

Housing conditions, level of feeding and presence of antibiotics in the feed shape rabbit cecal microbiota

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1 **Abstract**

2 **Background:** the effect of the production environment and different management practices
3 in rabbit cecal microbiota remains poorly understood. While previous studies have proved
4 the impact of the age or the feed composition, research in the housing conditions and other
5 animal management aspects, such as the presence of antibiotics in the feed or the level of
6 feeding, is still needed. Characterization of microbial diversity and composition of growing
7 rabbits raised under different conditions could help better understand the role these practices
8 play in cecal microbial communities and how it may result in different animal performance.

9 **Results:** four hundred twenty-five meat rabbits raised in two different facilities, fed under
10 two feeding regimes (*ad libitum* or restricted) with feed supplemented or free of antibiotics,
11 were selected for this study. A 16S rRNA gene-based assessment through the MiSeq Illumina
12 sequencing platform was performed on cecal samples collected from these individuals at
13 slaughter. Different univariate and multivariate approaches were conducted to unravel the
14 influence of the different factors on microbial alpha diversity and composition at phylum,
15 genus and OTU taxonomic levels. The animals raised in the facility harboring the most stable
16 environmental conditions had greater, and less variable, microbial richness and diversity.
17 Bootstrap univariate analyses of variance and sparse partial least squares-discriminant
18 analyses endorsed that the farm exerted the largest influence on rabbit microbiota since the
19 relative abundances of many taxa were found differentially represented between both
20 facilities at all taxonomic levels characterized. Furthermore, only five OTUs were needed to
21 achieve a perfect classification of samples according to the facility where animals were
22 raised. The level of feeding and the presence of antibiotics did not modify the global alpha
23 diversity but had an impact on some bacteria relative abundances, albeit in a small number

24 of taxa compared with the farm, which is consistent with the lower sample classification
25 power according to these factors achieved using microbial information.

26 **Conclusions:** this study reveals different degrees of influence attributable to environment
27 and animal management. It highlights the importance of offering a controlled breeding
28 environment that reduces differences in microbial cecal composition that could be causative
29 of different animal performance.

30

31 **Keywords**

32 cecal microbiota, meat rabbit, housing conditions, feed restriction, antibiotics, 16S MiSeq
33 Illumina sequencing, analysis of variance, multivariate approach

34

35 **Background**

36 Microbial communities that inhabit the gastrointestinal tract (GIT) of animals constitute a
37 complex ecosystem whose members constantly interact between them and with their host [1-
38 Gaskins, 1997]. These interactions ensure homeostatic balance maintenance since GIT
39 ecosystem components are involved in many physiological and immunological processes [2-
40 Belkaid and Hand, 2014]. In the case of the domestic meat rabbit (*Oryctolagus cuniculus*), a
41 small herbivorous mammalian belonging to the family *Leporidae*, cecum is the main organ
42 for microbial fermentation. Thus, it is not surprising that the rabbit cecum hosts the richest
43 and the most diverse microbial community of its GIT [3-Gouet and Fonty, 1979]. For this
44 reason, the cecum has been the organ preferably chosen in previous rabbit gut microbiota

45 assessments [4-Abecia et al., 2007; 5-Zou et al., 2016; 6-Zhu et al., 2017; 7-Chen et al.,
46 2019].

47

48 Thanks to the development of the next generation sequencing (NGS) technologies, and their
49 rapidly decreasing costs, it is currently possible to characterize the gut microbiota of a large
50 number of animals. This characterization allows a deeper comprehension of the differences
51 between animals concerning their microbial composition and diversity. It is hypothesized
52 that the production environment could partially mediate these differences. Our general aim
53 is to provide further evidence of the effect of different management and environmental
54 factors on the cecal microbial composition and diversity. In relation to this topic, there is a
55 certain amount of information already published. A growing number of studies have revealed
56 changes in rabbit cecal microbial communities exerted by the age [8-Combes et al., 2011] or
57 the type of feed provided to the kits after weaning [6-Zhu et al., 2017; 7-Chen et al., 2019].
58 Another factor that causes variation is the administration of antibiotics in the feed. Different
59 molecules have been widely administrated in rabbit meat production, especially after
60 weaning to curb mortality peaks (sometimes over 20%) as a result of the onset of
61 gastrointestinal symptoms [9-Gidenne et al., 2010]. Multiple studies have shown alterations
62 caused on gut microbiota by the administration of antibiotics in the feed [5-Zou et al., 2016;
63 10-Eshar & Weese, 2014]. Despite the European Union banned their use as growth promoters
64 in animal feeds since 2006 (EC 1831/2003), the administration of one type of antibiotic
65 molecule is still allowed to prevent or treat the emergence of potential infectious diseases on
66 farms. However, substantial efforts are being made towards searching for efficient
67 alternatives which allow a complete withdrawal of antibiotic in animal feeds. In this context,

68 the application of feed restriction during the growing period was proposed as an interesting
69 alternative to the use of antibiotics. Quantitative feed restriction is a widely applied
70 commercial practice that consists of reducing by a certain percentage the amount of feed the
71 animal would consume when the food is provided *ad libitum*. Gidenne et al. (2009) [11-
72 Gidenne et al., 2009] demonstrated that feed restriction, despite penalizing animal growth,
73 improves feed efficiency and reduces mortality due to enteric disorders. It is
74 hypothesized that these positive effects could be partially explained by changes in gut
75 microbial composition or activity originated by the application of feed restriction. However,
76 techniques used so far to study this possible association have not found evidence of it [11-
77 Gidenne et al., 2009].

78 This study, which comprises a large number of animals in an experimental design involving
79 different management and environmental factors, is intended to unravel changes in diversity
80 and composition of rabbit cecal microbial communities associated with these factors. It will
81 allow a better understanding of how the housing conditions associated with the farm where
82 the animal was raised, the presence of antibiotics in the feed, and feed restriction shape the
83 cecal microbiota of growing rabbits.

84

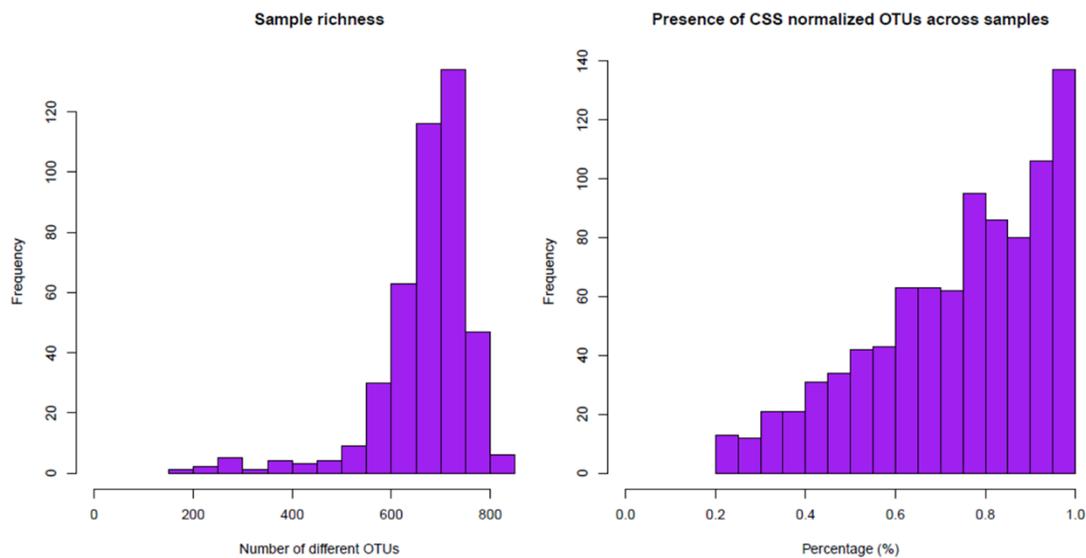
85 **Results**

86 **Sequence processing**

87 After the removal of doubletons and samples with low sequence counts, 425 rabbit cecal
88 samples (Additional file 1) were represented on 14,928,203 sequence counts clustered into
89 963 different OTUs. Each sample had on average 35,125 final sequences (range: 10,157-

90 678,798) and 677 OTUs (range: 197-841) (Additional files 2 and 3). Figure 1 shows two
91 histograms representing the sample richness and the proportion of OTUs present across
92 samples. Most of the samples had more than 700 different OTUs (mode = 748) and nearly
93 140 OTUs were present in all the samples.

94



95

96 **Figure 1 Sample richness and presence of CSS-normalized OTUs across samples.**

97

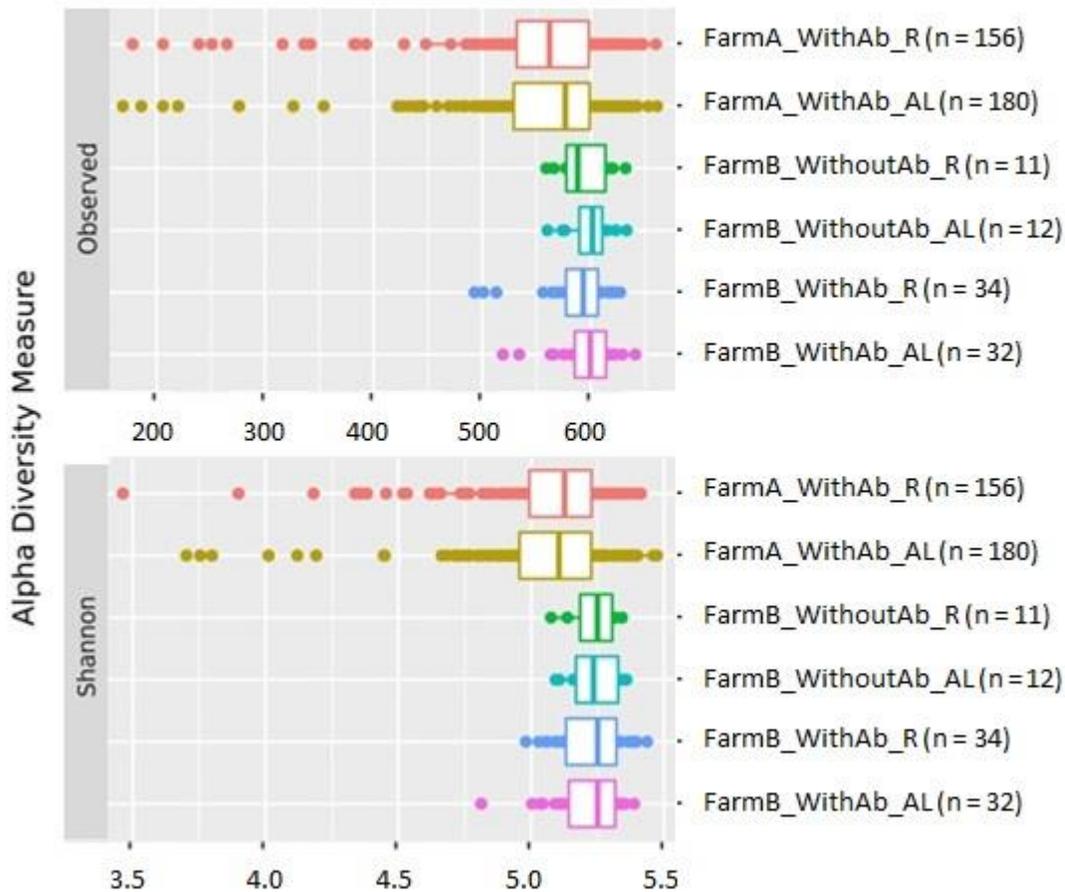
98 Taxonomic assignment of representative OTUs against the Greengenes reference database
99 gg_13_5_otus (Additional file 4) revealed the presence of 8 different known phyla with an
100 average of 8 phyla per sample (range: 7-8) (Additional file 5), and 28 different known genera
101 with an average of 24 genera per sample (range: 17-28) (Additional file 6).

102

103 **Animal management and farm environment shaping cecal microbial alpha diversity**

104 The study of alpha diversity was performed after rarefying the prefiltered and unnormalized
105 OTU table to 10,000 sequences per sample. Rarefaction generated a table which contained
106 the sequence counts of 963 different OTUs for 425 samples. The average (standard deviation)
107 number of observed OTUs within animal was 560.52 (75.03) and the average Shannon index
108 within animal was 5.09 (0.26). The comparison of alpha diversities revealed that the group
109 of animals raised in farm B had greater alpha diversity than the group of animals raised in
110 farm A (estimated differences of 40.20 (9.83) observed OTUs and 0.17 (0.03) Shannon index;
111 $P_{FDR} < 0.001$). Furthermore, larger variability in both indexes was observed in farm A than
112 in farm B. No significant differences for the two alpha diversity indexes were found between
113 feeding regimes within both farms (Figure 2, $P_{FDR} > 0.05$), nor between the presence and the
114 absence of antibiotics in the feed within the farm B (Figure 2, $P_{FDR} > 0.05$).

115



116

117 **Figure 2 Microbial richness and diversity between samples grouped according to**
 118 **management that animals received.** The cecal microbial richness and diversity were
 119 estimated by the observed number of different OTUs and the Shannon indexes, respectively.

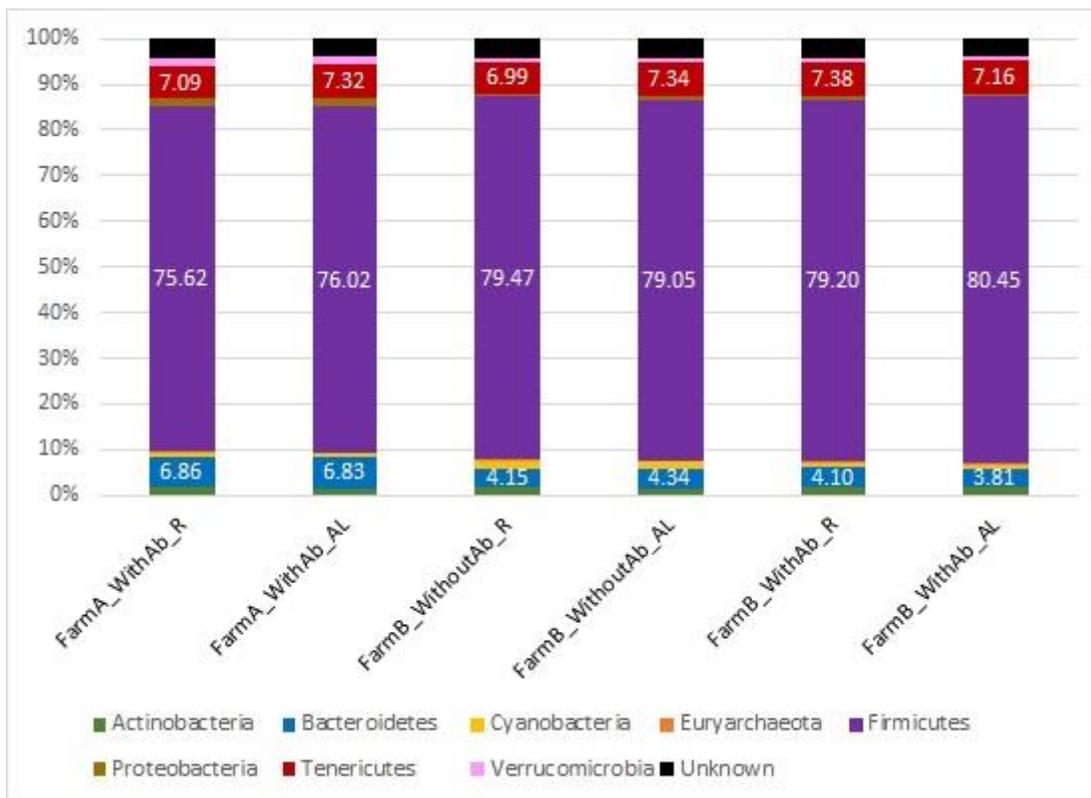
120

121 **Animal management and farm environment shaping cecal microbial composition**

122 According to the taxonomic assignment of representative sequences (Additional file 4)
 123 performed with the UCLUST consensus taxonomy assigner on the Greengenes reference
 124 database gg_13_5_otus, *Firmicutes* (76.74%), *Tenericutes* (7.22%) and *Bacteroidetes*

125 (6.26%) were the predominant phyla, accounting for more than 90% of the microbial
126 diversity, in the rabbit cecal samples studied (Figure 3).

127



128

129 **Figure 3 Phyla relative abundances of samples grouped according to farm, level of**
130 **feeding and presence of antibiotics in the feed.**

131

132 Differential growth and cecal microbial composition across farms

133 The facility where the animals were raised affected their growth performance. Animals raised
134 in farm B exhibited a faster growth (47.11 grams/day) than those raised in farm A (44.19
135 grams/day). The estimated average daily gain difference between farm B and farm A was

136 2.92 ± 0.94 grams per day ($P < 0.005$). Cecal samples of rabbits raised in farm A showed an
 137 overrepresentation of phyla *Bacteroidetes*, *Proteobacteria* and *Verrucomicrobia* while phyla
 138 *Euryarchaeota*, *Cyanobacteria* and *Firmicutes* were found to be overrepresented in cecal
 139 samples of rabbits raised in farm B (Table 1).

140

141 **Table 1 Microbial composition at phylum level in cecal samples of rabbits grouped by**
 142 **farm.**

| Phylum | Mean relative abundance in farm A (%) (SD) | Mean relative abundance in farm B (%) (SD) | Estimated difference farm A - farm B ± SE | P_{FDR} |
|------------------------|---|---|--|------------------------------------|
| <i>Actinobacteria</i> | 1.62 (0.67) | 1.84 (0.33) | -0.14 ± 0.08 | 0.09 |
| <i>Bacteroidetes</i> | 6.84 (1.81) | 4.03 (0.70) | 2.74 ± 0.22 | 0.00 |
| <i>Cyanobacteria</i> | 0.77 (0.40) | 1.05 (0.36) | -0.39 ± 0.05 | 0.00 |
| <i>Euryarchaeota</i> | 0.13 (0.19) | 0.44 (0.17) | -0.28 ± 0.02 | 0.00 |
| <i>Firmicutes</i> | 75.83 (3.34) | 79.66 (1.53) | -3.78 ± 0.41 | 0.00 |
| <i>Proteobacteria</i> | 1.83 (0.62) | 0.66 (0.12) | 1.14 ± 0.07 | 0.00 |
| <i>Tenericutes</i> | 7.21 (1.47) | 7.25 (0.93) | 0.00 ± 0.18 | 0.99 |
| <i>Verrucomicrobia</i> | 1.62 (0.45) | 0.91 (0.24) | 0.68 ± 0.05 | 0.00 |

143

144 Genera *Ruminococcus* (4.32%), *Blautia* (2.96%) and *Oscillospira* (2.37%) dominated the
 145 meat rabbit cecal microbiota. Most of the relative abundance differences at genus level were
 146 found differentially represented between animals raised in the different farms: genera
 147 *Bacteroides*, *Parabacteroides*, *Rikenella*, *Anaerofustis*, *Anaerostipes*, *Clostridium*,
 148 *Coprobacillus*, *Anaeroplasma* and *Akkermansia* were overrepresented in cecal samples of
 149 rabbits raised in farm A while genera *Adlercreutzia*, *Butyricimonas*, *Odoribacter*,
 150 *Methanobrevibacter*, *Blautia*, *Butyrivibrio*, *Coprococcus*, *Dehalobacterium*, *Dorea*,
 151 *Oscillospira*, *rc4-4* and *Oxalabacter* were overrepresented in cecal samples of rabbits raised

152 in farm B. Interestingly, genera *Epulopiscium*, *p-75-a5*, *Phascolarctobacterium*,
 153 *Campylobacter* and *Desulfovibrio* were only found in samples collected from farm A (Table
 154 2).

155

156 **Table 2 Relative abundances of genera, grouped by phylum, differentially represented**

157 **between farms ($P_{FDR} < 0.05$).**

| Genus | Mean relative abundance in farm A (%) (SD) | Mean relative abundance in farm B (%) (SD) | Estimated difference farm A - farm B \pm SE |
|------------------------------|---|---|---|
| <i>Actinobacteria</i> | | | |
| <i>Adlercreutzia</i> | 0.89 (0.47) | 1.14 (0.23) | -0.19 \pm 0.06 |
| <i>Bacteroidetes</i> | | | |
| <i>Bacteroides</i> | 1.88 (0.67) | 0.80 (0.35) | 1.10 \pm 0.08 |
| <i>Butyricimonas</i> | 0.16 (0.19) | 0.35 (0.17) | -0.19 \pm 0.02 |
| <i>Odoribacter</i> | 0.23 (0.21) | 0.44 (0.20) | -0.21 \pm 0.03 |
| <i>Parabacteroides</i> | 0.25 (0.18) | 0.07 (0.07) | 0.18 \pm 0.02 |
| <i>Rikenella</i> | 0.39 (0.24) | 0.18 (0.13) | 0.25 \pm 0.03 |
| <i>Euryarchaeota</i> | | | |
| <i>Methanobrevibacter</i> | 0.13 (0.19) | 0.44 (0.17) | -0.28 \pm 0.02 |
| <i>Firmicutes</i> | | | |
| <i>Anaerofustis</i> | 0.12 (0.08) | 0.08 (0.04) | 0.03 \pm 0.01 |
| <i>Anaerostipes</i> | 0.17 (0.08) | 0.12 (0.04) | 0.06 \pm 0.01 |
| <i>Blautia</i> | 2.86 (0.67) | 3.22 (0.46) | -0.36 \pm 0.08 |
| <i>Butyrivibrio</i> | 0.10 (0.07) | 0.13 (0.06) | -0.03 \pm 0.01 |
| <i>Clostridium</i> | 1.09 (0.26) | 0.87 (0.13) | 0.21 \pm 0.03 |
| <i>Coprobacillus</i> | 0.20 (0.27) | 0.14 (0.08) | 0.08 \pm 0.03 |
| <i>Coprococcus</i> | 1.96 (0.42) | 2.26 (0.29) | -0.28 \pm 0.05 |
| <i>Dehalobacterium</i> | 0.05 (0.08) | 0.18 (0.03) | -0.13 \pm 0.01 |
| <i>Dorea</i> | 0.46 (0.12) | 0.51 (0.09) | -0.05 \pm 0.02 |
| <i>Epulopiscium</i> | 0.14 (0.11) | 0.00 (0.00) | 0.15 \pm 0.01 |
| <i>Oscillospira</i> | 2.11 (0.53) | 2.85 (0.31) | -0.79 \pm 0.07 |
| <i>p-75-a5</i> | 0.13 (0.06) | 0.00 (0.00) | 0.13 \pm 0.01 |
| <i>Phascolarctobacterium</i> | 0.27 (0.24) | 0.00 (0.00) | 0.26 \pm 0.03 |
| <i>rc4-4</i> | 0.13 (0.06) | 0.23 (0.03) | -0.10 \pm 0.01 |
| <i>Proteobacteria</i> | | | |
| <i>Campylobacter</i> | 0.08 (0.08) | 0.00 (0.00) | 0.08 \pm 0.01 |

| | | | |
|-------------------------------|-------------|-------------|--------------|
| <i>Desulfovibrio</i> | 0.58 (0.22) | 0.00 (0.00) | 0.57 ± 0.03 |
| <i>Oxalabacter</i> | 0.10 (0.06) | 0.13 (0.03) | -0.03 ± 0.01 |
| <i>Tenericutes</i> | | | |
| <i>Anaeroplasma</i> | 0.23 (0.18) | 0.10 (0.09) | 0.12 ± 0.02 |
| <i>Verrucomicrobia</i> | | | |
| <i>Akkermansia</i> | 1.62 (0.45) | 0.91 (0.23) | 0.68 ± 0.05 |

158

159 The analyses on the CSS-normalized OTUs revealed that 648 out of the 946 OTUs showed
160 signatures significantly different between farms. Out of these, 276 were overrepresented in
161 farm A, while 372 were overrepresented in farm B. Table S1 shows the estimated difference
162 between farms for these OTUs, their sequences and their assignment at the lowest taxonomic
163 level. Only 9 of them could be assigned at species level and 129 were assigned to known
164 genera. These results showed remarkable coincidences with those obtained from the analyses
165 directly performed on the relative abundance of taxa at phylum and genera levels. An
166 example that illustrates this match is the overrepresentation of genus *Akkermansia* in farm
167 A. This genus is encompassed by phylum *Verrucomicrobia* that was also overrepresented in
168 rabbits raised in farm A, as well as 6 out of the 7 OTUs assigned to this phylum.

169

170 Differential growth and cecal microbial composition across feeding regimes

171 The feeding regime affected the rabbits' growth performance in both facilities. Animals fed
172 AL had a higher growth (48.74 and 55.77 grams/day in farms A and B, respectively) than
173 those fed R (38.95 and 38.65 grams/day in farms A and B, respectively). The estimated
174 average daily gain difference between AL and R groups was 9.79 ± 0.58 and 17.12 ± 1.08
175 grams per day in farms A and B, respectively ($P < 0.001$). An overrepresentation of phyla
176 *Cyanobacteria* (estimated difference R - AL = 0.11 ± 0.04 ; $P_{FDR} = 0.04$) and

177 *Verrucomicrobia* (estimated difference R - AL = 0.11 ± 0.05 ; $P_{FDR} = 0.04$) was found in cecal
178 samples of rabbits fed R and raised in farm A. On the other hand, phylum *Euryarchaeota*
179 was overrepresented in animals fed R and raised in farm B (estimated difference R - AL =
180 0.14 ± 0.04 ; $P_{FDR} < 0.001$). At genus level, the only significant contrast was observed for
181 *rc4-4* which resulted overrepresented in samples from animals fed AL in farm A (estimated
182 difference R - AL = -0.03 ± 0.01 ; $P_{FDR} < 0.001$) while in farm B none of the genera resulted
183 differentially represented ($P_{FDR} > 0.05$) between feeding regimes. The contrasts based on the
184 CSS-normalized OTUs revealed 51 and 9 OTUs differentially represented between feeding
185 regimes within farms A and B, respectively. Within the farm A, 32 OTUs were
186 overrepresented in cecal samples of rabbits that were fed AL and 19 OTUs in the samples
187 from rabbits fed R. Within the farm B, 7 OTUs were overrepresented in cecal samples of
188 rabbits that were fed AL and 2 OTUs were overrepresented in rabbits that were fed R. Table
189 S2 shows the estimated difference between feeding regime within farm of these OTUs, their
190 sequences and their assignment at the lowest taxonomic level. The analyses based on the
191 CSS-normalized OTUs within the farm A were in full accordance with the analyses
192 performed at genus level given that all OTUs assigned to genus *rc4-4* (phylum *Firmicutes*)
193 were overrepresented in cecal samples of rabbits fed AL.

194

195 Effect of the presence of antibiotics in the feed

196 The effect of the presence of antibiotics in the feed could only be assessed within the farm B
197 given that all rabbits raised in farm A received feed supplemented with antibiotics. Animals
198 that received antibiotics had a slightly higher growth (47.29 grams/day) than those that did
199 not received (46.59 grams/day). The estimated average daily gain difference between groups

200 was not significant (0.69 ± 2.43 grams per day; $P = 0.78$). Cecal samples of rabbits that
201 received feed free of antibiotics showed an overrepresentation of phyla *Cyanobacteria*
202 compared to those that received feed supplemented with antibiotics (estimated difference
203 without antibiotics - with antibiotics = 0.49 ± 0.09 ; $P_{FDR} < 0.001$). In addition, the analyses
204 on the CSS-normalized OTUs revealed an overrepresentation of 15 and 29 OTUs in cecal
205 samples of rabbits that received a feed supplemented or free of antibiotics; respectively.
206 Table S3 shows the estimated difference between the presence and the absence of antibiotics
207 in the feed for the OTUs in which the differences reached the significance threshold. The
208 OTU sequences as well as their assignment at the lowest taxonomic level are also shown in
209 Table S3. Only 1 of these OTUs could be assigned at species level (*Bacteroides fragilis*) and
210 2 OTUs at genus level (*Oscillospira* and *Coproccoccus*).

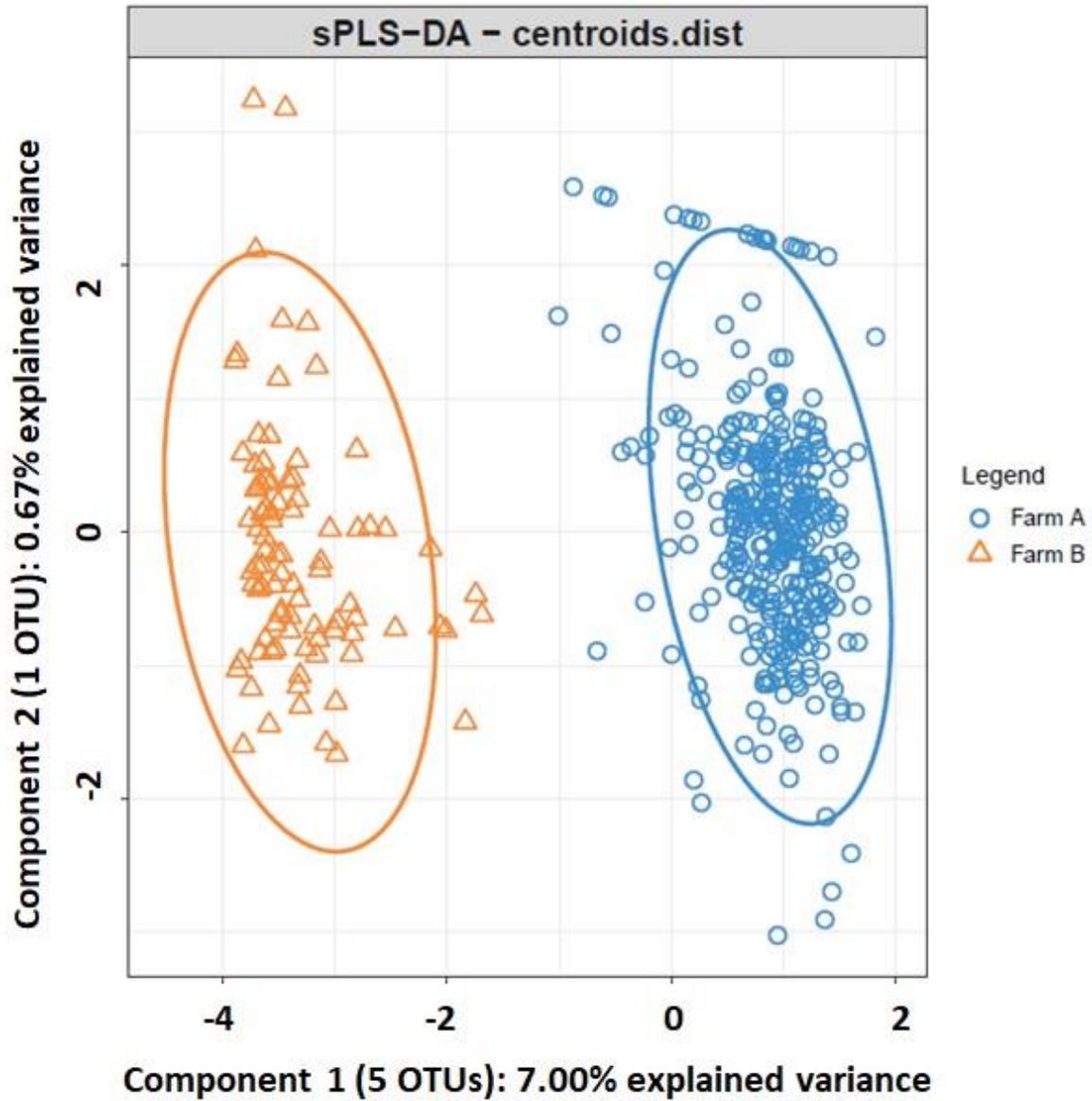
211

212 **Microbial information as a classifier of cecal samples according to farm environment** 213 **and animal management**

214 Sparse partial least squares-discriminant analyses (sPLS-DA) on the CSS-normalized OTUs
215 were conducted to discriminate samples according to the factors considered in this study (i.e.,
216 the farm where the animal was raised, the presence or the absence of antibiotics in the feed
217 and the feeding regime). The tuning process of the sPLS-DA conducted to discriminate
218 samples according to the farm where the rabbits were raised selected 5 OTUs for component
219 1 and 1 OTU for component 2 (Figure 4). Component 1 explained 7.00% of the total variance
220 while component 2 explained 0.67%. The classification performance of this sPLS-DA could
221 be said to be perfect since its overall and balanced error rate (BER) per class across 1000

222 replicates of 5-folds cross-validation runs was 0.00 (0.00). Furthermore, two OTUs of
223 component 1 had a stability higher than 0.9.

224



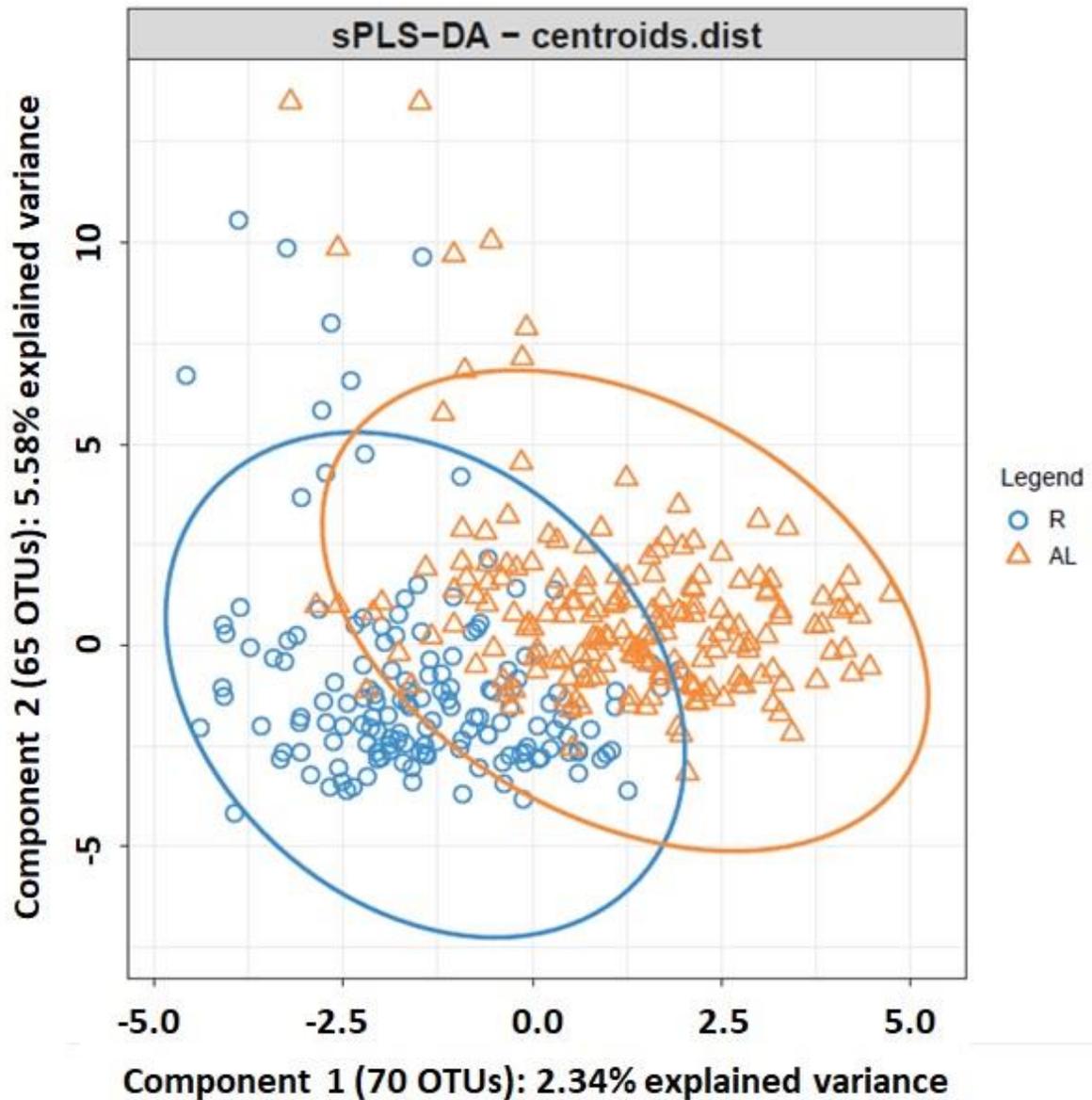
225

226 **Figure 4 Sparse partial least squares discriminant analysis representing cecal samples**
227 **of rabbits raised in farm A (blue) and in farm B (orange).**

228

229 The sPLS-DA performed to discriminate samples across feeding regimes within the farm A
230 selected 70 OTUs for component 1 and 65 OTUs for component 2 (Figure 5). Component 1
231 explained 2.34% of the total variance while component 2 explained 5.58%. The cross-
232 validation assessment of the classification performance of this sPLS-DA showed an overall
233 and BER per class of 0.27 (0.02). The stability of 18 and 5 OTUs selected in components 1
234 and 2, respectively, across the different cross-validation folds was higher than 0.9.

235



236

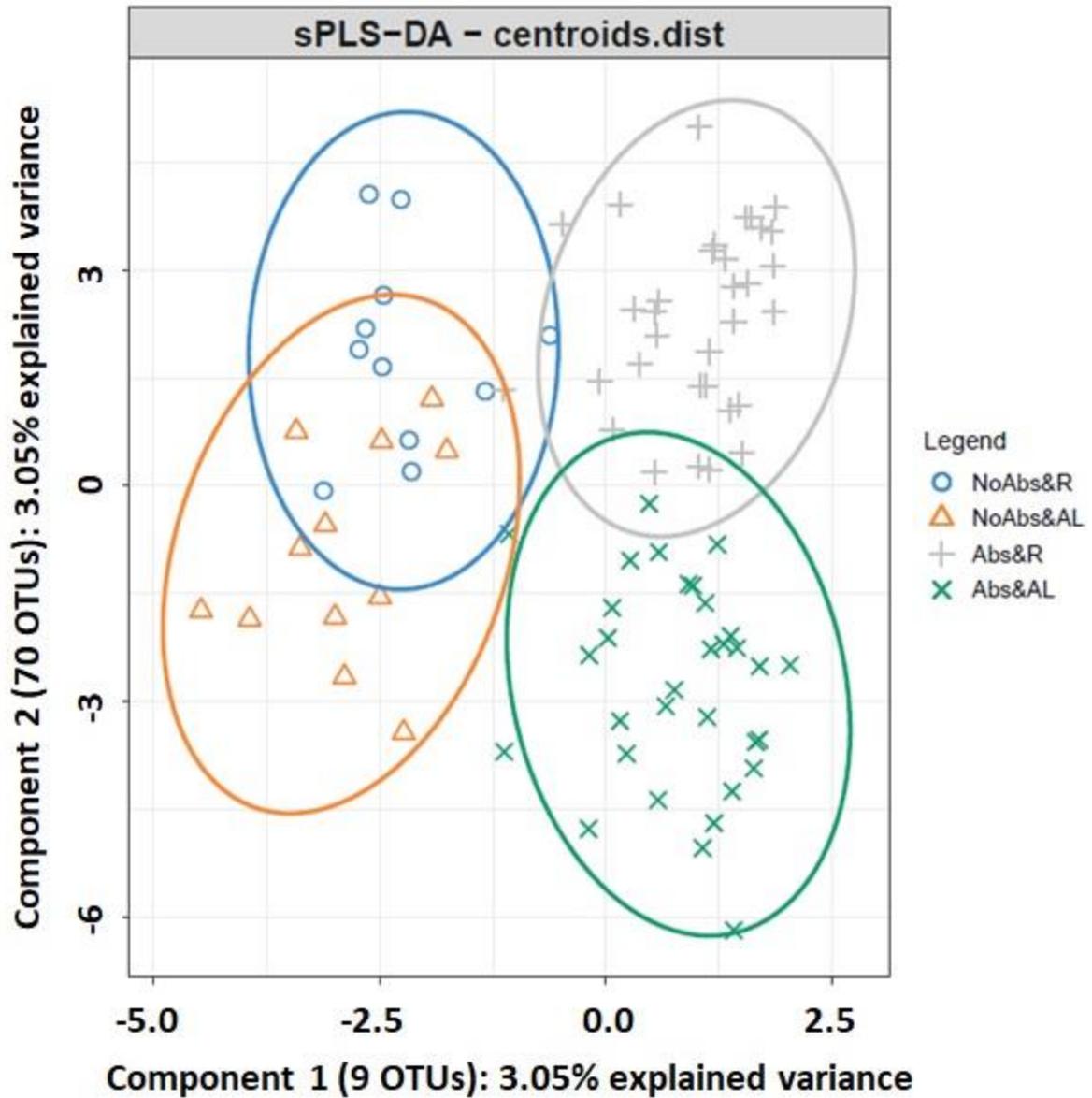
237 **Figure 5 Sparse partial least squares discriminant analysis representing cecal samples**
 238 **of rabbits raised in farm A and fed R (blue) or AL (orange).**

239

240 Finally, the sPLS-DA conducted to discriminate samples of animals raised within the farm
 241 B according to the combination of the presence or not of antibiotics in the feed and the
 242 feeding regime selected 9 OTUs for component 1 and 70 OTUs for component 2 (Figure 6).

243 Component 1 explained 3.05% of total variance and defined the discrimination between
244 samples from animals fed with antibiotics and those fed without antibiotics. On the other
245 hand, component 2 explained 3.05% of total variance and defined the discrimination between
246 samples from animals fed R and those belonging to animals fed AL. The cross-validation
247 assessment of the classification performance of this sPLS-DA showed an overall BER of
248 0.32 (0.15). The BER per class was 0.34 (0.12) for samples fed R without antibiotics, 0.46
249 (0.14) for samples fed AL without antibiotics, 0.29 (0.11) for samples fed R with antibiotics,
250 and 0.20 (0.07) for samples fed AL with antibiotics. The stability of 3 and 11 OTUs selected
251 in components 1 and 2, respectively, across the different cross-validation folds was higher
252 than 0.9.

253



254

255 **Figure 6 Sparse partial least squares discriminant analysis representing cecal samples**
 256 **of rabbits raised in farm B and fed R without antibiotics (blue), fed AL without**
 257 **antibiotics (orange), fed R with antibiotics (gray) and fed AL with antibiotics (green).**

258

259

260 **Discussion**

261 The influences of farm environment and common commercial practices of animal
262 management on their gut microbiota are not yet well known in many livestock species. In
263 this study, we have aimed to disentangle potential changes in microbial diversity and
264 composition of meat rabbit cecal communities as a result of being raised in different farms
265 and subjected to different handling during their growing period. To shed light on this matter,
266 we conducted a microbiota comparison of a large number of rabbits raised under different
267 housing conditions, feeding regimes, and fed with feed supplemented or free of antibiotics.

268

269 **16S rRNA gene-based characterization of meat rabbit cecal microbiota**

270 The Illumina MiSeq sequence processing of samples collected from these animals revealed
271 that phyla *Firmicutes*, *Tenericutes* and *Bacteroidetes* dominate the growing meat rabbit cecal
272 ecosystem representing more than 90% of its entire microbial composition. This fact is in
273 accordance with previous studies that have characterized the rabbit cecal microbiota [5-Zou
274 et al., 2016; 7-Chen et al., 2019; 12-Velasco-Galilea et al., 2018] and reported *Firmicutes* as
275 the predominant phylum. However, there are discrepancies between studies in establishing
276 which other phyla are also prevalent in this ecosystem. Whereas we found phyla *Tenericutes*
277 and *Bacteroidetes* representing 7.22% and 5.93% of the cecal microbial composition,
278 respectively, Chen et al. 2019 [7] and Zou et al. 2016 [5] reported *Bacteroidetes* as the second
279 predominant phylum representing 18% and 20% of New Zealand White and Rex rabbit cecal
280 microbial composition, respectively. Conversely, other studies that have previously
281 characterized meat rabbit fecal microbiota identified phyla *Proteobacteria* and

282 *Verrucomicrobia* in higher relative abundances [13-Kylie et al., 2018; 10-Eshar and Weese,
283 2014]. Velasco-Galilea et al. 2018 [12] reported *Firmicutes* (76.42%), *Tenericutes* (7.83%)
284 and *Bacteroidetes* (7.42%) as the predominant phyla of meat rabbit fecal and cecal microbial
285 communities. These discrepancies found across studies could be attributed to technical issues
286 (e.g., pair of primers, sequencing platform, bioinformatic pipeline employed to process raw
287 sequences or reference database used for the taxonomic assignment of the representative
288 sequences) or to purely biological reasons (e.g., breed, age or section of the GIT sampled).
289 Nonetheless, Kylie et al. (2018) [13] depicted that the relative increase in less beneficial
290 phyla, such as *Proteobacteria*, could be related to seasonal climate changes that directly
291 impact rabbits' health. This impact affects the susceptibility to enteritis and possibly feed
292 conversion efficiency. In any case, this phylum was more prevalent in farm A where the
293 animals were more exposed to changes in climate conditions.

294

295 **Farm environment modify alpha diversity**

296 Regarding the alpha diversity assessment, Shannon and the observed number of OTUs
297 indexes revealed the existence of significant differences between housing conditions (i.e., the
298 experimental farm where the rabbits were raised). Cecal samples collected from rabbits raised
299 in farm B had greater richness and diversity than those belonging to animals raised in farm
300 A. This could be explained by more stable environmental conditions in farm B (i.e., facility
301 better insulated) than in farm A. This combined with the fact that samples of animals raised
302 in farm A were collected from rabbits produced in 4 different batches, could also explain the
303 larger variability in both indexes observed in this farm [13-Kylie et al., 2018]. Despite not
304 having observed significant differences between the presence or not of antibiotic in the feed,

305 nor between feeding regimes, it is noteworthy to mention that samples collected from animals
306 fed AL in both farms had a greater, although not significant, richness than those fed R. This
307 fact is consistent with previous studies in mice that observed a lower alpha diversity in
308 animals with a restricted level of feeding [14-O'Neil et al., 2017; 15-Chen et al., 2016; 16-
309 Zarrinpar et al., 2014]. Surprisingly, but in agreement with our results, studies performed in
310 pigs [17-Soler et al., 2017], chicken [18-Kumar et al., 2018] and Rex rabbits [5] also did not
311 show clear significant differences on alpha diversity indexes between animals fed on diets
312 with antibiotics with respect to those on diets free of antibiotics. Nevertheless, these studies
313 were able to detect differences in the relative abundances of some specific species between
314 diets. For example, Kumar et al. 2018 [18] found that the inclusion of bacitracin in the feed
315 did not affect the chicken bacterial phyla. However, they observed differences between the
316 control and the bacitracin-fed group in the ileal and cecal bacterial populations at lower
317 taxonomic levels. It is worth noting that the antibiotic withdrawal at the beginning of the last
318 week of the rabbits' lives equalized the diets of both groups and possibly their microbial
319 populations, which may explain some lack of differences between them.

320

321 **Farm environment has a large impact on rabbit cecal microbiota**

322 Despite the lack of differences in microbial diversity and richness across management factors
323 (except for the farm); univariate studies revealed differential microbial composition across
324 the studied factors. In addition, the performed multivariate analysis evidenced a certain
325 classification power of the samples on the different levels of management and environment
326 factors based on the microbial composition of the samples.

327 As it might be expected, analyses of variance confirmed that the largest modification of meat
328 rabbit cecal microbial composition is generated by the housing conditions (in this case
329 represented by the farm factor). Our results revealed that the relative abundances of 6 out of
330 8 phyla are differentially represented between both farms. At genus level, we detected
331 significant differences in the relative abundances of almost all of them. Genera *Bacteroides*,
332 *Parabacteroides*, *Rikenella*, *Anaerofustis*, *Anaerostipes*, *Clostridium*, *Coprobacillus*,
333 *Anaeroplasma* and *Akkermansia* were enriched in cecal samples of rabbits housed in farm A.
334 The first three belong to phylum *Bacteroidetes* and genus *Bacteroides* is the most abundant
335 of them in meat rabbit cecum. Species of this genus are anaerobic Gram-negative members
336 of the family *Bacteroidaceae* that play an important role in the degradation of vegetal
337 polysaccharides and amino acid fermentation in the mammal GIT [19-Fang et al.,2017; 20-
338 Dai et al., 2011]. Moreover, this genus is involved in propionic acid and lactate formation
339 depending on nitrogen organic availability. Nonetheless, some authors showed that great
340 amounts of *Bacteroides* could predict obesity tendency. *Parabacteroides* is also an anaerobic
341 Gram-negative bacterium (family *Porphyromonadaceae*) involved in amino acid transport
342 and metabolism, energy production and conversion, lipid transport and metabolism,
343 recombination and repair, cell cycle control, cell division, and cell motility in the intestinal
344 microbiota of the growing rabbit [21-Sun et al., 2020]. This genus was specifically found in
345 the cecal microbiota of mice raised in conventional conditions and absent in those raised in
346 pathogen-free facilities in a study performed under different housing conditions [22- Müller
347 et al., 2016].

348 Within the phylum *Firmicutes*, genus *Clostridium* (family *Clostridiaceae*) is an anaerobic
349 Gram-positive bacterium that inhabits the GIT of many mammals where it acts by degrading

350 cellulose. However, some *Clostridium* species (e.g., *C. perfringens* and *C. difficile*) are
351 pathogenic, and an enrichment of this genus has previously been described in rabbits affected
352 by epizootic rabbit enteropathy [23- Bäuerl et al., 2014]. This genus, together with genus
353 *Bacteroides*, was found enriched in the cecal microbiota of mice housed in open cages
354 compared with those kept in individual ventilated cages [24-Thoene-Reineke et al., 2014].
355 Both genera have been associated with an exacerbation of the intestinal inflammatory
356 response in mammals [25-Terán-Ventura et al., 2010]. Genus *Anaerofustis* (family
357 *Eubacteriaceae*) has been found enriched in cecal samples of rabbits affected by
358 paratuberculosis infection (*Mycobacterium avium*) [26-Arrazuria et al., 2016].

359 Within the phylum *Verrucomicrobia*, genus *Akkermansia* is an anaerobic Gram-negative
360 bacterium that encompasses mucin degrader species [27-Belzer et al., 2012]. In the cecum, a
361 proper enrichment of this genus could maintain a suitable mucosal turn-over, thus exerting a
362 protective effect that could help the animal to deal with inflammatory processes.

363 It is worth mentioning that we have detected genera *Epulopiscium*, *p-75-a5*,
364 *Phascolarctobacterium*, *Campylobacter* and *Desulfovibrio* only in the cecal samples of
365 rabbits housed in farm A. The first three are encompassed within the phylum *Firmicutes*.
366 Genus *Epulopiscium* is a large size Gram-positive bacterium that has a nutritional symbiotic
367 relationship with surgeonfish that eats algae and detritus. This bacterium is physically similar
368 to the phylogenetically related *Metabacterium polyspora* which is an endospore-producing
369 bacterium isolated from the cecum of guinea pigs [28-Angert et al., 1996]. On the other hand,
370 genera *Campylobacter* and *Desulfovibrio* are Gram-negative bacteria that belong to phylum
371 *Proteobacteria*. Some species of these genera are pathogens responsible for infections and
372 diarrheas in mammals. The exclusive presence of these genera in farm A could indicate the

373 existence of a potential dysbiosis of the animals raised in that facility that could affect their
374 sanitary status and growth. While farm A was a semi-open-air facility, farm B was artificially
375 ventilated and offered more controlled environmental conditions that favor animal growth.
376 Moreover, the presence of sulfate-reducing bacteria (SRB) such as *Desulfovibrio* could be
377 enhanced by sulfate-secreting bacteria (SSB) such as *Rikenella* in farm A where this genus
378 is significantly more predominant. It is noteworthy to mention that SRB could also obtain
379 sulfate via *cross-feeding* mediated by *Bacteroides*-encoded sulfatases [29-Rey et al., 2013],
380 and interestingly, this phylum is more prevalent in farm A.

381 Regarding sample classification based on the sPLS-DA study, given the important
382 differences in gut microbial composition found between farms, a perfect classification of the
383 samples can be achieved with only 5 OTUs. One of these 5 OTUs was overrepresented in
384 farm B and belonged to family *S24-7* (phylum *Bacteroidetes*). The remaining 4 were
385 overrepresented in farm A and belonged to family *Barnesiellaceae* (phylum *Bacteroidetes*),
386 order *Bacteroidales* (phylum *Bacteroidetes*), and genera *Desulfovibrio* (phylum
387 *Proteobacteria*) and *Bacteroides* (phylum *Bacteroidetes*). It is worth mentioning that these 5
388 OTUs were also declared as differentially represented between farms by the univariate
389 analyses.

390

391 **Administration of antibiotics impact on some taxa relative abundances**

392 Within the farm B, the effect of the presence of antibiotics in the feed was assessed by
393 comparing the microbial cecal composition of rabbits fed with antibiotics with that of some
394 animals that received feed without antibiotics. As stated above, we did not detect significant

395 differences in alpha diversity, nor in genera relative abundances, between both groups.
396 However, some significant differences were observed at phylum and OTU levels. An
397 overrepresentation of phylum *Cyanobacteria* was found in rabbits fed without antibiotics.
398 The detection of this bacterial phylotype, commonly assigned to photosynthetic activity, in
399 the rabbit cecum could suggest contamination during the GIT sampling. However, Zeng et
400 al. 2015 [30] previously reported its presence in rabbit feces. In the present study, all OTUs
401 taxonomically assigned to phylum *Cyanobacteria* are as well encompassed in the order *YS2*.
402 Interestingly, it was demonstrated that this order does not really have photosynthetic capacity
403 and it is currently classified within the candidate phylum *Melainabacteria* [31-Di Rienzi et
404 al., 2013]. The non-photosynthetic cyanobacteria *YS2*, now named *Gastranaerophilales*, is a
405 fermenter gut-associated order present in humans and other animals such as squirrels, where
406 its exact role is unknown but it has the capacity to produce hydrogen, fix nitrogen and
407 synthesize vitamins B and K [31-Di Rienzi et al., 2013; 32-Monchamp et al., 2019; 33-Liu
408 et al., 2020]. Our results, in accordance with Kylie et al. 2018 [13], revealed that rabbits fed
409 without antibiotics exhibited higher abundances of OTUs assigned to phylum *Bacteroidetes*
410 than those fed with antibiotics. In addition, samples of rabbits that received antibiotics had a
411 significant increase of an OTU taxonomically assigned to genus *Coprococcus*. Interestingly,
412 a study that evaluated the differences in bacterial communities of Rex rabbits fed with
413 different antibiotics also found an overrepresentation of this bacterium in animals treated
414 with zinc bacitracin [5-Zou et al., 2016]. *Coprococcus* is an anaerobic bacterium that may
415 protect against colon cancer in humans by producing butyric acid [34-Ai et al., 2019]. We
416 hypothesized that the administration of antibiotics could modulate the abundance of some
417 *Coprococcus* species to provide intestinal protection on meat rabbits. However, it is
418 important to recognize that the reduced sample size of the group of rabbits fed without

419 antibiotics may have limited the statistical power to detect microbial composition differences
420 associated with this factor.

421

422 **Feed restriction modify *Euryarchaeota* and some bacteria relative abundances**

423 Within the farm B, the effect of the feeding regime in microbial composition was also
424 assessed by comparing samples of animals fed R with those fed AL. The main difference
425 found was for phylum *Euryarchaeota* which was overrepresented in animals fed R in farm
426 B. All *Euryarchaeota* species found in the rabbit cecum belong to genus *Methanobrevibacter*
427 that encompasses different hydrogenotrophic methane-producing species. Previous studies
428 in humans [35-Shen and Maitin, 2015] and cattle [36-McCabe et al., 2015; 37-McGovern et
429 al., 2017] found an overrepresentation of *Methanobrevibacter* species in individuals
430 submitted to feed restriction and a negative correlation between the abundance of this
431 bacterium and body mass index. A prevalence of *Methanobrevibacter* species could be a
432 positive indicator of a healthy microbiota since restricted animals showed an
433 overrepresentation of this genus. The main purpose of applying feed restriction is to improve
434 intestinal health, reducing weaning mortality. The growth of *Methanobrevibacter* is
435 supported by fermenters such as *Gastranaerophilales* and butyrate-producing bacteria such
436 as *Anaereostipes* via interspecies formate/hydrogen transfer [37-Bui et al., 2019]. A study in
437 mice determined that *Methanobrevibacter smithii* facilitates *Bacteroides thetaiotaomicron*
438 capacity to digest glycans resulting in increased production of short-chain fatty acids [38-
439 Samuel and Gordon, 2006]. The same study defined *M. smithii* as a “power broker” that
440 regulates polysaccharide fermentation efficiency that influences the fat stores. The lower
441 prevalence of methanogenic archaea in farm A could be explained by the high presence of

442 SRB that outcompetes with methanogens for hydrogen consumption. This fact could favor
443 hydrogen sulfide production and compromise the rabbits' health.

444

445 Regarding the sample classification based on the sPLS-DA study conducted within the farm
446 B, component 1 and component 2 discriminated between animals that did or did not received
447 antibiotics in the feed and between feeding regimes, respectively. It is worth mentioning that
448 8 out of 9 OTUs selected in component 1 were also declared as differentially represented
449 between the presence or the absence of antibiotics in the feed by the univariate analyses.
450 Within the farm A, an sPLS-DA was also performed to classify samples according to the
451 feeding regime using microbial information. Although a large number of OTUs were selected
452 as classifier variables in the tuning process of this sPLS-DA, the classification error rate was
453 high. It implied a poor discrimination capacity of samples according to the feeding regime
454 the animal received. Nevertheless, bootstrap univariate analyses of variance detected some
455 significant differences at all taxonomic levels analyzed between feeding regimes within the
456 farm A. At genus level, *rc4-4* was overrepresented in animals fed AL. This genus belongs to
457 phylum *Firmicutes* and it is known as an obesity-associated bacterium [39- Ziętak et al.,
458 2016] and as a pathogenic candidate identified in mice with multiple sclerosis [40-Gandy et
459 al., 2019]. A potential pro-inflammatory role has been proposed for this genus [40] what
460 could be related to a reduced incidence of enteric disorders when feed restriction is applied.
461 It is worth mentioning that family *Peptococcaceae*, which encompasses genus *rc4-4*, is
462 strongly related to total rabbit weight gain from weaning to 12-week old [41-North et al.,
463 2019]. Although in our study this genus was prevalent in animals fed AL, its association with

464 weight gain is not clear since the greater growth exhibited by these animals was consequence
465 of higher feed intake.

466

467 **Rabbit cecal microbiota is shaped by farm environment and animal management**

468 Different approaches have been applied in this study to evaluate the effect of different
469 environments and management practices, commonly used in rabbit production, in their cecal
470 microbial composition and diversity. Our results confirmed that the most important effect is
471 exerted by the environment provided by the farm where the animals were raised. Those raised
472 in the best insulated facility (farm B) appear to have a microbiota characteristic of healthier
473 animals than those raised in the open-air facility (farm A). It is worth mentioning that the
474 rabbits were housed in cages interspersed with feeding regime. This fact could make possible
475 the exchange of microorganisms between animals of different feeding regimes and therefore
476 have reduced the differences observed between regimes. However, the joint consideration of
477 70 OTUs in the sPLS-DA made possible a certain discrimination power of samples according
478 to the level of feeding received by each animal raised in farm A. It implies the existence of
479 cecal microbiota content patterns characteristic of each regime which could be revealed
480 thanks to the univariate analyses conducted at different taxonomic levels. Similarly, the
481 sPLS-DA performed within the farm B also involved the consideration of 70 OTUs to
482 discriminate samples according to the amount of feed consumed. Within this farm, the
483 classification of samples regarding the presence or the absence of antibiotics in the feed
484 needed a smaller number of OTUs than the feeding regime but greater than the farm. This
485 suggests that the effect of the presence of antibiotic in feed is stronger than the feeding level.
486 The implication of the discussed microbial composition and diversity differences originated

487 by the studied management and environmental factors on the animals' performance still
488 needs to be investigated. In future studies the role of specific groups of bacteria in rabbit
489 growth and feed efficiency will be analyzed.

490

491 **Conclusions**

492 The analysis of a large number of animals from a paternal rabbit line has allowed a deeper
493 comprehension of the role played by different management and environmental factors
494 shaping the composition and diversity of cecal microbial communities. It reveals that the
495 housing conditions offered to the rabbits during their growing play a key role that can result
496 in different microbial alpha diversity and composition of almost all species that inhabit the
497 rabbit GIT. This highlights the importance that a stable and controlled environment could
498 have in the intestinal health and, consequently, in animal performance. It seems clear that the
499 better insulated conditions of farm B favored the presence of a gut microbiota characteristic
500 of healthier animals. Although the level of feeding and the presence of antibiotics in the feed
501 did not modify the global diversity of cecal microbial communities, these factors can increase
502 or decrease the prevalence of specific bacteria which could lead to a microbial composition
503 potentially beneficial for the animal or, at the other extreme, to an origin of future intestinal
504 dysbiosis.

505

506 **Methods**

507 Animals and experimental design

508 All biological samples used in the study were collected from animals of an experiment
509 conducted at the Institute of Agrifood Research and Technology (IRTA) in different periods
510 and involving two different farms. The objective of that experiment was to estimate the effect
511 of the interaction between the genotype and the feeding regime (i.e., the amount of feed
512 provided during fattening) on growth, feed efficiency, carcass characteristics, and health
513 status of the animals [42-Piles and Sánchez, 2019]. For this particular study, 425 meat rabbits
514 from Caldes line [43-Gómez et al., 2002] of that experiment were randomly selected. Most
515 of them (336) were raised in 4 different batches in a semi-open-air facility (farm A). The
516 remaining animals (89) were produced in a single batch in another facility under better
517 controlled environmental conditions (farm B). Rabbits raised in farm A were housed in
518 collective cages containing 8 kits each one while those raised in farm B were housed in cages
519 with 6 kits each one. All animals were raised under the same management conditions and
520 received the same standard pelleted diet. Twenty-three rabbits raised in farm B received a
521 diet free of antibiotics and the remaining sixty-six received the same diet but supplemented
522 with antibiotics. Those raised in farm A received oxytetracycline, valnemulin, and colistin
523 while those in farm B received oxytetracycline, valnemulin and neomycin. At the time this
524 experiment was conducted, it was possible to use up to four types of molecules to prevent or
525 treat the emergence of potential infectious diseases on farms. However, nowadays, only one
526 antibiotic molecule is allowed. During the last fattening week all the animals received an
527 antibiotic free diet. Feed was supplied once per day in a feeder with three places for the 4-5
528 weeks that the fattening lasted. Water was provided *ad libitum* during the whole fattening
529 period. The animals were under two different feeding regimes: (1) *ad libitum* (AL) or (2)
530 restricted (R) to 75% of the AL feed intake. The amount of feed supplied to the animals under
531 R feeding regime in a given week for each batch was computed as 0.75 times the average

532 feed intake of kits on AL from the same batch during the previous week, plus 10% to account
533 for a feed intake increase as the animal grows. Kits were randomly assigned to one of these
534 two feeding regimes after weaning (32 days of age). They were categorized into two groups
535 according to their size at weaning (big if their body weight was greater than 700 g or small
536 otherwise) aiming to obtain homogenous groups regarding animal size within feeding
537 regime. A maximum of two kits of the same litter were assigned to the same cage in order to
538 remove the possible association between cage and maternal effects on animal growth during
539 the fattening period. The distribution of these animals across the different levels of
540 management factors is shown in Table 3. The body weight of each animal was weekly
541 recorded. The individual average daily gain was computed as the slope of the within animal
542 regression of all body weight measurements recorded during the growing period.

543 **Table 3 Distribution of rabbits in groups according to different management factors.**

| Farm | Batch | Feed | Feeding regime | Number of rabbits |
|-------------|--------------|---------------------|-----------------------|--------------------------|
| A | 1 | With antibiotics | Ad libitum | 27 |
| A | 1 | With antibiotics | Restricted | 30 |
| A | 2 | With antibiotics | <i>Ad libitum</i> | 35 |
| A | 2 | With antibiotics | Restricted | 41 |
| A | 3 | With antibiotics | <i>Ad libitum</i> | 61 |
| A | 3 | With antibiotics | Restricted | 53 |
| A | 4 | With antibiotics | <i>Ad libitum</i> | 57 |
| A | 4 | With antibiotics | Restricted | 32 |
| B | 5 | With antibiotics | <i>Ad libitum</i> | 32 |
| B | 5 | With antibiotics | Restricted | 34 |
| B | 5 | Without antibiotics | <i>Ad libitum</i> | 12 |
| B | 5 | Without antibiotics | Restricted | 11 |

544

545 Sample processing, DNA extraction and sequencing

546 Animals were slaughtered (at 66 and 60 days of age in farm A and farm B, respectively) and
547 cecal samples of each rabbit were collected in a sterile tube, kept cold in the laboratory (4°C)
548 and stored at -80°C. DNA extraction, amplification, Illumina library preparation and
549 sequencing followed methods described previously [12-Velasco-Galilea et al., 2018]. Whole
550 genomic DNA was extracted from 250 mg of each cecal samples using ZR Soil Microbe
551 DNA MiniPrep™ kit (ZymoResearch, Freiburg, Germany) according to manufacturer's
552 instructions with the following modification: cecal samples were mechanically lysed in a
553 FastPrep-24™ Homogenizer (MP Biomedicals, LLC, Santa Ana, CA, United States) at a
554 speed of 1 x 6 m/s for 60 s allowing an efficient lysis of archaea and bacteria species. Integrity
555 and purity of DNA extracts were measured with a Nanodrop ND-1000 spectrophotometer
556 equipment (NanoDrop products; Wilmington, DE, United States) according to Desjardins
557 and Conklin's protocol [44- Desjardins and Conklin, 2010]. All DNA extracts had adequate
558 integrity and purity (absorbance ratio 260 nm/280 nm > 1.6) to avoid PCR inhibition issues.

559

560 A fragment of the 16S rRNA gene including the V4-V5 hypervariable regions was amplified
561 with F515Y/R926 primer combination (5'-GTGYCAGCMGCCGCGGTAA-3', 5'-
562 CCGYCAATTYMTTTRAGTTT-3') [45-Parada et al., 2016] and then re-amplified in a
563 limited-cycle PCR reaction to add sequencing adaptors and 8 nucleotide dual-indexed
564 barcodes of multiplex Nextera® XT kit (Illumina, Inc., San Diego CA, United States)
565 following manufacturer's instructions. The initial PCR reactions were performed for each
566 sample using 12.5 µl 2x KAPA HiFi HotStart Ready Mix, 5 µl forward primer, 5 µl reverse
567 primer and 2.5 µl template DNA (5 ng/µl). The initial PCR conditions were as follows: initial

568 denaturation for 3 minutes at 95 °C, 25 cycles of 30 seconds at 95 °C, 30 seconds at 55 °C
569 and 30 seconds at 72 °C; and final extension for 2 minutes at 72 °C. The addition of indexes
570 and sequencing adaptors to both ends of the amplified regions took place in a second PCR
571 by using 25 µl 2x KAPA HiFi HotStart Ready Mix, 5 µl index i7, 5 µl index i5, 10 µl PCR
572 Grade water and 5 µl concentrated amplicons of initial PCR. The second PCR conditions
573 were as follows: initial denaturation for 3 minutes at 95 °C, 8 cycles of 30 seconds at 95 °C,
574 30 seconds at 55 °C and 30 seconds at 72 °C; and final extension for 5 minutes at 72 °C. Final
575 libraries were cleaned up with AMPure XP beads, validated by running 1 µl of a 1:50 dilution
576 on a Bioanalyzer DNA 1000 chip (Agilent Technologies, Inc., Santa Clara, CA, United
577 States) to verify their size, quantified by fluorometry with PicoGreen dsDNA quantification
578 kit (Invitrogen, Life Technologies, Carlsbad, CA, United States), pooled at equimolar
579 concentrations and paired-end sequenced in 5 parallel plates in a Illumina MiSeq 2 x 250
580 platform at the Genomics and Bioinformatics Service (SGB) of the Autonomous University
581 of Barcelona (UAB).

582

583 Bioinformatic pipeline for OTU calling

584 Sequence processing was performed using QIIME software (version 1.9.0) [46- Caporaso et
585 al., 2010]. In a first step, the resulting paired-ended V4-V5 16S rRNA gene reads were
586 assembled into contigs with the python script *multiple_join_paired_ends.py*. Then the
587 contigs were curated using the script *split_libraries.py* with default parameters in order to
588 assign them to samples and to discard those with a low-quality (Q19 was the minimum
589 acceptable quality score). Chimeric sequences generated during the process of DNA
590 amplification were detected with UCHIME algorithm [47- Edgar et al., 2011] and removed.

591 The totality of filtered contigs were clustered into operational taxonomic units (OTUs) with
592 a 97% similarity threshold using the script *pick_open_reference_otus.py* with default
593 parameters [48- Rideout et al., 2014] that grouped, through UCLUST algorithm [49- Edgar,
594 2010], the sequences against Greengenes reference database (version gg_13_5_otus) and also
595 made a *de novo* clustering of those that did not match the database. The generated OTU table
596 was filtered at: (1) sample level: by discarding samples with less than 5,000 final sequence
597 counts and at (2) OTU level: by removing the doubleton ones. The filtered OTU table
598 contained the sequence counts of 963 OTUs for 425 samples. Taxonomic assignment of
599 representative sequences of each OTU defined (963) was conducted by mapping them to the
600 Greengenes reference database gg_13_5_otus with the UCLUST consensus taxonomy
601 assigner (QIIME default parameters). The raw sequence data were deposited in the sequence
602 read archive of NCBI under the BioProject accession number PRJNA524130. Metadata, the
603 prefiltered and normalized OTU tables, and corresponding taxonomic classifications are also
604 included as Additional files 1, 2, 3 and 4, respectively.

605

606 Models and statistical methods

607 In order to study differences in diversity and richness between rabbits grouped according to
608 the farm environment and the management that they received, two alpha diversity indexes
609 (Shannon and the observed number of OTUs) were computed from the OTU table rarified to
610 10,000 sequences per sample with “phyloseq” R package [50-phyloseq]. The statistical
611 method chosen to assess alpha diversity differences between these groups of animals was an
612 analysis of variance that included a factor resulting from the combination of four factors (the

613 farm where the animal was raised, the batch, the presence or the absence of antibiotics in the
614 feed and the feeding regime). The significance threshold was set at 0.05 for type I error.

615

616 Different approaches were considered to assess the influence of the environments and
617 management factors on microbial composition. A bootstrap analysis of variance was
618 individually implemented for each OTU to test whether it was differentially represented
619 between the different categories of the factors studied. This univariate analysis was
620 conducted by normalizing the OTU table with the cumulative sum scaling (CSS) method [51-
621 Paulson et al., 2013] and only for those OTUs which were detected in at least 5% of the
622 samples and had a sum of its counts resulting in a frequency greater than 0.01% of the total
623 sum of all OTUs counts across all samples. It was implemented by fitting a model defined
624 by the combination of the four aforementioned factors by using *lm()* function in R [52- R].
625 Then, the differences between the CSS-normalized OTUs counts in the different levels of the
626 studied factors were tested. The significance between the levels of the main factors: farm,
627 presence of antibiotics in the feed and feeding regime was assessed using an F statistic. When
628 the involved interaction terms were significant, the contrasts of interest were studied nested
629 within the levels of other interacting factors, i.e. feeding regime were studied within farm
630 levels. When the interaction terms were not significant, the effects of the different levels were
631 averaged, i.e. the effects of the levels of the batches within the farm A were averaged to
632 present the effect associated with this farm. In the performed F tests instead of relying on the
633 theoretical distribution of the statistic under the null hypothesis to define the p-values, they
634 were empirically computed using bootstrap after 1,000 permutations of the dependent
635 variable with respect to the design matrix of factors in the model. The use of bootstrapping

636 enabled the hypothesis test to be done without the need of assuming that data are normally
637 distributed, which is an assumption that fails for OTUs counts. *P*-value was defined as the
638 proportion of bootstrap rounds having an F statistic value equal or greater than that obtained
639 with the original dataset. *P*-values were corrected defining a false discovery rate (FDR) of
640 0.05 [53- Benjamini and Hochberg, 1995]. This bootstrap analysis of variance approach was
641 also implemented to study the effect of the management factors on the relative abundance of
642 bacteria at phylum and genus levels.

643

644 The value of the microbial information to classify samples into the three factors considered
645 in our study was explored using multivariate techniques. In particular, sparse partial least
646 squares-discriminant analysis (sPLS-DA) [54-Le Cao et al., 2008] was used to find the
647 combination of OTUs that allowed the best classification of cecal samples according to: (1)
648 the farm where the animals were raised, (2) the feeding regime within the farm A and (3) the
649 combination of feeding regime and the presence or absence of antibiotics in the feed for the
650 animals raised in farm B. This approach was implemented through the R package
651 “mixOmics” [55-mixomics]. In a first step, the function *tune.splsda()* was used to select the
652 optimal sparsity parameters of the sPLS-DA model: the number of components and the
653 number of variables (OTUs) per component. For the tuning process, a 5-fold cross-validation
654 repeated 10 times was performed one component at a time, with a maximum of 4
655 components, on an input grid of values that indicate the number of variables to select on each
656 component. The sparsity parameters were defined, based on the BER and centroids distance,
657 and then included in the final sPLS-DA model. Samples were represented on the first two
658 components and colored according to their class (e.g., R or AL in the case of the feeding

659 regime) in a sample plot with the function *plotIndiv()*. The performance of the sPLS-DA
660 model was assessed with a 5-fold cross-validation repeated 1,000 times that randomly split
661 the data in training and validation sets. In this data partition, it was ensured that 20% of the
662 samples within each level of the discriminant factor were assigned to the validation set. Five
663 different partitions were performed for each replicate to guarantee a different sample
664 distribution in each validation set. The sPLS-DA model with the sparsity parameters
665 previously defined was adjusted in the training set and its classification performance was
666 assessed in the validation set using the overall and BER per class as criteria. The stability of
667 the OTUs selected on each component was also assessed in the cross-validation by computing
668 the selection frequency of each variable across the replicates.

669

670 **Additional files**

671 **Additional file 1:** metadata.txt. Metadata associated with the 425 rabbit cecal samples
672 analyzed in this study.

673 **Additional file 2:** otu_table_prefiltered_unnormalized.txt. Prefiltered and unnormalized
674 OTU table used for statistical analyses in this study.

675 **Additional file 3:** otu_table_filtered_CSSnormalized.txt. Filtered and CSS-normalized OTU
676 table used for statistical analyses in this study.

677 **Additional file 4:** rep_OTUs_tax_assignments.txt. Taxonomic assignments for all OTUs in
678 Additional file 2.

679 **Additional file 5:** phyla_table.txt. Relative abundances phyla table built from the collapse
680 of the filtered and CSS-normalized OTU table at phylum level.

681 **Additional file 6:** genera_table.txt. Relative abundances genera table built from the collapse
682 of the filtered and CSS-normalized OTU table at genus level.

683 **Additional file 7: Table S1.** OTUs differentially represented between farms.

684 **Additional file 8: Table S2.** OTUs differentially represented between feeding regimes
685 within farms.

686 **Additional file 9: Table S3.** OTUs differentially represented between the presence and the
687 absence of antibiotics in the feed within the farm B.

688

689 **List of abbreviations**

690 **AL:** *ad libitum*

691 **BER:** balanced error rate

692 **CSS:** cumulative sum scaling

693 **FDR:** false discovery rate

694 **GIT:** gastrointestinal tract

695 **NGS:** next generation sequencing

696 **OTU:** operational taxonomic unit

697 **PCR:** polymerase chain reaction

698 **R:** restricted

699 **sPLS-DA:** sparse partial least squares-discriminant analysis

700 **SRB:** sulfate-reducing bacteria (SRB)

701 **SSB:** sulfate-secreting bacteria (SRB)

702

703 **Declarations**

704 *Ethics approval and consent to participate*

705 This study was carried out in accordance with the recommendations of the animal care and
706 use committee of the Institute for Food and Agriculture Research and Technology

707 (IRTA). The protocol was approved by the committee of the Institute for Food and
708 Agriculture Research and Technology (IRTA).

709

710 *Consent for publication*

711 Not applicable.

712

713 *Availability of data and materials*

714 The raw sequence data were deposited in the sequence read archive of NCBI under the
715 accession number SRP186982 (BioProject PRJNA524130). Metadata, the prefiltered and
716 unnormalized OTU table, the filtered and CSS-normalized OTU table and corresponding

717 taxonomic assignments have all been included as Additional files 1, 2, 3 and 4, respectively.
718 Relative abundances phyla and genera table have also been included as Additional files 5 and
719 6, respectively. OTUs differentially represented between the studied factors, their sequences
720 and their assignment at the lowest taxonomic level have been included as Additional files 7,
721 8 and 9.

722

723 *Competing interests*

724 The authors declare that they have no competing interests.

725

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733

734 *Authors' contributions*

735 JS, MP and OR conceived the experimental design. JS, OR, MP and MVG collected
736 biological samples. MVG, OGR, MP, MG and AS processed the samples in the laboratory.
737 MVG processed and analyzed the sequencing data, interpreted data, prepared figures and

738 tables, and wrote the manuscript. JS and YRC helped analyzing the sequencing data. JS, MG,
739 MP, MV and YRC helped interpreting the data, and wrote and revised the manuscript. All
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741

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747

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