouari et al. 2005).

285

286 **3.2 Next generation sequencing of PCR amplicons**

287 The development of Next Generation Sequencing (NGS) technologies has overcome three of the
288 main limitations of the cloning/sequencing technology, as these new methods are: faster,
289 cheaper and high-throughput. Millions of sequences can be obtained in a single run in a
290 complete automatic method within a few hours or days. Sequencing technologies are based on
291 the detection of incorporated nucleotides by different chemistries. In all cases, sequences are
292 generated without the need of a conventional, vector-based cloning procedure (Shokralla et al.
293 2012). A complementary approach is the adoption of nucleotide barcodes in the amplification
294 primers for multiplexing samples. In this way, up to 120 samples from different origins can be
295 mixed in one run and after sequencing their data can be separated according to their barcode
296 (Parameswaran et al. 2007; Meyer et al. 2007). This approach decreases the cost per sample

297 since more samples can be pooled in the sequencing run rather than sequencing fewer samples
298 to greater depth (Cardenas and Tiedje 2008).

299 The first step is the PCR of the desired gene, as the length of the sequences obtained is shorter
300 than in the Sanger methods, specific primers for this technology are used (Cardenas and Tiedje
301 2008, Wang and Qiang 2009). This step is followed by high-throughput sequencing of the
302 resulting amplicons libraries by means of a selected NGS platform. There are four commercially
303 available platforms which use PCR based sequencing systems: Roche 454 Pyrosequencing
304 Genome Sequencer (Roche Diagnostics Corp. Branford, CT, USA), MiSeq and HiSeq 2000
305 (Illumina Inc. San Diego, CA, USA), AB SOLiD System (Life Technologies Corp. Carlsbad,
306 CA, USA) and Ion Personal Genome Machine (Life Technologies, South San Francisco, CA,
307 USA) (Shokralla et al. 2012).

308 These new sequencing technologies are based on different concepts: 454 Pyrosequencing, Ion
309 Personal Genomic Machine (Ion PCM) and Illumina uses real-time sequencing-by-synthesis. In
310 these technologies each nucleotide incorporated by DNA polymerase generates a signal detected
311 by a specific detector system. In the case of 454 Pyrosequencing each nucleotide incorporated
312 release a pyrophosphate molecule which is linked to the production of light by the action of the
313 enzyme luciferase (Margulies et al. 2005). The Ion Personal Genome Machine detects the
314 changes in the hydrogen ion concentration produced when a nucleotide is incorporated into a
315 strand of DNA by the polymerase action (Rothberg et al. 2011). The Illumina platform
316 sequencing is based on the detection of fluorescent signal release after incorporation of a
317 modified nucleotide (Shendure and Ji 2008). These three techniques includes the
318 immobilization of the library fragments on beads (454 and Ion PGM) or a surface flow cell
319 (Illumina) whose surfaces carry oligonucleotides complementary to specific adapter sequences
320 ligated or PCR-generated onto both ends of the fragmented library. After a step of
321 amplification, the plate/chip/flow cell is sequenced en masse by the instrument.

322 Unlike the previous three platforms, SOLiD uses a sequencing by-oligo ligation technology.
323 This process couples oligo adaptor-linked DNA fragments with complementary oligos
324 immobilized on the surface of 1-mm magnetic beads. After the ligation step, a fluorescent
325 readout identifies the ligated 8-mer oligo, which corresponds to one of the four possible bases
326 (Shokralla et al. 2012).

327 Each technology has advantages and disadvantages, the major advantages of the 454
328 Pyrosequencing platform are the long read length obtained and its relatively short run time.
329 Longer sequences generated through 454 provide higher flexibility in terms of accurate
330 annotation of reads in ecological applications involving non model organisms. This has made
331 454 the most commonly used NGS platform for the analysis of environmental DNA for

332 ecological applications. The Ion PGM present a cheap alternative with relative long reads (>200
333 bp) but lower coverage than 454-pyrosequencing. The advantage of both Illumina and SOLiD
334 systems is the high output per run compared to 454 pyrosequencing and Ion PGM. The main
335 drawback of these systems is the relative short-read length. Then, before choosing a platform, it
336 is necessary to evaluate the need of higher coverage or higher sequences length (Shokralla et al.
337 2012).

338 The interpretation of the huge amount of produced data requires appropriate bioinformatic tools
339 and a specific know-how in addition to efficient computational resources. Three bioinformatic
340 pipelines are widely used to analyze high amounts of sequences: QIIME (Caporaso et al. 2010),
341 Mothur (Schloss et al. 2009) and the online tool developed by the Ribosomal Database Project
342 (<http://pyro.cme.msu.edu/>) (Cole et al. 2014).

343 Although this technology was developed ten years ago (Margulies et al. 2005), several
344 applications to the analysis of microbial communities from anaerobic digesters have been
345 reported over the past years. As the sequencing cost has decreased markedly the technologies
346 are more popular and several samples can be analyzed in a short time with reasonable costs.
347 This has allowed to perform studies comparing the microbiology of high number of reactors
348 operated during long time periods. Werner et al. (2011), analyzed the microbiology of 9 full
349 scale methanogenic bioreactors treating brewery wastewater by 454-pyrosequencing. The high
350 amount of datasets was analyzed using specialized bioinformatic tools and correlated with
351 environmental data using statistical analysis. The authors found that each of the nine facilities
352 had a unique community structure with an unprecedented level of stability. Syntrophic
353 populations were highly stable, resilient, and specific for function and environment. This
354 indicates that resilience, rather than dynamic competition, plays an important role in
355 maintaining the necessary syntrophic population. They also found that communities with a
356 greater phylogenetic variability functioned more efficiently.

357 Similarly, Sundberg et al. (2012) studied the bacterial and archaeal compositions of 21 full scale
358 biodigesters operated in thermophilic or mesophilic conditions to produce energy using different
359 solid waste. The reactors were fed with sewage sludge and various combinations of sewage
360 sludge in codigestion with other wastes. The predominant populations in sewage sludge-fed
361 reactors included *Actinobacteria*, *Proteobacteria*, *Chloroflexi*, *Spirochetes*, and *Euryarchaeota*,
362 while *Firmicutes* was the most prevalent in the codigestion reactors fed with other waste (Table
363 1). The main bacterial class found in all reactors was *Clostridia*. Acetoclastic methanogens were
364 detected in sewage sludge, and not in other reactors. Their absence suggests that methane
365 formation from acetate takes place mainly via syntrophic acetate oxidation in these reactors, as
366 previously reported (Karakashev et al. 2006). Statistical analysis revealed that the microbial
367 composition was mainly governed by the type of substrate and the process temperature.

368 Other more specific projects studied the relationship of the microbial communities with
369 different environmental variables as fed sources (Ziganshin et al. 2013) or temperature (Tian et
370 al. 2015).

371 The principal advantage of the sequencing-based methods is that the identity of the
372 microorganisms present in a sample can be determined. The Sanger sequencing method gives
373 more accurate taxonomic assignments as full length sequences can be retrieved (1,500 bp), but,
374 the method is onerous, time-consuming and nowadays widely replaced by NGS methods.
375 However, the recent improvements of the NGS equipment allow achieving longer sequences
376 maintaining a low cost. For instance, in the new 454 GS FLX+System (Roche) reads up to
377 1,000 bp can be obtained. The bottleneck of the new methodology is now to develop
378 bioinformatic tools to analyze the high amount of data generated. The tools available require
379 informatic skills and important computer capacity.

380 Functional insight into the process can be inferred from 16S-based studies by searching for
381 closely related cultured species. However, this should be done with caution as closely related
382 organisms can be functionally different (Jaspers and Overmann 2004). Metabolic functionality
383 can also be inferred by correlating community composition and dynamics with operational
384 conditions and performance parameters using multivariate statistical tools. This correlation is
385 particularly important in anaerobic digestion, where niche differentiation occurs and syntrophic
386 groups are responsible for fulfilling essential steps in the process (Vanwonterghem et al. 2014).

387

388 **3.3 DNA microarrays**

389 DNA microarrays or microchips, are a fast, semi-quantitative technique for phylogenetic
390 identification of bacterial and archaeal species or for detection of functional genes. It is based on
391 the hybridization between extracted DNA sample or 16S rRNA amplicons, which are
392 fluorescently labeled, and complementary oligonucleotide probes that are immobilized on a
393 glass slide (microchip). When hybridization occurs, fluorescence can be detected using a laser.
394 DNA microarrays can detect thousands of genes at the same time (high-throughput) and enables
395 screening of microbial structure and activities (Čater et al. 2013). Moreover, it is fast and once
396 the chip has been design and fabricated the cost of applying the technique is not high. The main
397 disadvantage is that the genes detected are limited to genes previously included in the
398 microchip. Taking into account this limitation, the technique is very useful to compare
399 communities from samples taken under different conditions.

400 To study AD, an ANAEROCHIP was used targetting methanogens which includes 103 probes.
401 This chip has been used to investigate the methanogenic community from a thermophilic

402 continuously stirred tank reactor (Franke-Whittle et al. 2009) and to follow the start up of solid
403 waste digesters (Goberna et al. 2015).

404 Functional genes can be studied using a special chip which particularly includes different genes
405 from the environment (as the GeoChip) (Tu et al. 2014). This strategy was used to study
406 changes in the microbial community during operation of a combined nitrification-anammox
407 reactor established to treat anaerobic digestion supernatant (Zhao et al. 2014). As far as we
408 know, there is no available microarray especially designed with functional genes present in
409 anaerobic digesters communities. This functional microarray may have high potential
410 application to monitor bioreactors operation and start up.

411

412 **4. Microbial dynamics (How does the community change over time?)**

413

414 **4.1 Fingerprinting techniques: DGGE, T-RFLP, SSCP**

415 The most common fingerprinting methods include Denaturing Gradient Gel Electrophoresis
416 (DGGE), Single-Strand Conformation Polymorphism (SSCP), Terminal-Restriction Fragment
417 Length Polymorphism (T-RFLP) and Ribosomal Intergenic Spacer Analysis (RISA). These
418 methods are based on the analysis of PCR products amplified from 16S rRNA genes (DGGE,
419 SSCP and T-RFLP) or from the ribosomal intergenic region between 16S and 23S rRNA genes
420 (RISA). With these methods, a community fingerprint based on sequence polymorphism is
421 generated which gives an assessment of community structure and fluctuation over time in
422 ecological studies (Schmalenberger and Tebbe 2002; Rastogi and Sani 2011). Fingerprints of
423 the active part of a community may also be obtained after RNA extraction and reverse
424 transcription of the 16S rRNA (Delbès et al. 2000; Delbès et al. 2001). The possibility to get
425 fingerprinting patterns from both 16S rDNA and 16S rRNA confers a great advantage to
426 DGGE, SSCP and T-RFLP with respect to the RISA method. Other genes can be targeted with
427 fingerprinting methods and it is not restricted to ribosomal genes (Figure 3). In anaerobic
428 digestion, monitoring the activity of methanogens is a primary target. The *mcrA* gene carried by
429 methanogens, which encodes the key enzyme called Methyl Coenzyme M Reductase catalysing
430 the last step of methane formation, was established as a molecular marker (Lueders et al. 2001)
431 since all known methanogens express the Mcr enzyme (Ferry, 1999). Nowadays, a satisfactory
432 number of databank entries for *mcrA* gene are available, and sequence analyses reveal a huge
433 diversity of methanogens. The presence of this enzyme provides a reliable diagnostic of
434 methanogenesis in diverse environments (Reeve et al. 1997; Luton et al. 2002; Steinberg and
435 Regan 2009).

436

437

438 *4.1.1 DGGE as one of the most popular and used molecular tool for monitoring microbial*
439 *community structure and dynamics in bioreactors*

440 DGGE is one of the most popular and used techniques for biodiversity assessment in bioreactor
441 samples (Boon et al. 2002; Arooj et al. 2007; Connaughton et al. 2006; Miura et al. 2007;
442 Pereira et al. 2014; Ramos et al. 2014).

443 DGGE protocols are relatively simple and straightforward and produce results in relatively short
444 time: after extraction of genomic DNA or RNA, target gene sequences (mostly 16S rRNA gene
445 fragments, or functional genes) are amplified using specially designed primers with GC clamps
446 in the PCR reaction, and PCR amplicons of equal length are separated electrophoretically in a
447 denaturing gradient gel (Muyzer et al. 1993) (Figure 3). The interpretation of DGGE gels is
448 rapid and straightforward, if only band patterns and relative band intensities are of interest. This
449 renders this method popular for fast comparative analysis of communities from different
450 reactors (Boon et al. 2002; Ramos et al. 2014) or from the same reactor operated under different
451 conditions (Arooj et al. 2007; Miura et al. 2007; Luo and Angelidaki 2012; Ramos et al. 2013;
452 Shi et al. 2013; Pereira et al. 2012). Sequencing of the DNA bands excised from gels allows the
453 identification of different members of the microbial community and further phylogenetic
454 analysis (Arooj et al. 2007; Connaughton et al. 2006; Díaz et al. 2011; Luo and Angelidaki
455 2012; Ramos et al. 2013; Ramos et al. 2014; Pereira et al. 2012).

456 Nevertheless, DGGE as any other techniques has some disadvantages and limitations. Reports
457 on methodological difficulties of DGGE include issues such as comigration of DNA from
458 different species in the same band (Sekiguchi et al. 2001b; Speksnijder et al. 2001) and
459 formation of multiple bands in the amplification of genes from single genomes (Brosius et al.
460 1981; Nübel et al. 1996). Moreover, this method has limitations with respect to its resolving
461 power, as typically only up to 50 bands can be distinguished in a gel lane and it is clearly
462 limited to the dominant members of the microbial community (with a threshold of roughly 0.1%
463 of the total) (Van Elsas and Boersma 2011). In spite of these reports, many researchers employ
464 this technique for the identification of important community members and dominant organisms
465 found in bioreactors.

466 The number of bands is assumed to accurately reflect the diversity of microbes in the sample
467 (Miura et al. 2007; Luo and Angelidaki 2012; Ziembinska-Buczynska et al. 2014) whilst the
468 relative intensity of each band is thought to reflect the relative abundance of the particular
469 organism represented by the band in the community (Dar et al. 2005; Connaughton et al. 2006;
470 Díaz et al. 2006; Shi et al. 2013).

471 Since identification of important community members as well as numerically most dominant
472 members of a community is a key aspect of microbial community analysis of bioreactor
473 samples, it is very important to know whether DGGE is a reliable technique to obtain such
474 community data. In a systematic study conducted by Araujo and Schneider (2008) with artificial
475 consortia, this technique was tested under conditions where results would not be affected by
476 differences in DNA extraction efficiency from cells. They demonstrated that DGGE was
477 suitable for identification of all important community members in the three-membered artificial
478 consortium, but not for identification of the dominant organisms in this small community.
479 Multiple DGGE bands obtained for single organisms with the V3 primer pair (targeting 16S
480 rRNA) could greatly confound interpretation of DGGE profiles.

481 Despite all the limitations mentioned above, PCR-DGGE has turned into a routine
482 fingerprinting method to study microbial diversity because it has a reasonable resolving power
483 and allows fast comparative analysis between different samples (Van Elsas and Boersma 2011).

484 DGGE has been used to analyze the biodiversity of granular sludge (Roest et al. 2005; Zhang et
485 al. 2005), to determine the microbial composition and structure of different types of granules in
486 a UASB reactor from an industrial brewery (Díaz et al. 2006) and in a UASB reactor treating
487 domestic wastewater (Pereira et al. 2012), to compare the start-up and evolution of mesophilic
488 and thermophilic UASB reactors (LaPara et al. 2000; Liu et al. 2002), to investigate changes in
489 the microbial population in continuously stirred tank reactors (Rincon et al. 2006; Ueno et al.
490 2001), and to analyze the microbial population dynamics in anaerobic reactors treating landfill
491 leachate or trichloroethane (Calli et al. 2006; Tresse et al. 2005). DGGE was also used to
492 investigate the effect of operating temperatures on the microbial community profiles in a high
493 cell density hybrid anaerobic reactor (Kundu et al. 2012), to analyze the effect of the carbon
494 source on the microbial community structure and performance of an anaerobic reactor (Kundu
495 et al. 2013a), and to evaluate changes in microbial communities in a hybrid anaerobic reactor
496 with organic loading rate and temperature (Kundu et al. 2013b).

497 The correlation of microbial community structure with wastewater composition and reactor's
498 performances was investigated by Kundu et al. (2013a). Self-immobilized granules were
499 developed in synthetic wastewaters based on different carbon sources (glucose, sugarcane
500 molasses, and milk), in three hybrid anaerobic reactors operated at 37°C. To study archaeal
501 community structure, a polyphasic approach was used with both qualitative and quantitative
502 analysis. While DGGE of 16S rRNA gene did not reveal major shifts in diversity of *Archaea*
503 with change in substrate, quantification of different groups of methanogens and total bacteria by
504 real-time PCR showed variations in relative abundances with the dominance of
505 *Methanosaetaceae* and *Methanobacteriales*. These data were supported by differences in the
506 ratio of total counts of *Archaea* and *Bacteria* analyzed by catalyzed reporter deposition –

507 fluorescence in situ hybridization. During hydraulic and organic shocks, the molasses-based
508 reactor showed the best performances followed by the milk-and glucose-based reactors. This
509 study indicates that carbon source shapes the microbial community structure more in terms of
510 relative abundance with distinct metabolic capacities rather than its diversity.

511 Ramos et al. (2014) used DGGE to evaluate the impact of O₂ on microbial communities in an
512 industrial-pilot scale sewage sludge reactor. Ziembinska-Buczynska et al. (2014) analyzed the
513 bacterial community performing methanogenesis in a pilot scale anaerobic chamber during the
514 shift from mesophilic to thermophilic conditions to point at the group of temperature tolerant
515 microorganisms and their performances, by using DGGE. Taken together, these examples
516 indicate that DGGE has turned into a routine fingerprinting method and is still one of the most
517 used molecular methods to study microbial diversity in anaerobic reactors.

518

519 *4.1.2 T-RFLP a molecular tool to monitor bacterial and archaeal communities in anaerobic* 520 *reactors*

521 T-RFLP emerged as a molecular tool for the study of microbial communities when Liu (1997)
522 adapted an existing method (RFLP) in order to perform a rapid analysis of complex
523 environments. T-RFLP analysis involves amplification of target genes from whole-community
524 DNA extracts by using specific primer pairs, one of which is fluorescently labeled.
525 Subsequently, amplicons are digested with restriction enzymes, typically using 4-base cutters,
526 and fragments are size separated and detected on automated sequencers (Figure 3). As only
527 terminal fragments (T-RFs) are detected the complexity of the profiles is considerably reduced.
528 The polymorphism depends on the fragment length and an internal size standard labeled with a
529 different fluorophore running with each sample allows a precise assignment of the fragment
530 lengths with single base pair resolution. The patterns of T-RFLP for a sample can be converted
531 into a numerical form and allows, after applying standardization procedures, to compare
532 samples using a variety of statistical approaches. In this way, by using T-RFLP profiles,
533 questions related to the diversity and dynamics of complex ecosystems might be revealed.
534 Moreover, a presumptively phylogenetic assignment of T-RFs might be possible relating the T-
535 RFLP profiles to *in silico* restriction of OTUs obtained from cloning libraries of the same or
536 related samples. This is especially useful in simple ecosystems or when a specific population is
537 targeted, where the number of *in silico* T-RFs retrieved from the OTUs of a cloning library is
538 low. For more complex communities, the taxonomic assignment for all T-RF might not be
539 possible. Nonetheless, the predominant T-RFs are generally identified. One shortcoming is that
540 discrepancies between *in silico* and observed T-RF sizes might occur which makes the T-RF
541 assignment difficult (Schütte et al. 2008). Another disadvantage of T-RFLP is the occurrence of

542 pseudo T-RFs which leads to an overestimation of the diversity (Egert and Friedrich 2003). In
543 order to improve the reproducibility and accuracy of the method and the taxonomic assignment
544 of T-RFs several technical improvements have been suggested, eg. the use of more than one
545 restriction enzyme (Osborne et al. 2006; Padmasiri et al. 2007; Collins et al. 2003), an
546 appropriate choice of restriction enzymes and primers (Schütte et al. 2008) and labeling both
547 forward and reverse primers with two different fluorochromes (Collins et al. 2003). Technical
548 improvements have also been proposed in the last years to improve data analysis (for more
549 detailed information see Nocker et al. 2007; Schütte et al. 2008; Prakash et al. 2014).

550 Even though T-RFLP is a very simple method, data analysis and taxonomic T-RF assignment
551 might require an additional effort. Several steps are needed before the data can be statistically
552 analyzed. First, a fluorescence threshold needs to be established to remove baseline noise.
553 Second, standardization between samples is needed to avoid the influence of variations in DNA
554 concentration between samples. Finally, size binning is necessary between samples to be able to
555 compare them. Further analysis is needed using database information or clone sequence *in silico*
556 restriction for taxonomic T-RF assignment. As a consequence of the time-consuming analysis of
557 T-RFLP data, several web-based tools have been developed to aid in T-RF assignment and
558 identify plausible members of a microbial community based on T-RFLP data. These tools are
559 also performing additional tasks such as profile comparison, statistical analysis of data and
560 representation of similarity in the form of a dendrogram. Some examples are TAP-TRFLP
561 (<http://rdp.cme.msu.edu>), torast (<http://www.torast.de>), MiCA (<http://mica.ibest.uidaho.edu/>)
562 and T-RFPred (<http://nodens.ceab.csic.es/trfpred/>), with phylogenetic assignment tools (PAT)
563 likeT-Align (http://inismor.ucd.ie/~*talign/), ARB software integrated tool, TRF-CUT
564 (<http://www.mpi-marburg.mpg.de/braker/trfcut.zip>), TRiFLe
565 (<http://cegg.unige.ch/trifle/trifle.jnlp>), and with T-RFLP statistical data analysis software
566 (http://www.ibest.uidaho.edu/tool/T-RFLP_stats/index.php). Afterwards, statistical analysis can
567 be performed with standard softwares in order to interpret the results.

568 T-RFLP has been extensively used to monitor structural changes and dynamics of *Archaea* and
569 *Bacteria* in anaerobic reactors (Collins et al. 2003; Carballa et al. 2011; Pycke et al. 2011; Feng
570 et al. 2010; Pervin et al. 2013; Zhang et al. 2014; Whang et al. 2015). In general, the community
571 is monitored in order to relate changes in the microbial community by modifying one or several
572 operational parameters. Bacterial and archaeal communities can be monitored in lab scale or full
573 scale reactors and different statistical tools might be used to interpret T-RFLP results. In this
574 sense, Blasco et al. (2014) studied changes in bacterial and archaeal communities by only using
575 T-RFLP in two lab scale stirred tank reactors fed with autoclaved and untreated waste. Strong
576 statistical analysis of the data was performed including PCoA, ANOVA and Spearman's rank
577 correlation coefficient. The authors were able to determine that autoclaving as a pretreatment as

578 well as change of OLR influenced the microbial community structure, especially the bacterial
579 one. Another investigation used T-RFLP to study the microbial community richness, dynamics,
580 and organization of four full-scale anaerobic digesters during a time-course study of 45 days
581 (Pycke et al. 2012). In this case the authors calculated the parameters established by Marzoratti
582 et al. (2008) (Richness (Rr), Functional organization (Fo), Dynamics (Dy)) and performed
583 cluster analysis. The authors demonstrated that the temperature has a strong effect on both
584 bacterial and archaeal communities. The dynamics of change was very high and varied for both
585 *Bacteria* and *Archaea* at a rate of change between 20–50% per 15 days. Moreover, the
586 community organization of the *Bacteria* changed more in the thermophilic reactors, compared
587 with the mesophilic ones and the *Archaea* community structure was more stable.

588 Several authors have used T-RFLP to determine the community structure and dynamics of
589 methanogens, one of the key populations of the anaerobic process in anaerobic digesters
590 (McHugh et al. 2004; Padmasiri et al. 2007; Ziganshin et al. 2013). By choosing the correct
591 restriction enzymes it is possible to study the relation between hydrogenotrophic and
592 acetoclastic methanogens. Padmasiri et al. (2007) were able to determine that hydrogen utilizing
593 methanogens increased in abundance during a period of poor reactor performances when
594 studying the effect of the degree of shear to which the biomass in an AnMBR is exposed. The
595 authors used three restriction enzymes and were able to study the dynamics of
596 *Methanomicrobiales* and *Methanobacteriales* (hydrogenotrophic) and *Methanosaetaceae* and
597 *Methanosarcinales* (acetoclastic) without performing a cloning library for T-RF assignment.
598 When combining T-RFLP (using one enzyme but forward and reverse primers labeled) with
599 cloning and sequencing a more exact determination of the structure, diversity, dynamics and
600 composition of the methanogenic community was achieved. Xu et al. (2010) studied the effect
601 of two methanogenic inhibitors, BES and CHCl_3 , on the methanogenic community structure
602 present in an anaerobic sludge. The phylogenetic analysis of the archaeal sequences obtained
603 from the clone library allowed determining the archaeal community composition which
604 consisted of *Methanosaetaceae*, *Methanomicrobiales*, and *Methanobacteriales* and of yet
605 uncultured archaeal lineages such as RC-III. Correspondingly, six fragments were detected as
606 major peaks in the T-RFLP profiles. *In silico* digestion of the representative clones allowed to
607 assign three T-RFs to *Methanomicrobiales*, one T-RF to *Methanobacteriales* and one T-RF to
608 *Methanosaetaceae*. Finally, one T-RF was associated with two lineages, which could be
609 differentiated by forward digestion. By using this strategy the authors were able to determine
610 that acetoclastic methanogens were more sensitive to inhibitors than hydrogenotrophic
611 methanogens as the relative abundance of *Methanosaetaceae* decreased compared to the control
612 experiment and the replacement of acetoclastic methanogens was more pronounced in the
613 CHCl_3 versus the BES incubations.

614 Another interesting option used to study methanogens is to perform T-RFLP profiles on the
615 methyl coenzyme M reductase gene (*mcrA*) (Zhang et al. 2014; Ma et al. 2013; Ács et al. 2013).
616 One recent example revealed which methanogens were involved in tetramethylammonium
617 hydroxide (TMAH) degradation in three full scale bioreactors (Whang et al. 2015).

618 T-RFLP has also been used to monitor microbial communities present in lab scale hydrogen
619 producing reactors, combined with sequencing and strain isolation, and it has been shown that a
620 mixed microbial community developed while non hydrogen producing bacteria were also
621 present in the reactors (Castelló et al. 2009; Perna et al. 2013; Ferraz et al. 2014). For example,
622 when using raw cheese whey as substrate a low hydrogen yield was obtained which could be
623 explained by the selection of a mixed fermentative population with the presence of hydrogen-
624 producing organisms (*Clostridium*, *Ruminococcus* and *Enterobacter*) and other non-hydrogen
625 producing fermenters (*Lactobacillus*, *Dialister* and *Prevotella*) in the same reactor (Castelló et
626 al. 2011). Knowing how these organisms outcompete might be important in order to further
627 optimize hydrogen production in UASB reactors. These examples demonstrate that T-RFLP has
628 become a rapid, inexpensive, sensitive, robust and reproducible technique for the study of
629 bacterial and archaeal communities in anaerobic bioreactors.

630

631 *4.1.3 SSCP fingerprinting as a rapid and accurate molecular tool for monitoring microbial* 632 *dynamics in anaerobic bioprocesses*

633 The SSCP-based analysis relies on the propensity for single-strand DNA (ssDNA) to take a
634 folded secondary structure in non-denaturing conditions. The secondary structure is determined
635 by the nucleotide sequence and intramolecular interactions. Therefore, the SSCP method
636 involves a step of heat denaturation of the PCR products where the ssDNA is formed. An
637 electrophoresis is then performed under non-denaturing conditions, where ssDNA fragments are
638 separated into different bands according to their difference in electrophoretic motility (Lee et al.
639 1996; Delbès et al. 2001). Nowadays, the procedure was adapted to an automated capillary
640 DNA sequencer and is so-called Capillary Electrophoresis-SSCP (CE-SSCP) (Zumstein et al.
641 2000). An internal SSCP ladder is added in each sample to compare data and eliminate
642 capillary-to-capillary or run-to-run variability, which has been a real technical improvement as
643 compared to the original SSCP analysis (Quéméneur et al. 2011b). Since automation of the CE-
644 SSCP electrophoresis allows a rapid and easy process of a maximum of 96 samples at a time,
645 CE-SSCP fingerprinting is particularly useful to study the dynamics of the microbial
646 populations over a long time series (Zumstein et al. 2000). Considering this, further statistical
647 analysis can be carried out. Usually, CE-SSCP are processed and computed either with the

648 graphical interface Stat-Fingerprints program working under the R software environment
649 (Michelland et al. 2009) or the SAFUM toolbox written in MATLAB (Zemb et al. 2007).

650 The V3 variable region of the 16S rRNA gene is one of the most common target for
651 characterizing archaeal and bacterial communities in anaerobic digesters (Zumstein et al. 2000;
652 Leclerc et al. 2001; Leclerc et al. 2004; Mnif et al. 2012; Jáuregui-Jáuregui et al. 2014). In
653 literature, SSCP analysis is widely reported as useful to compare microbial diversity in response
654 to changes in environmental or process operation conditions especially when the diversity is low
655 (Leclerc et al. 2004; Ye et al. 2007). For instance, the archaeal diversity in anaerobic digesters is
656 often limited to a dozen of peaks, and diversity can be rapidly characterised by simple peak
657 counting on SSCP profiles (Leclerc et al. 2004). Bacterial diversity may also be monitored by
658 SSCP in anaerobic ecosystems that are simplified by applying stringent conditions, *e.g.* in
659 hydrogen-producing bioreactors where only few major bacterial species dominate and are
660 mostly affiliated to the *Clostridium* genus (Rafrafi et al. 2013). Similarly, the CE-SSCP analysis
661 is perfectly adapted to monitor the decrease of microbial diversity all along an enrichment
662 procedure, *e.g.* from anaerobic sludge to a co-culture of anaerobic syntrophs (Trably et al.
663 2008). The diversity of more complex bacterial community can also be estimated but only when
664 considering the background signal corresponding to the information that is not enclosed in
665 peaks areas (Loisel et al. 2006).

666 Interestingly, Bauer et al. (2008) developed and validated successfully a functional PCR-SSCP
667 method followed by a cloning step of the interesting bands and a sequence analysis. They
668 evaluated this method in combination with direct PCR cloning and sequence analysis. A novel
669 and highly degenerated *mcrA/mrtA* pair of PCR-primers targeting all known methanogens was
670 developed and tested with reference methanogens. DNA extracts from biogas fermenters fed
671 with maize silage and operated at different conditions were analyzed and the dominant residing
672 methanogens were characterized to assess their functional importance during the process stages.
673 Whereas direct PCR cloning was more suited to determine the relative abundances of
674 methanogens, the SSCP analysis easily detected population changes and the involved sequences
675 specifically. Although most of the bands were not sharp probably because of high degenerated
676 primers, the authors concluded that the SSCP technique was more suitable than DGGE for
677 monitoring methanogens. The SSCP method was further refined and employed to investigate
678 the presence and dynamics of distinct sub-populations of methanogens during biogas process
679 acidification (Munk et al. 2010). The authors concluded that technical details associated with
680 degeneration of the primers should be improved since different sequences comigrated in one
681 SSCP band and identical sequences were found in different bands. This method, in combination
682 with classical chemical parameters, could help to quickly indicate changes in population
683 structure due to process conditions, and thus prevent dysfunction of the process.

684 The correlation between the expression of *hydA* gene and H₂ production in batch or continuous
685 reactors has been well established (Tolvanen et al. 2008a; Wang et al. 2008a; Tolvanen et al.
686 2008b; Wang et al. 2008b). Quéméneur et al. (2010) developed a successful CE-SSCP method
687 based on functional *hydA* [Fe-Fe]-hydrogenase genes for monitoring hydrogen (H₂)-producing
688 clostridial populations. These authors designed and validated a set of non-degenerated primers,
689 named *hydAClosF/hydAClosR*, by binding two conserved regions of the active site domain (H-
690 cluster) of the [Fe-Fe]-H₂ases-coding genes. These primers showed a high *in silico* specificity
691 and selectivity for the detection of the *hydA* genes from a broad range of *Clostridium* strains
692 belonging to different phylogenetic clusters. This method was validated using experimental H₂-
693 producing mixed cultures and was shown to be very effective in monitoring the shift of H₂-
694 producing bacterial populations in relation to environmental changes, such as temperature
695 (Quéméneur et al. 2010), pH (Quéméneur et al. 2011b) and the type of carbohydrate as substrate
696 (Quéméneur et al. 2011a). The *hydA* sequences amplified with this new set of primers clearly
697 showed that the functional diversity of *hydA*-carrying *Clostridium* sp. strains impacted the H₂
698 production yields. Interestingly, the 16S rDNA-based fingerprints were found to be less
699 correlated to changes in H₂ production performances. The difficulty in distinguishing
700 *Clostridium* cluster I species in H₂ dark fermentation systems by using universal bacterial
701 primers was also reported by (Chang et al. 2008).

702 Hence, functional CE-SSCP tools provide highly reliable and throughput analysis of the
703 microbial communities which could be used as a complement to 16S rDNA phylogenetic
704 markers. Such first successful approach for determining functional microbial dynamics in
705 bioprocesses by the CE-SSCP technique makes possible its rapid extent to other functional
706 groups found in anaerobic digestion systems, e.g. methanotrophic (*pmoA*), and sulfate reducing
707 (*dsrAB*) bacteria.

708

709 The three fingerprinting methods presented here are highly suitable to study the microbial
710 diversity, structure and dynamics of microbial communities in AD reactors. Overall, the CE-
711 SSCP method offers several advantages in terms of rapidity and ease of obtaining data, as well
712 as a high reproducibility between runs and data processing. Regarding the DGGE-based
713 methods, which relies on the use of clamped primers and gradient gels, with only a limited
714 number of samples per gel that are analysed, SSCP and T-RFLP are simpler and more straight
715 forward techniques (Marzorati et al. 2008; Rastogi and Sani 2011, Schütte et al. 2008).
716 However, DGGE is widely used and has turned into a routine fingerprinting method to study
717 microbial diversity because it has reasonable resolving power and allows fast comparative
718 analysis between different samples (Van Elsas and Boersma 2011). On the other hand, T-RFLP
719 has been shown to be a reproducible and accurate tool for community fingerprinting (Osborn et

720 al. 2000) but, the data analysis requires an additional effort. Additionally, even though DGGE is
721 widely used, special equipment, but very affordable, is required while for SSCP and T-RFLP
722 the samples can be analysed in any sequencing service commercially available.

723

724 **5 – Tools for microbial quantification (How many microorganisms from the different** 725 **groups are present in the reactor?)**

726

727 *5.1 Quantification and detection of particular organisms in anaerobic digesters by FISH*

728 The FISH method involves application of fluorescently labeled probes to ribosomal rRNA in
729 permeabilized whole microbial cells. The probes consist of short pieces of DNA (usually 15-25
730 nucleotides in length) that are designed to specifically hybridize to their complementary target
731 sequence on the rRNA structures (16S and 23S subunits are typically used for *Bacteria* and
732 *Archaea*) (Amann et al. 1995). From the composition of the probe, it is possible to design it to
733 specifically target a narrow phylogenetic group (down to the species level) or any other higher
734 phylogenetic hierarchical group (Amman et al. 1995). No probes will hybridize to those cells
735 without target sequences. Cells containing the target sequence will on the other hand retain the
736 hybridized probe and due to the large number of ribosomes only the active cells (from 10^3 to
737 10^5 ribosomes) become fluorescently labeled. FISH provides an effective means of identification
738 and qualitative and/or quantitative microbial population analysis in natural and engineered
739 environments (Amann et al. 2001).

740 The optimisation process in FISH method is generally carried out with pure cultures of
741 microorganisms that contain rRNA with a perfect match to the probes (positive controls) and
742 with microorganisms that contain rRNA with preferably one (optimally centrally located) or
743 several mismatches to the probe (negative controls) (Manz et al. 1992). There are several
744 variables that can be exploited to modify the strength of a hybrid between an oligonucleotide
745 and its perfect target. These include temperature, ionic strength of the medium and organic
746 solvent concentration (Amann et al. 1995).

747 Several researches applied FISH in order to study microbial ecology in aerobic and/or anaerobic
748 reactors (Amann et al. 1992; Raskin et al. 1994; Power et al. 1999; Rocheleau et al. 1999;
749 Sekiguchi et al. 1999; Araujo et al. 2000; Araujo et al. 2004; Imachi et al. 2000; Zheng and
750 Raskin 2000; Egli et al. 2003; Ginige et al. 2004; Díaz et al. 2006). One of the early studies
751 (Raskin et al. 1994) is very important to mention because authors designed different
752 oligonucleotide probes for ex situ hybridization in order to quantify different groups of
753 methanogenic *Archaea*. The second study, which is also worth to mention, was of Sekiguchi et
754 al. (1999). This was the first study that used FISH combined with confocal laser scanning

755 microscopy (CLSM) to elucidate the spatial distribution of microbes within two types of
756 methanogenic granular sludge (mesophilic and thermophilic) in UASB reactors fed with
757 sucrose-, acetate-, and propionate-based artificial wastewater. In situ hybridization with
758 archaeal- and bacterial-domain probes within granule sections clearly showed that both
759 mesophilic and thermophilic granules had layered structures and that the outer layer harbored
760 mainly bacterial cells while the inner layer consisted mainly of archaeal cells. *Methanosaeta*-,
761 *Methanobacterium*-, *Methanospirillum*-, and *Methanosarcina*-like cells were detected with
762 oligonucleotide probes specific for the different groups of methanogens, and they were found to
763 be localized inside the granules, in both types of which dominant methanogens were members
764 of the genus *Methanosaeta*.

765 In the study of Crocetti et al. (2006), 3000 Euryarchaeota 16S rRNA gene sequences were
766 phylogenetically analyzed and previously published oligonucleotide probes were evaluated for
767 target accuracy. Where necessary, modifications were introduced or new probes were designed.

768 Despite all the studies mentioned above that used FISH, there are some limitations associated
769 with this method, such as: 1-poor cell permeability (a pretreatment is necessary to fix Gram
770 positive cells or some archaea (Davenport et al. 2000; Zheng and Raskin 2000). 2-The detection
771 limit is not very low (varies from 10^3 cells/mL (Li and Gu 2011) to 10^4 cells /mL (Amann
772 1995). 3-Insufficient ribosome content, as inactive or dormant cells will not give fluorescent
773 signal or it can be very low (Morgenroth et al. 2000). 4- Sample autofluorescence, specially
774 some methanogenic *Archaea* can produce autofluorescence that interfere with fluorescent
775 labels.

776 Some of these limitations, especially the detection of cells with low ribosomal content present in
777 oligotrophic conditions, can be overcome by applying the catalyzed reporter deposition FISH
778 (CARD-FISH) (Pernthaler et al. 2002). In this approach (CARD-FISH) the critical step is the
779 diffusion of the large molecules, in this case the horse radish peroxidase (HRP)-labelled probe,
780 into whole cells embedded in an agarose matrix. A directed permeabilization of prokaryotic cell
781 walls prior to the hybridization step is of crucial importance to enable the penetration of the
782 probe (Pernthaler et al. 2002). In the study of Wilhartitz et al. (2007), CARD-FISH resulted in
783 substantially higher recovery efficiency than the conventional FISH-approach and therefore it is
784 more suitable for the enumeration of specific prokaryotic groups in ultra-oligotrophic
785 environments such as ground- and drinking water samples.

786 Therefore, knowledge of the nature and applicability of the sample as a uniform protocol with
787 application of the proper controls are of fundamental importance for obtaining solid and
788 comparable information. By using FISH it is possible to observe the morphology and to
789 quantify numbers of bacteria or equivalent biovolume. Although the method is very time

790 consuming and tedious, the principal advantage is that it is possible to quantify and at the same
791 time, determine the position of the different cells in the community.

792

793 *5.2 Q-PCR for quantification in anaerobic digestion processes*

794 The quantitative polymerase chain reaction (q-PCR) or real-time quantitative PCR (RT-PCR) is
795 a molecular technique that detects and quantifies DNA molecules in solution, by means of
796 fluorescence signal detection. Based on the end point or regular PCR, a target region of the
797 template DNA is amplified with specific primers, although a key difference is that products are
798 detected in each cycle during the exponential phase of the process (Higuchi et al. 1992).

799 The two detection methods most widely used in environmental microbiology are: i) the addition
800 of intercalating fluorescent dyes (e.g. SYBR Green I), which emits fluorescence only when
801 binds to double-stranded DNA (Wittwer et al. 1997) or, ii) sequence-specific oligonucleotide
802 probes (complementary to an internal sequence) labeled with a fluorophore and with a quencher
803 (e.g. TaqMan) (Gudnason et al. 2007). The TaqMan technology (Heid et al. 1996) is quite more
804 specific than SYBR Green I due to the third oligonucleotide probe used in the reaction that
805 drastically reduces false positives. On the other hand, SYBR Green I is a less expensive method
806 and sometimes the only choice, when the target sequence hinders the primer and probe design.

807 Hence, the quantification of the fluorescent signal can be correlated to the amount of target
808 DNA molecules present in the solution. Quantification can be absolute or relative; the former
809 method is currently the most used in microbial ecology and consists in the interpolation of the
810 inquired samples in a standard curve constructed with known concentrations of target
811 sequences, often cloned in plasmids or as PCR products.

812 The specificity, sensitivity but mainly the quantitative basis of the q-PCR filled a gap in
813 microbial ecology. Since the pioneer studies in 2000s (Suzuki et al. 2000; Takai and Horikoshi
814 2000), this approach has been increasingly applied to investigate microbial ecology questions
815 that could not be addressed with other techniques such as clone libraries or even the actual high-
816 throughput DNA sequencing. It has been used to track genes and/or transcripts of ribosomal or
817 functional genes across temporal or spatial scales in different environments such as biofilms (de
818 Gregoris et al. 2011), rumen (Bekele et al. 2011) seawater (Mincer et al. 2007) and alpine soils
819 (Brankatschk et al. 2011). A recent work published by Kim et al. (2013) describes the state of
820 the art in the quantification of the different groups of microorganisms found in wastewater
821 treatment systems.

822 In the anaerobic wastewater treatment research field, q-PCR has been applied mainly to monitor
823 the methanogen community. The strategy consists in targeting the 16S rRNA gene (Yu et al.
824 2005) or the *mcrA* gene (Alvarado et al. 2014). By targeting the 16S rRNA gene, some studies

825 have shown a switch from acetoclastic to hydrogenotrophic methanogenesis when stressing
826 conditions such as temperature or ammonia are applied to lab-scale reactors (Lee et al. 2009;
827 Mc Keown et al. 2009; Song et al. 2010; Traversi et al. 2011; Jang et al. 2014; Town et al.
828 2014). On the other hand, a direct correlation between specific methanogenic activity and *mcrA*
829 gene copy numbers was reported recently (Morris et al. 2013).

830 Some considerations must be stressed before interpreting or comparing q-PCR results. The first
831 issue to consider is the DNA yield bias introduced by the different methods of nucleic acids
832 extraction (Martin-Laurent et al. 2001). Another important aspect is the operon copy number
833 variation (CNV) of the target gene, in the case of bacterial 16S rRNA, variations of 1-15 copies
834 between different species or even in different phases of the cell cycle were reported
835 (Klappenbach et al. 2000; Sun et al. 2013). In the case of organisms from the *Archaea* domain,
836 ribosomal operon variation is less pronounced, ranging 1-4 copies per genome (Kim et al.
837 2013). As an attempt to cope with the gene CNV, alternative molecular markers with only one
838 copy per genome have been used. The *mcrA* gene, mentioned above has been proved to be a
839 good molecular marker for tracking methanogens in the archaeal community (Luton et al.
840 2002).

841 The specificity of the primers and probes used is another key aspect of this approach. However,
842 there is a vast set of primers designed to monitor the wastewater treatment microbiology. For
843 instance, a complete review of primers and probes designed to monitor methanogenic *Archaea*
844 have been compiled by Narihiro and Sekiguchi (2011).

845 Due to the potential applications of this technique in wastewater treatment research, efforts must
846 be done to jointly improve the robustness of the approach and standardize protocols to compare
847 results between different works. Although far from a golden standard, q-PCR is currently a
848 promising technique to assess quantitatively the structure of microbial communities. The
849 principal advantages of this quantification technique are the low detection limits (it is possible
850 to detect a single DNA molecule depending of the efficiency of the reaction), the possibility to
851 perform several samples at the same time and the short time required to obtain results. q-PCR
852 analysis give access to absolute quantification, and thus complement very well sequencing and
853 FISH data that give relative abundances of targeted microorganisms.

854

855 **6. Tools to analyze microbial function (What is the function of the microorganisms inside** 856 **the reactor?)**

857 Understanding how complex microbial communities function inside anaerobic digesters and
858 how environmental conditions may affect interspecies relationships is currently one of the big
859 challenges for both environmental biotechnologists and microbiologists. Only by getting more

860 insight into microbial function we will be able to fully exploit and manage microbial
861 communities for biotechnology purposes (Verstraete 2007). Over the last years, several culture-
862 dependent and culture-independent techniques have been developed that allow to study
863 community composition and function in complex ecosystems, such as detection of specific
864 enzyme activity or of specific enzyme-coding genes, functional microarrays, radioactive and
865 stable isotope labeling, probing and advanced imaging, and meta-omics approaches (Su et al.
866 2012; Vanwonterghem et al. 2014). These techniques, alone or in combination with each other
867 and with differential bioreactor operation, can offer insights into the metabolic activities of
868 microbial communities in AD processes and it is in this perspective that they are discussed in
869 the following sections.

870

871 *6.1 Stable isotope probing and in situ advanced microscopic methods*

872 Stable isotope probing (SIP) is performed by amending a stable isotope (for example ^{13}C -
873 labeled substrate to microbial communities and further analyzing microbial composition in
874 heavy DNA or rRNA fraction, *e.g.* by cloning and sequencing (Figure 4a) (Radajewski et al.
875 2000). Heavy DNA/rRNA represents the active populations in the community because those are
876 the ones incorporating the isotopes within their biomass.

877 In recent years, with advances in high resolution imaging and spectroscopy, SIP has been
878 complemented with other techniques, such as FISH, microautoradiography combined with FISH
879 (MAR-FISH), Raman microspectroscopy combined with FISH (Raman-FISH), or nanoscale
880 secondary ion mass spectrometry (NanoSIMS) combined with FISH (SIMSISH) (Chapleur et
881 al. 2013; Huang et al. 2007; Lee et al. 1999; Li et al. 2008). These techniques allow to study
882 taxonomic identity and activity of microbial communities at single cell resolution (Figure 4b).
883 This is a field of fast evolution and complementary techniques are rapidly emerging. One
884 example (not yet applied to anaerobic digestion systems) is the utilization of amino acid tagging
885 and click chemistry for *in situ* visualization of newly synthesized proteins (biorthogonal non-
886 canonical amino acid tagging, BONCAT) (Hatzenpichler et al. 2014). Combination of
887 BONCAT technique with FISH (BONCAT-FISH) enables a direct link between taxonomic
888 identity and translational activity. Comparison of BONCAT-FISH results and SIMISH for
889 anabolic nitrogen assimilation showed concordance, with BONCAT-FISH offering a less
890 expensive solution for these studies.

891 Several studies have addressed the key players in anaerobic digesters by performing separate
892 feeding studies with different ^{13}C -labelled carbon sources, such as ^{13}C -cellulose, ^{13}C -glucose,
893 ^{13}C -acetate and ^{13}C -propionate, followed by DNA-SIP and FISH analysis (a summary of the
894 main results obtained in these studies is shown in Table 2). In this way, the metabolic activity

895 for independent steps in the cellulose/glucose to methane conversion pathway in methanogenic
896 reactors could be assigned to specific taxonomic groups, *e.g.* cellulose degraders were shown to
897 belong to *Clostridia* and *Acetivibrio*, while subsequent glucose conversion was mainly
898 performed by bacteria belonging to the *Clostridia* and *Porphyromonadaceae* (Li et al. 2009; Li
899 et al. 2008). The involvement in cellulose conversion of novel identified groups in anaerobic
900 digesters, such as the WWE1 candidate division, could also be confirmed using these methods
901 (Limam et al. 2014). The strategy to perform feeding studies with different ¹³C-labelled
902 substrates, corresponding to the different levels of the conversion pathway, was also
903 successfully applied in elucidating the microbial groups involved in each of the sequential steps.
904 In a glucose-degrading anaerobic digester it was elegantly demonstrated that glucose was
905 converted to propionate by members belonging to *Actinobacteria*, *Bacteroidetes* and
906 *Chloroflexi*, and that propionate was further converted to acetate by *Syntrophobacter* and
907 *Smithella*, and acetate finally utilized by members belonging to the *Synergistes* group 4 and the
908 methanogen *Methanosaeta* (Ito et al. 2011; Ito et al. 2012).

909

910 6.2 The meta-omics era

911 Omics techniques (genomics, transcriptomics, proteomics) were first used to study and
912 characterize cultivable isolates but, with the development of more powerful instrumentation and
913 bioinformatics tools, application of these techniques has been extended to more complex
914 ecosystems. This is important because interspecies interaction and responses to environmental
915 conditions, which result in multiple community behaviors, can only be identified if the
916 ecosystem is assessed as a whole. Meta-omics techniques - metagenomics, metatranscriptomics,
917 metaproteomics, and metabolomics - can be applied to ecosystems to unravel composition and
918 function of uncultured microorganisms (Figure 5). Methods such as SIP and advanced imaging
919 techniques discussed in the previous section allow analyzing structure-function in ecosystems,
920 but only associated with a particular physiological trait. Omics techniques, on the other hand,
921 have the potential to reveal the complete picture of the microbial functionalities in an
922 ecosystem.

923

924 6. 2. 1 Metagenomics

925 Metagenomics has greatly facilitated the mining of microbial diversity and metabolic potential
926 of highly diverse microbial communities. In particular, in recent years the advances in NGS
927 technology and accompanying bioinformatic tools have enabled the generation of large
928 sequence datasets and streamlined analysis of the complex sequence data. Through these
929 methods, in-depth 16S rRNA gene-sequencing analysis of production-scale biogas digesters has

930 generated many novel details on the microbial composition (and compositional changes) during
931 anaerobic digestion. These studies confirmed that *Clostridia* comprise the most prevalent
932 bacterial class in different types of AD reactors (Jaenicke et al. 2011; Li et al. 2013;
933 Rademacher et al. 2012; Schlüter et al. 2008), but also identified hitherto neglected taxa such as
934 *Streptococcus*, *Acetivibrio*, *Garciella*, *Tissierella*, *Gelria* (Jaenicke et al. 2011) or
935 *Psychrobacter* and *Anaerococcus* (Li et al. 2013). Methanogenic archaea were mainly
936 represented by the hydrogenotrophic *Methanoculleus*, *Methanosarcina*, *Methanobacterium* and
937 *Methanothermobacter*, or the acetoclastic archaea *Methanosaeta*. This corresponded to results
938 from previous studies. In addition one study also demonstrated the presence of
939 *Thermacetogenium* (Rademacher et al. 2012). The type strain of this genus, *Thermacetogenium*
940 *phaeum*, oxidizes acetate with *Methanothermobacter thermoautotrophicus* (Hattori et al. 2000).

941 Besides phylogenetic information, the metagenomics analysis also provides more details on the
942 functional potential of the microbial community and its correlation to the anaerobic process
943 under study. In AD reactors treating plant material waste a high number of sequence reads
944 related to cellulose and hemicellulose conversion as well as other carbohydrate-degrading genes
945 were detected, which in one study could be assigned to the predominant phylogenetic taxa
946 identified (Jaenicke et al. 2011). *Acetivibrio* species had previously been recovered in the heavy
947 DNA-fraction during SIP experiments with cellulose (Li et al. 2009); taxonomic and functional
948 genes results obtained by Jaenicke et al. (2011) support once more that this genus most probably
949 play a role in cellulose degradation. Some *Clostridia* may be involved in acetogenesis, as
950 deduced by mapping bacteria taxa to metagenome hits representing the Wood-Ljungdahl
951 pathway (Jaenicke et al. 2011). Reads encoding for enzymes required for hydrogenotrophic and
952 acetoclastic methanogenesis pathways were also detected, corresponding to the taxonomic
953 observations (Figure 6). Enzymes necessary for catalyzing CO₂ conversion to methane were all
954 found within the metagenomes of the bioreactor sludges treating cellulosic materials and waste
955 activated sludge (Rademacher et al. 2012; Wong et al. 2013). Acetate kinase and
956 phosphotransacetylase, involved in the first step of acetoclastic methanogenesis, were not
957 detected in these studies. Wong et al. (2013) proposed an alternative route in which an acetate
958 transporter coupled to acetyl-CoA synthetase and the hydrolysis of pyrophosphate by inorganic
959 pyrophosphatase drives this reaction forward, similar to the pathway present in *Methanosaeta*
960 *thermophila* (Smith and Ingram-Smith 2007).

961 One of the main drawbacks of metagenomics studies is still the large amount of sequence reads
962 with unidentified microbial origin or gene prediction. This may indicate that the anaerobic
963 digestion process is also conducted by yet unknown microorganisms or species for which no
964 genomic information is available. Most metagenomics studies rely on complete or draft
965 genomes to identify fragmentary sequences, which limits the ability to resolve metagenomics

966 data deriving from unknown microbial diversity (Temperton and Giovannoni 2012). Therefore,
967 it is important to continue the effort of isolating new microorganisms and sequencing the
968 genome of representatives of the different groups of microorganisms present in bioreactors,
969 following a similar strategy that was successfully applied with the human microbiote.

970

971 6. 2. 2 Metatranscriptomics

972 While metagenomics gives information on the metabolic and functional potential of a microbial
973 community, it cannot provide information on the activity of the genes present or to differentiate
974 between expressed and non-expressed genes. To get an estimate of the actual metabolic activity,
975 total mRNA from a microbial environment needs to be retrieved and sequenced.
976 Metatranscriptomics in general generates less complex datasets to analyze (only transcribed
977 genes are retrieved) than metagenomics does. However, mRNA molecules have a short half-life,
978 and ribosomal RNA represents the majority of RNA isolated, thereby reducing sequencing-
979 depth for mRNA reads that represent the expressed genes and pathways.

980 Recently, a metatranscriptome study investigated the transcriptionally active microbial fraction
981 of a biogas-producing reactor and identified the most abundant transcripts during biogas
982 production (Zakrzewski et al. 2012). This is the only metatranscriptomics study performed on a
983 full-scale anaerobic digester thus far. The most abundant mRNA-derived sequence reads from
984 this metatranscriptome dataset were from transcripts encoding enzymes involved in substrate
985 hydrolysis, acidogenesis, acetate formation and methanogenesis, indicating that these pathways
986 were highly active in the anaerobic digestion process. Taxonomic profiling of the active
987 community, based on 16S rRNA tags, revealed that members of the *Firmicutes* and
988 *Euryarchaeota* were the most abundant bacteria and archaeal phyla in the system, respectively;
989 low numbers of 16S rRNA tags were associated to the phyla *Bacteroidetes*, *Actinobacteria*, and
990 *Synergistetes*. In addition, comparison of the metatranscriptome dataset with 16S rRNA-gene
991 metagenomic sequence tags derived from the same community, indicated that the archaeal
992 species present were highly active and that in general the most abundant species in the
993 community contributed to the majority of the retrieved transcripts.

994

995 6. 2. 3 Metaproteomics

996 Where metagenomics identifies the microbial community and its metabolic/functional capacity,
997 and metatranscriptomics the transcriptional activity of specific metabolic pathways/microbes,
998 metaproteomics can be used to characterize highly expressed proteins within an environmental
999 microbial consortium. This can provide a more functional evidence of key metabolic pathways
1000 that are active and important for anaerobic digestion processes. In contrast to metagenomics and

1001 metatranscriptomics, the main limitation of metaproteomics is to extract a sufficient quantity of
1002 high quality proteins that are representative of the sample. Extracting a representative protein
1003 fraction for analysis is complicated due to the complex microbial communities involved, the
1004 presence of interfering compounds, and the heterogeneity of natural environments.
1005 Nevertheless, metaproteomics has a great potential to link the genetic diversity and activities of
1006 microbial communities with their impact on ecosystem function (Su et al. 2012).

1007 Metaproteomics studies in anaerobic digestion are still limited. One study analysing a lab-scale
1008 anaerobic bioreactor treating glucose-based wastewater assigned putative functions to the
1009 proteins detected, and demonstrated that they were mostly related to cellular processes such as
1010 methanogenesis from both CO₂ and acetate, glycolysis and the pentose phosphate pathway
1011 (Abram et al. 2011). The number of identified proteins was very low though, only 18 distinct
1012 proteins were assigned to different functions (from 70 proteins excised from two-dimensional
1013 electrophoresis gel). No metagenomics data was available from the reactor, which decreases the
1014 rate of identification. The protein assignment also indicated the presence of specific
1015 microorganisms in the bioreactor (*Actinobacteria*, *Firmicutes*, *Proteobacteria*,
1016 *Methanosarcinales* and *Methanobacteriales*), though little overlap was observed with the 16S
1017 rRNA gene clone libraries constructed for phylogenetic analysis of the bioreactor. In a similar
1018 work, Hanreich et al. (2013) analyzed the metaproteome of biogas batch fermentation systems
1019 co-digesting straw and hay. Members of the orders *Clostridiales*, *Bacteroidales* and
1020 *Flavobacteriales* were identified as key players in this ecosystem from the analysis of
1021 metagenomics data. A higher number of gene sequences codifying for enzymes from the
1022 glycoside hydrolase families GH5, GH48 and GH94 (comprising cellulase, cellobiohydrolyse
1023 and cellobiose phosphorylase activities) were present at the beginning of the incubations
1024 (together with higher numbers of *Clostridiales*-related microorganisms); over time, an increase
1025 in *Bacteroidales* and *Flavobacteriales* was observed with an increase in the number of
1026 sequences encoding enzymes of the GH2, GH13, GH29, GH77 and GH92 families (comprising
1027 enzymes capable of cleaving a variety of monosaccharides and polysaccharides). Nevertheless,
1028 only three hydrolytic enzymes could be detected during the proteomics analysis – a putative α -
1029 amylase, a pectaselyase and a glycosyde hydrolase family protein). On the other hand, a high
1030 number of enzymes involved in methanogenesis could be detected in the protein fractions (e.g.
1031 methyl-coenzyme M reductase), although methanogens represented a minor part of the
1032 community as inferred from metagenomic data (orders *Methanosarcinales*,
1033 *Methanomicrobiales*, and *Methanobacteriales*). Recently, Lü et al. (2014) performed a much
1034 more comprehensive metaproteomics study on thermophilic anaerobic cellulose degradation,
1035 with the identification of more than 500 non-redundant protein functions. The taxonomic
1036 distribution of the identified proteins suggested the presence of a few dominant bacterial groups,

1037 namely *Caldicellulosiruptor* species ($\approx 41\%$), *Coprothermobacter proteolyticus* ($\approx 20\%$) and
1038 *Clostridium thermocellum* ($\approx 16\%$). *Methanothermobacter thermoautotrophicus* ($\approx 5\%$) was the
1039 most abundant methanogen. A large number of proteins potentially involved in cellulose and
1040 hemicellulose binding and hydrolysis or in oligosaccharide metabolism were identified, and
1041 their synthesis linked to bacteria related to *Caldicellulosiruptor* spp. and *Clostridium*
1042 *thermocellum*. A cooperation between these two groups of bacteria was suggested, in which
1043 *Caldicellulosiruptor* spp. are responsible for hydrolysing the cellulosic part of the substrate,
1044 whereas *Clostridium thermocellum* is also hydrolysing hemicellulose in addition to cellulose.
1045 Regarding methanogenesis, data generated by pyrosequencing, metaproteomics and isotopic
1046 analyses together strongly support the predominance of the hydrogenotrophic pathway by
1047 strains of the genus *Methanothermobacter*. None of the enzymes specific for acetoclastic
1048 methanogenesis were identified, which opens the hypothesis of the occurrence of syntrophic
1049 acetate oxidation (SAO) in this ecosystem. Nevertheless, the lack of known protein specific for
1050 the SAO pathways and the limited knowledge about thermophilic SAO microorganisms did not
1051 allow further confirming this. SAO is currently a topic of interest as new microorganisms and
1052 theories for this pathway are emerging (Nobu et al. 2015).

1053

1054 **7- Statistical analysis to link results from molecular data to reactors performance**

1055 The results obtained by molecular methods can be linked to reactor performances using
1056 different statistical analysis. The link with environmental data is very important for the
1057 interpretation of the results. Although some conclusions can be drawn just by observing
1058 fingerprinting profiles or OTU tables, a deeper analysis is essential in order to arrive at strong
1059 and robust conclusions.

1060 There are different kinds of analyses that can be performed with different levels of complexity.
1061 The most frequently used are the hierarchical clustering analysis, the ordination methods and the
1062 calculation of diversity indices (Talbot et al. 2008). The methods can be applied to either
1063 fingerprinting data or sequencing analysis datasets. The first step is to perform a matrix with the
1064 samples and the presence/absence or abundance of each “species” (represented by OTUs,
1065 chromatogram peaks or DGGE bands). With this species matrix several analyses can be
1066 performed.

1067

1068 **7.1 Multivariable statistical tools**

1069 With the hierarchical cluster analysis the samples are grouped according to the dissimilarity
1070 between them, the results are visualized in a dendrogram constructed to represent how the
1071 samples are grouped and different coefficients and algorithms can be applied for the clustering

1072 (Ramette 2007). This analysis is very useful to determine the changes in the communities due to
1073 differences in operation conditions. The samples can also be ordinated according to the
1074 dissimilarities between them in two dimensional spaces using ordination methods as Principal
1075 Component Analysis (PCA), Non-metric Multidimensional Scaling (NMDS) or Canonical
1076 Correspondence Analysis (CCA). In constrained (canonical) ordination analysis methods it is
1077 possible to include in the original matrix additional environmental data such as: pH,
1078 temperature, organic loading rate (OLR), or other environmental data. The resulting ordination
1079 diagrams display samples, species, and environmental variables so that ‘fitted species samples’
1080 and ‘species environment’ relationships can be derived as easily as possible from angles
1081 between arrows or distances between points and arrows (Ramette 2007).

1082 Another objective may be to test whether differences between groups of objects (rows) in a
1083 multivariate table are significantly different based on the set of their attributes (columns), i.e. to
1084 test whether similarities within groups are higher than those between groups. Here,
1085 nonparametric multivariate ANOVA (NPMANOVA) and analysis of similarities (ANOSIM),
1086 which are commonly found in standard statistical packages, can be used (Ramette 2007).

1087 An overview of the analysis that can be applied in microbial ecology could be found in the
1088 review published by Ramette (2007) and in the web tool developed by Buttigieg and Ramette
1089 (2014).

1090

1091 **7.2 Diversity indices**

1092 Using the same matrix of species abundance, diversity indices can be calculated. The indices
1093 reflect the organization of the community and can be used to determine differences within
1094 microbial communities (Talbot et al. 2008). Three main diversity indices: richness, evenness (or
1095 equitability), and Shannon index, can be calculated from fingerprinting or sequencing data to
1096 define the structure of a microbial community (Hill et al. 2003). Richness is simply defined as
1097 the number of different OTUs obtained by the fingerprinting method or sequencing approach.
1098 Evenness is a measure of the equitability of abundance. In other words, an environmental
1099 sample with more even abundance is more diverse than a sample with predominant and sparse
1100 species. Finally, Shannon index (H_0) is calculated from the relative abundances of each
1101 band/peak/OTU and is also influenced by the number of different OTUs. There are other
1102 diversity indices described in the bibliography that could be also determined, more detailed
1103 information can be found at the work of Hill et al. (2003).

1104 The clustering, ordination methods and diversity indices can be calculated using different free
1105 softwares that are easily downloaded from the web (R package; <http://www.r-project.org/>,
1106 PAST package: http://palaeo-electronica.org/2001_1/past/issue1_01.htm, Hammer 2001).

1107

1108 **7.3 Microbial Resource Management concept**

1109 Marzoratti et al. (2008) proposed a new concept to obtain valuable information from
1110 fingerprinting data that can be applied for the management of microbial communities for
1111 biotechnological applications. The idea of the authors was to define parameters that describe the
1112 organization of a community and reflects its functionality. These parameters can be used to
1113 explain and predict the performance of reactors.

1114 Three parameters were proposed to describe the structure and assembly of the microbial
1115 communities: Richness (Rr), Functional organization (Fo), Dynamics (Dy).

1116 The Rr represents the number of different individuals (species) in a sample and is calculated as
1117 the number of peaks in a fingerprinting profile (T-RFLP or SSCP), the number of bands in
1118 DGGE or the number of different OTUs in a cloning library. In the case of DGGE the authors
1119 propose a rarified coefficient to give different weight according the identity of the
1120 microorganisms represented by a band. This coefficient was calculated according to the
1121 migration of the bands in the gel.

1122 The Fo coefficient is a single value that describes a specific degree of evenness, measuring the
1123 normalized area between a given Lorenz curve and the perfect evenness line. This coefficient
1124 can be determined by constructing Lorenz curves and calculating Gini coefficients (Rousseau et
1125 al. 1999) or by constructing Pareto-Lorenz curves (Wittebolle et al. 2008). The higher the Fo
1126 coefficient, the more uneven a community is.

1127 The Dy is a coefficient to determine the rate of change of a community along time and is
1128 calculated from the similarity values within two samples taken in a frame of time (for example
1129 if the samples are taken weekly the frame of time is one week). The similarity between two
1130 samples is determined from the cluster analysis.

1131 The authors propose that, according to the ecosystems functionality, the community will adapt
1132 to a particular assembly, represented by the three parameters that indicate if the community is in
1133 a “good or bad” organization to perform a function, deviation from these “normal” values may
1134 represent problems in the function (Marzoratti et al. 2008).

1135 These parameters were used to describe the communities from full scale anaerobic bioreactors
1136 (Pycke et al. 2009), to understand methanogenic bioreactors failures after overload (Carballa et
1137 al. 2009) and to correlate microbial communities with different operational conditions (Werner
1138 et al. 2011; Ciotola et al. 2013; Chen et al. 2014).

1139

1140 **8- Lessons from the application of molecular tools to study the anaerobic digestion process**
1141 **and future challenges**

1142 During the last three decades molecular biology tools have been more and more applied to the
1143 study of the anaerobic digestion process. Each method present different advantages and
1144 limitation which have to be taken into account (Table 3). Important questions have been
1145 addressed and important information was generated. The key point now is how to use this
1146 information to improve the process.

1147 From the 16S rRNA gene approach we now know that the microbial communities involved in
1148 the anaerobic digestion present a general microbial composition at the phylum level that is
1149 shared in most cases. These following phyla: *Chloroflexi*, *Firmicutes*, *Proteobacteria*,
1150 *Bacteroidetes*, *Synergistetes*, *Verrucomicrobia* predominate within the *Bacteria* Domain
1151 (Nelson et al. 2011; Werner et al. 2011; Sundberg et al. 2011). Species of known anaerobic
1152 microorganisms with important functions for the anaerobic digestion (hydrolysis, fermentation,
1153 acetogenesis) are included in these phyla. However, three of these phyla (*Chloroflexi*,
1154 *Verrucomicrobia* and *Synergistetes*) are represented by very few isolates and large amount of
1155 environmental sequences, then, there is a big hole in the knowledge of these organisms and their
1156 role in anaerobic digestion.

1157 In particular, the organisms from the *Chloroflexi* phylum presented high abundance in the
1158 biomass from methanogenic reactors, and form part of the “core group” of organisms always
1159 found in anaerobic digesters (Riviere et al. 2009; Werner et al. 2011). It was postulated that they
1160 play an important role in the hydrolytic and fermentative step and in the formation of granules
1161 or flocs (Sekiguchi et al. 2001; Yamada et al. 2005). They were also related to bulking problems
1162 in some UASB reactors (Yamada et al. 2007; Borzacconi et al. 2008). These organisms are very
1163 difficult to culture and isolate; therefore, new isolation procedures and new molecular
1164 techniques should be applied to know the role of these organisms so abundant in AD systems.

1165 On the other hand, it is known that the syntrophic organisms are a very specialized group of
1166 bacteria with low abundance in reactors and that they present high resilience (Werner et al.
1167 2011). Important efforts have been made to understand the ecology of these key
1168 microorganisms (Stams et al. 2012) and there is still much work to be done in the physiology of
1169 these organisms with such a particular way of obtaining energy.

1170 It is also known that *Archaea* from the *Euryarcheota* phylum are responsible for the
1171 methanogenic step. Within methanogens two metabolic pathways have been found: the
1172 hydrogenotrophic and acetoclastic. Although it was postulated for several years that the
1173 acetoclastic pathway predominate in the AD bioreactors (García et al. 2000; Batstone et al.
1174 2006), the molecular data now are questioning this paradigm as the hydrogenotrophic

1175 methanogens have been found to be more abundant in several investigations (Kampmann et al.
1176 2012). This is not trivial because the different groups of methanogens present different kinetics
1177 and different tolerances to toxic compounds. Hence, there is still a need of knowledge within
1178 this important group.

1179 Although the general composition of the biomass from AD process is, in general, maintained at
1180 the phylum level, the particular species that predominate in each system change. These changes
1181 during operation and between different systems demonstrate the high redundancy within the
1182 different guilds. It was postulated that this high redundancy is one of the keys to maintain the
1183 general function of the process in a changeable environment (Fernández et al. 1999). It was also
1184 demonstrated by several authors that the AD community is very dynamic (Fernández et al.
1185 1999, Zumstein et al. 2000; Pycke et al. 2011). This high dynamic was proven at the species
1186 level (represented by OTUs, fingerprinting peaks or bands), but it is not known whether this
1187 dynamic indicates change in the metabolic pathways or not. Future works using metagenomic
1188 and metaproteomics approaches are necessary to answer this important question.

1189 Several research groups have been focused in trying to explain the environmental factors that
1190 drive the AD process; from these works, it is possible to conclude that temperature of operation
1191 (Levén et al. 2007), type of substrate (Zhang et al. 2014) and inoculum are important factors
1192 that select different populations (Ali Shah et al. 2014). This is an important issue which should
1193 be taken into account for practical applications as they could aid in decrease start-up periods of
1194 full scale reactors and predict failures due to changes during operation. There is a need of
1195 compilation of all the information to have an overview of the different variables that affect the
1196 microbial communities in AD processes and how the communities cope with the changes during
1197 the operation. In that sense, a web site where all the information on the physiology of all the
1198 organisms present in the AD process, as was performed for the activated sludge process
1199 (<http://midasfieldguide.org>), would be a good initiative to follow.

1200 Finally, all the information generated should be applied to try to explain the function of reactors
1201 and help to solve engineering problems. Figure 7 gives an overview of the possible application
1202 of molecular biology tools to improve AD processes. The development of new molecular
1203 techniques focusing on the function, the application of deeper statistical analysis and the
1204 improvement of bioinformatic tools to manage the high amount of data generated will help to
1205 walk in that direction. An effort to integrate the four disciplines: microbial ecology,
1206 environmental engineering, mathematical modeling and bioinformatics will be necessary to
1207 achieve this goal.

1208

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1218

1219 **Conflict of interest**

1220 The authors declare no conflict of interest.

1221

1222 **References**

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Figure 1- Schematic view of the anaerobic degradation of organic matter

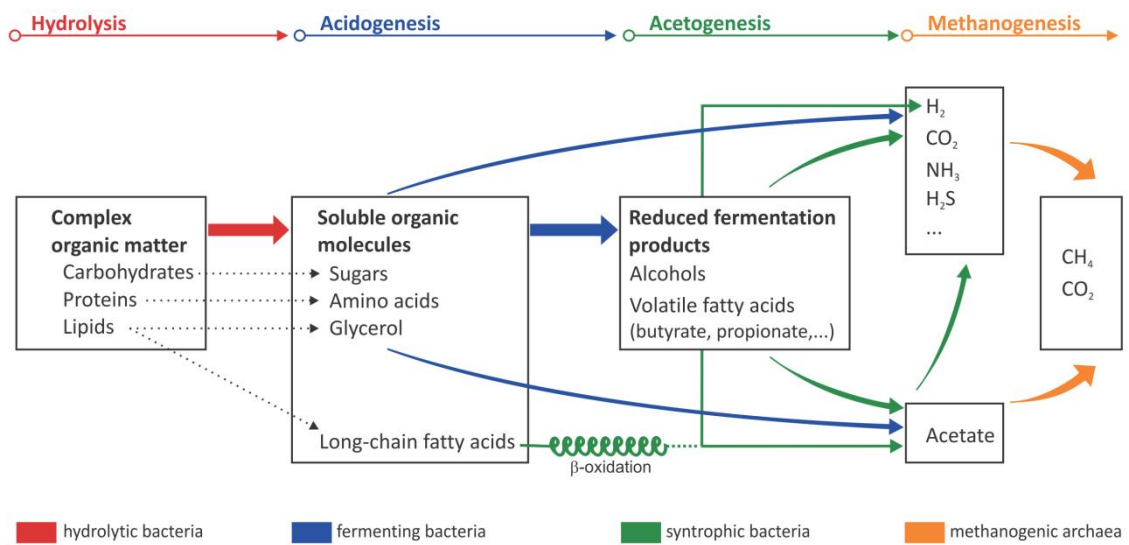


Figure 2- Questions to be formulated and molecular techniques that could be used to answer them, results obtained and statistical analysis.

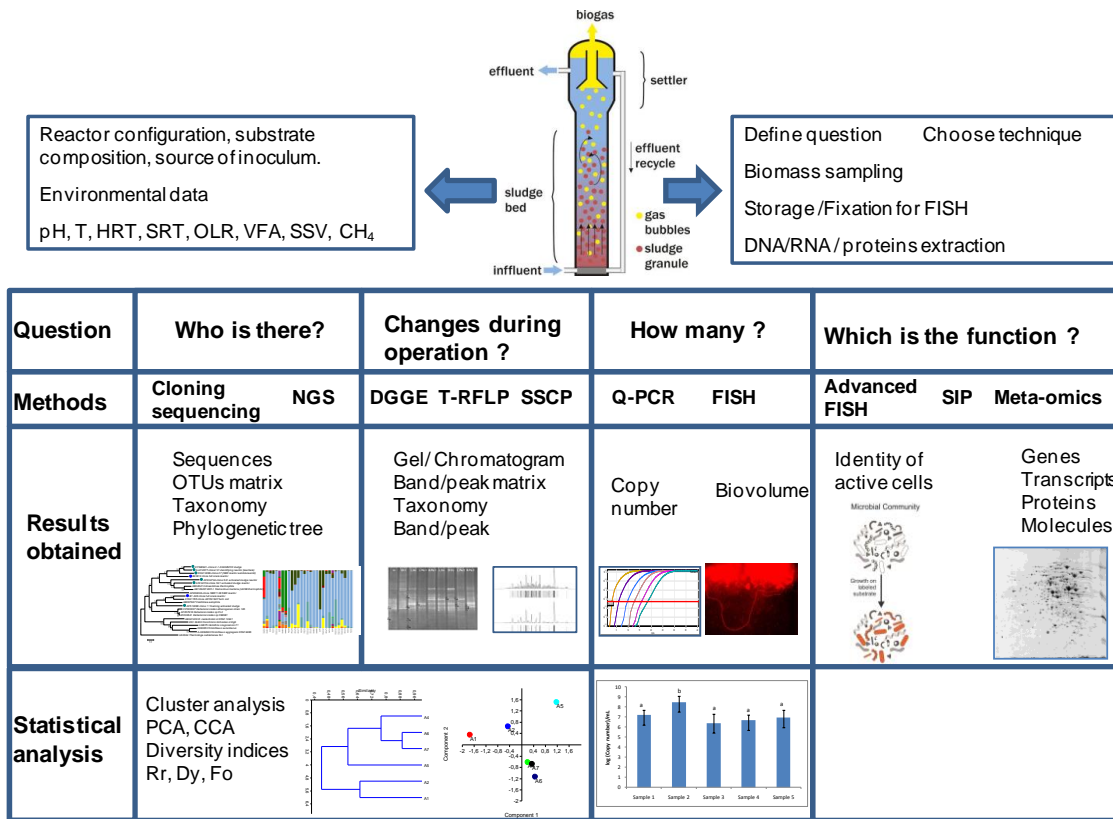


Figure 3 Methods to study microbial community structure and dynamics based on the amplification of 16S rRNA genes or functional genes: DGGE, T-RFLP and SSCP.

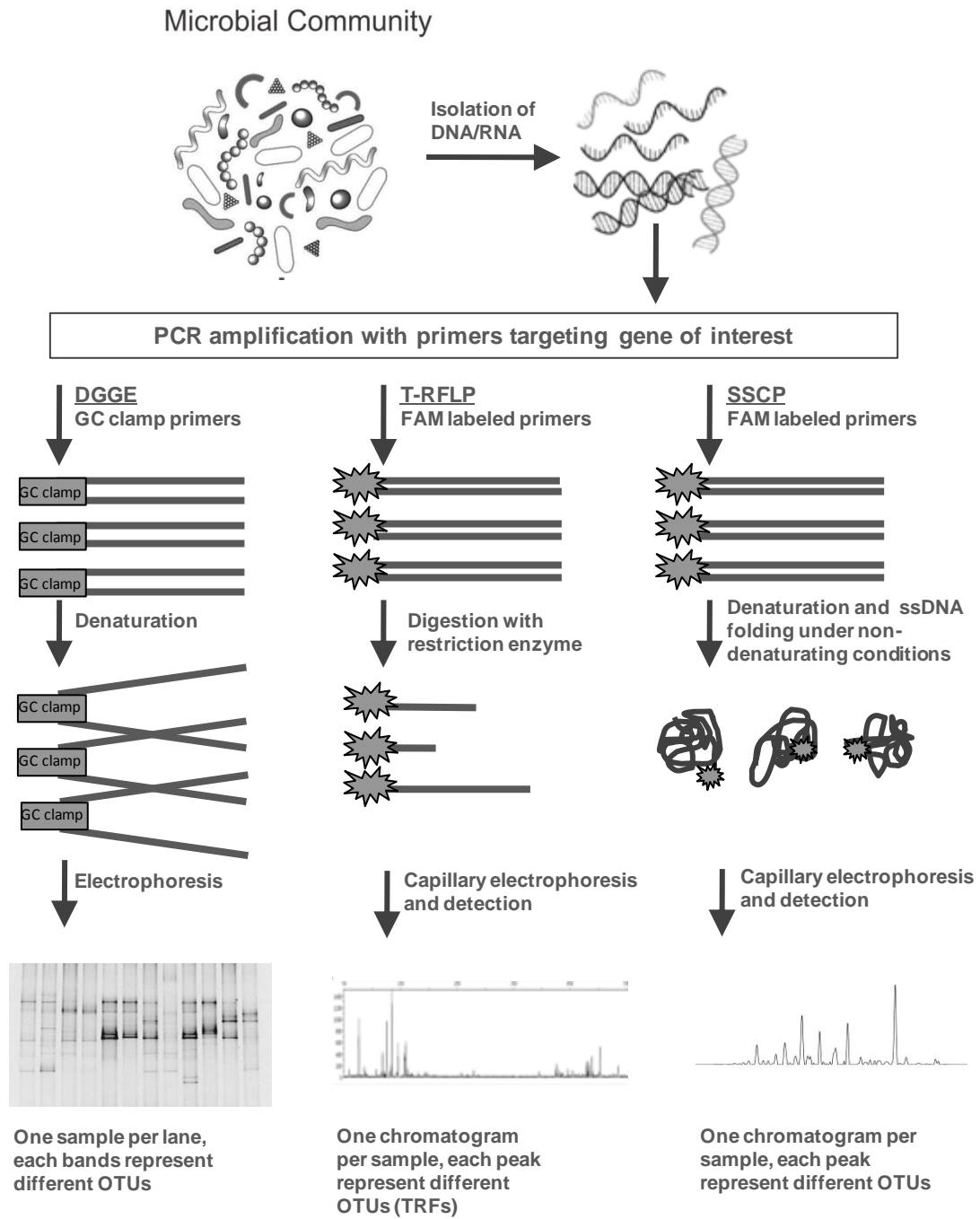


Figure 5. Possibilities for multi-omics analysis of complex microbial communities.

Metagenomics analysis provides information on taxonomy (based on retrieved 16S rRNA gene sequences) and on the metabolic genes (potential functional metabolic pathways) present in the metagenome. Metatranscriptomics enables the investigation of the actively transcribed ribosomal and messenger RNA from a community, giving more insight on the active fraction of the community; it can be used to study how communities and gene expression change in response to environmental changes. Metaproteomics is an excellent tool to for studying functionality in an ecosystem because proteins are the direct responsible for cell phenotype and are therefore more informative than the identification of functional genes or their corresponding messenger RNAs. Metabolomics has been recently introduced in systems biology approaches and studies the complete set of metabolites formed by the whole microbial community as a result of its interaction with biotic and abiotic factors of its environment.

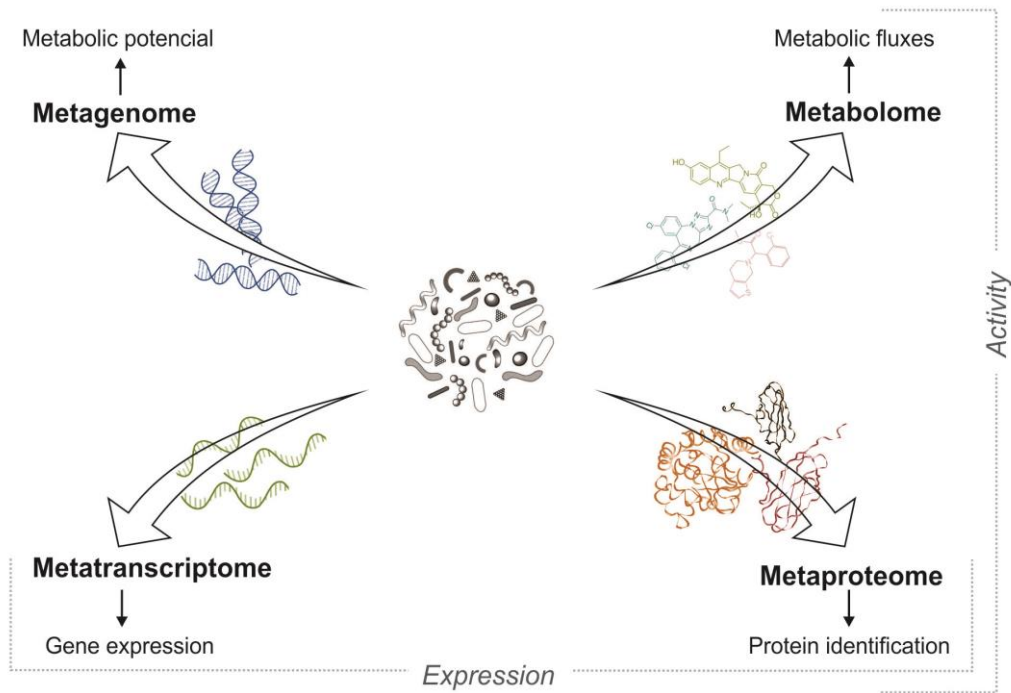


Figure 6. Hydrogenotrophic and acetoclastic methanogenesis pathways. Genes detected in methagenomic analyses of anaerobic digesters are indicated in blue - data from Rademacher et al.(2012) and Wong et al. (2013); in red, genes that could not be detected in those studies. Dotted lines represent an alternative pathway for the activation of acetate to acetyl-CoA, as suggested by Wong et al. (2013).

MF - methanofuran; MPT - methanopterin; CoM - coenzyme M; CoB - coenzyme B; CoA - coenzyme A; F₄₂₀ - coenzyme F₄₂₀; Fd - ferredoxin.

fmd - formylmethanofuran dehydrogenase; *ftr* - formylmethanofuran:H₄MPT formyltransferase; *mch* - methenyl-H₄MPT cyclohydrolase; *hmd* - H₂-forming methylene- H₄MPT dehydrogenase; *mer* - F₄₂₀-dependent methylene-H₄MPT reductase; *mtr* - methyl- H₄MPT:coenzyme M methyltransferase; *mcr* - methyl coenzyme M reductase; *ack* - acetate kinase; *pta* - phosphotransacetylase; *acs*- acetyl-CoA synthetase; *ppa*- inorganic pyrophosphatase; *cdh* - carbon monoxide dehydrogenase/acetyl-CoA synthase.

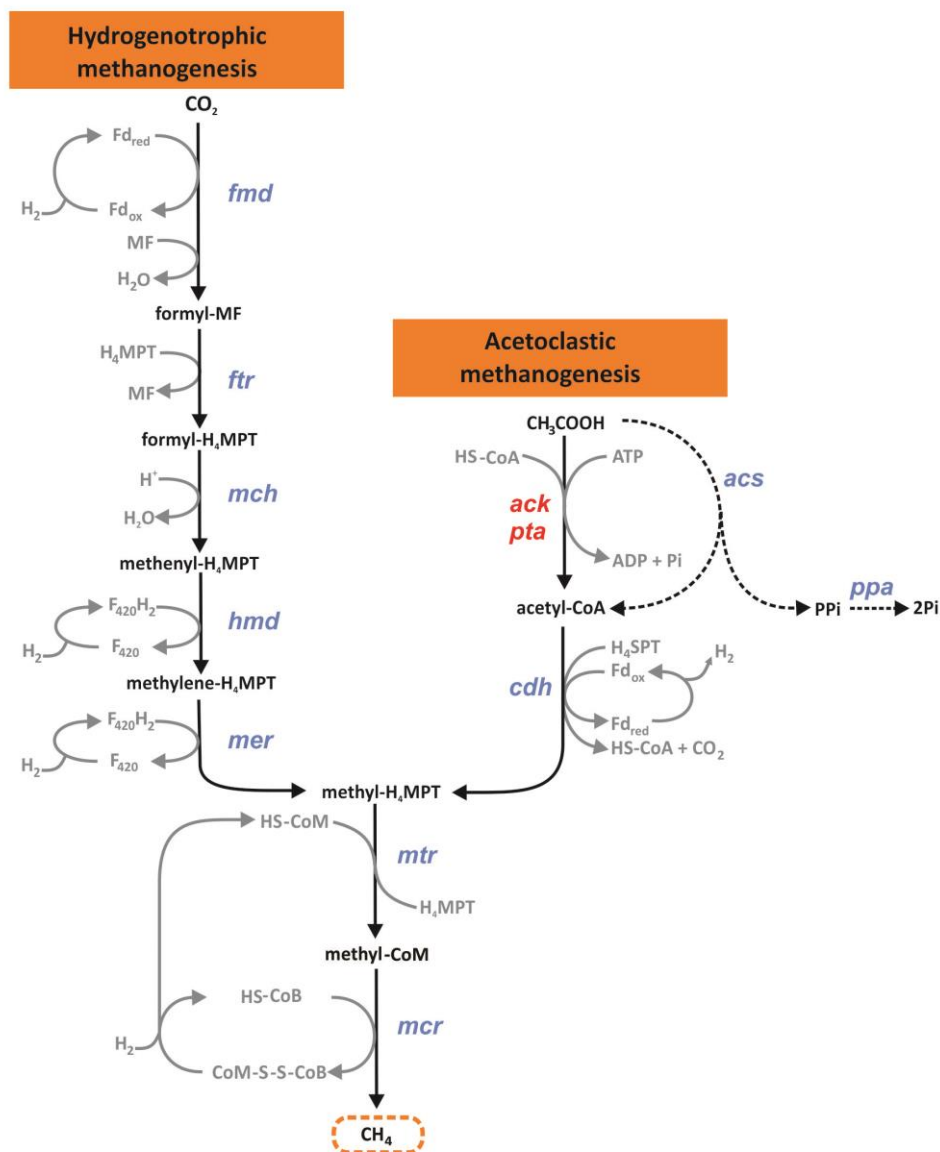
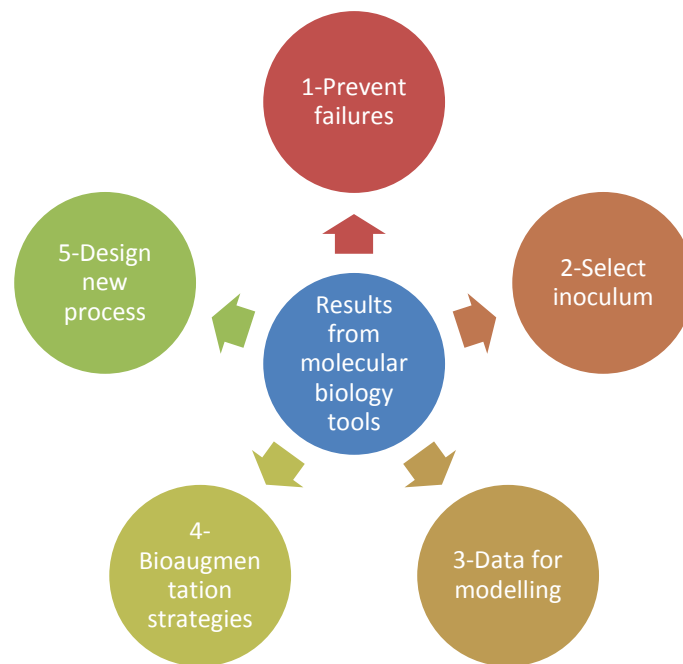


Figure 7- Possible applications of molecular biology techniques results to improve AD processes.

- 1- Prevent failures- By the regular monitoring of the microbial community of a bioreactor, information about the dynamic and the organization of the community during bad and good performance could be retrieved and used to predict the failures. The knowledge of microorganisms that cause problems (bulking, acidification, etc.) could be also used to prevents their growth..
- 2- Select inoculum- With the NGS available, it is possible to have a complete view of the microbial composition of a sample in one or two days with low cost this information could help to choose an adequate source of inoculum for full scale reactors avoiding cost of long periods of start up or reinoculation.
- 3- Data for modeling- Molecular data could be used to design systems of automatic control of reactors and for modeling. The AD model is based in data from kinetic and bulk measurements of the biomass. An effort of conceptualization of the data and integration of the two disciplines should be done.
- 4- Bioaugmentation- With the knowledge of key microorganisms important for the process it will be possible to design strategies for bioaugmentation and follow the growth of these organisms in the reactor. This could be very useful for the degradation of recalcitrant compounds (like detergents, pharmaceutical products) or to improve the degradation of difficult substrates (as lipids).
- 5- Design new process-With the discover of new metabolic functions of the microorganisms involved in the AD process it could be think different engineering process to solve some practical issues.



1 Table 1. Over view of the main bacterial phyla that composed the microbial communities from full scale anaerobic reactors. The values are mean of the values
 2 presented in the reports according to the results obtained from 16S rRNA gene sequencing analysis using different methods. Report 1: Nelson et al. (2011),
 3 Report 2: Riviere et al. (2009), Report 3: Sunberg et al. (2013).The data from unclassified sequences and some groups with low sequence abundance were not
 4 presented.

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Report	Method	Reactors analyzed	Total sequences within Bacteria	Bacterial Phylum (%)								
				<i>Chloroflexi</i>	<i>Proteobacteria</i>	<i>Firmicutes</i>	<i>Bacteroidetes</i>	<i>Synergistetes</i>	<i>Actinobacteria</i>	<i>Spirochaetes</i>	<i>Verrucomicrobia</i>	<i>Termotogales</i>
1	Sequences retrieved from database	NR*	16,519	23	22	15	15	6	2	1	1	1
2	Cloning/Sanger sequencing	7 UASB reactors treating municipal sewage sludge	7,000	32	18	9	11	6	2	1	1	1
3	454 pyrosequencing	21 CSTR reactors treating solid waste **	164,822									
		7 SS		1.7	5.7	25	15		4.7	4.6	0.9	1.6
		10 CD		0.1	0.4	69	14		1	0.1	0.1	0.3
		4 CDT		0.3	0.2	67	7		0.3	0.1	0.1	19

6 *NR= not reported.

7 ** The reactors studied were operated in different conditions: 7 reactors were fed with sewage sludge (SS), 10 reactors were fed with compose digestion (CD), and 4 reactors with compose
 8 digestion in thermophilic conditions (CDT).

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2 Table 2. Overview of main functional groups detected in anaerobic digestion sludges by using SIP, probing and advanced imaging techniques

Sample origin	Technique	Test substrate	Main bacterial groups detected*	Main archaeal groups detected*	Observations	Reference
Full-scale mesophilic anaerobic municipal sludge digester	MAR-FISH	¹⁴ C-glucose	Chloroflexi Spirochaetes <i>Spirochaeta</i> Candidate phylum TM7		β -Proteobacteria was the most dominant group in butyrate-, propionate- and acetate-degrading communities. MAR-positive β -Proteobacteria in each sample were collected by micromanipulation and their 16S rRNA gene cloned and sequenced. Half of the clones under the three conditions were affiliated with the genus <i>Variovorax</i> . <i>Smithella</i> and <i>Syntrophobacter</i> species play an important role in propionate degradation. Relative numbers of different populations within these genus vary with propionate concentration.	(Ariesyady et al. 2007a; Ariesyady et al. 2007b)
		¹⁴ C-butyrate	β -Proteobacteria Firmicutes <i>Syntrophomonas</i>			
		¹⁴ C-propionate	β -Proteobacteria δ -Proteobacteria <i>Smithella</i> <i>Syntrophobacter</i>			
		¹⁴ C-acetate	β -Proteobacteria γ -Proteobacteria Firmicutes Synergistetes <i>Synergistes</i>	Methanosarcinales <i>Methanosaeta</i> <i>Methanosarcina</i>		
Full-scale mesophilic anaerobic municipal solid waste digester	DNA- SIP	¹³ C-cellulose	Firmicutes <i>Clostridia</i> , <i>Acetivibrio</i> <i>Clostridia</i> , group 4	Methanomicrobiales <i>Methanoculleus</i>	<i>Acetivibrio</i> species were detected in aggregates around cellulose fibers using FISH, indicating that these bacteria are major cellulose degraders. Detection of <i>Methanoculleus</i> in the 'heavy' DNA may result from the assimilation of marked CO ₂ deriving from cellulose fermentation.	(Li et al. 2009)
		¹³ C-glucose	Firmicutes <i>Clostridia</i> , <i>Clostridium</i> Bacteroidetes <i>Porphyromonadaceae</i> <i>Rikenellaceae</i>		Hybridization with specific FISH probes confirmed abundant representation of <i>Clostridium</i> and <i>Porphyromonadaceae</i> (mainly in the planktonic phase).	
		¹³ C- acetate	γ - Proteobacteria <i>Pseudomonadaceae</i>	Methanomicrobiales <i>Methanoculleus</i>	FISH confirmed abundant representation of <i>Pseudomonadaceae</i> -related bacteria.	

					Syntrophic acetate oxidation between <i>Pseudomonadaceae</i> -related bacteria and <i>Methanoculleus</i> species was hypothesized.
Lab-scale mesophilic anaerobic fed-batch reactors fed with powdered whole milk	RNA-SIP MAR-FISH	¹³ C-glucose ¹⁴ C-glucose	Actinobacteria <i>Olsenella</i> <i>Propionibacterium</i> Bacteroidetes Chloroflexi Synergistetes <i>Synergistes</i> , group 4 δ-Proteobacteria <i>Syntrophobacter</i> <i>Smithella</i>	Methanosarcinales <i>Methanosaeta</i>	Actinobacteria, Bacteroidetes and Chloroflexi linked to glucose conversion to propionate and acetate. These labelled intermediates were further assimilated by propionate utilizers (<i>Syntrophobacter</i> and <i>Smithella</i>) and acetate utilizers (<i>Synergistes</i> group 4 and <i>Methanosaeta</i>). Actinobacteria was the most abundant MAR-FISH positive microbial group; increase of this group coincided with decreasing glucose concentration. <i>Smithella</i> has higher affinity to propionate than <i>Syntrophobacter</i> and, therefore, predominated when propionate concentrations were low.
Municipal solid waste digester	DNA-SIP SIMSISH	¹³ C-cellulose	Candidate division WWE1		Implication of candidate division WWE1 in anaerobic digestion of cellulose was confirmed at single-cell level by combined FISH-NanoSIMS analysis. Nevertheless, a major part of ¹³ C-enriched microbes did not exhibit positive hybridization signals with WWE1-specific probe indicating that different community members were using ¹³ C-cellulose.

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2 * For SIP analysis, main microbial groups detected in heavy fractions; for MAR-FISH, main MAR-positive groups are indicated

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1 Table 3. Advantages and limits of the molecular methods mostly used in anaerobic digestion; n/a: not appropriate.

Name	NGS	Cloning-sequencing	Microarrays	Fingerprint (DGGE, T-RFLP, SSCP)	qPCR	FISH, advanced FISH	SIP	Transcriptomic	Proteomic
Molecule targeted	Nucleic acids	Nucleic acids	Nucleic acids	Nucleic acids	Nucleic acids	Nucleic acids	Nucleic acids	RNA	Protein
PCR based	Yes	Yes	No	Yes	Yes	No	No	Yes	No
Physiological state of the cells	n/a	n/a	n/a	n/a	n/a	Alive or active	Active	Active	Active
Sampling and storage caution	-	-	-	-	-	Fixation	-	Rapid freezing	Rapid freezing
Specific procedure and equipment	Sequencer Bioinformatic data processing	Sequencer	Pre-designed array	Gel/capillary electrophoresis	Real-time PCR	Fluorescent/ confocal microscope	density gradient centrifugation	Sequencer Bioinformatic data processing	Mass spectro. Bioinformatic data processing
Technical easiness	+	+	-	+++	+++	++	+	+	+
Data processing requirements	+++	++	++	+	-	-	-	+++	+++
High throughput	+++	-	+	++	+	-	-	++	++
Phylogenetic identification	Yes	Yes	Yes	n/a	Yes	Yes	n/a *	Yes	Yes
Diversity measurement	Yes	Yes	Yes	Yes	n/a	n/a	n/a *	Yes	Yes
Spatial structuration	n/a	n/a	n/a	n/a	n/a	Yes	n/a*	n/a	n/a
Current use in AD	+++	+	-	+++	+++	+	+	+	+
Effort in technical development	+++	-	+	+	++	+	-	++	++

1 *Yes, if combined with sequencing (for phylogenetic identification and diversity measurement) or imaging (for spatial structuration) – *cf.* figure 4

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