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Metataxonomic comparison between internal transcribed spacer and 26S ribosomal large subunit (LSU) rDNA gene

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19 Highlights

| 20 | - | Primer selection plays critical roles in the sensitivity and tracking fungal communities to |
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| 21 | | assess reliable and accurate ecological populations |
| 22 | - | The 26S target region exploited in rRNA sequencing demonstrated greater taxonomical |
| 23 | | depth for fungal communities |
| 24 | - | Preferential amplification phenomenon contributes to underestimations and overestimations |
| 25 | | of fungal species |

26 - The limited availability of updated databases to assess ecological populations.

28 Abstract

Next-generation sequencing has been used to strengthen knowledge about taxonomic diversity and 29 ecology of fungi within food ecosystems. However, primer amplification and identification bias 30 31 could edge our understanding into the fungal ecology. The aim of this study is to compare the performance of two primer pairs over two nuclear ribosomal RNA (rRNA) regions of the fungal 32 kingdom, namely the ITS2 and 26S regions. Fermented cocoa beans were employed as biological 33 material and the fungal ecology during fermentation was studied using amplicon-based sequencing 34 tools, making use of a manually curated 26S database constructed in this study, and validated with 35 SILVA database. To explore potential biases introduced by PCR amplification of fungal 36 communities, a mock community of known composition was prepared and tested. The relative 37 abundances observed for ITS2 suggest that species with longer amplification fragments are 38 39 underestimated and concurrently species that render shorter amplification fragments are overestimated. However, this correlation between amplicon length and estimation is not valid for all 40 the species analysed. Variability in the amplification lengths contributed to the preferential 41 42 amplification phenomenon. DNA extracted from twenty fermented cocoa bean samples were used 43 to assess the performance of the two target regions. Overall, the metataxonomic data set recovered similar taxonomic composition and provided consistent results in OTU richness among biological 44 45 samples. However, 26S region provided higher alpha diversity index and greater fungal rRNA taxonomic depth and robustness results compared with ITS2. Based on the results of this study we 46 suggest the use of the 26S region for targeting fungi. Furthermore, this study showed the efficacy of 47 the manually curated reference database optimized for annotation of mycobiota by using the 26S as 48 49 a gene target.

50 Keywords: Amplicon sequencing; fungal ecology; primer bias; Illumina; fungal database.

51 **1. Introduction**

Fungi are eukaryotic microorganisms which belong to one of the most diverse kingdoms on 52 Earth (Blackwell, 2011). They play an important role in the safety, quality, and stability of all 53 54 foodstuff to some degree, whether they are required during processing or whether they have a negative impact during shelf life. Therefore, tracking fungal communities of food systems has been 55 a concern in food research. To date, most recent studies on the microbial diversity of fermented 56 food such as vegetable, seafood, beverages, cheese, olives and spontaneously fermented American 57 cool ship ale fermentations have employed amplicon sequencing approaches (Bokulich et al., 2012; 58 Cocolin et al., 2013; Ercolini et al., 2012; Li et al., 2011; Roh et al., 2010). 59

60 Illumina sequencing platform has been currently providing a sensitive description of the microbial dynamics within food ecosystems. Some of the advantages of using this technology is 61 62 that it yields greater sequencing coverage and increased sample throughput at lower cost *per* sample 63 compared to other platforms (Caporaso et al., 2011; Quail et al., 2012). The sensitivity of this approach relies on the high coverage and accurate taxonomic resolution of short amplicon length 64 65 (Quail et al., 2012). Recent advances in the microbial diversity using next-generation sequencing technologies (NGS) have underlined the importance of the reliability of PCR primers targeting a 66 specific genetic marker (Bokulich and Mills, 2013). In spite of the importance of the amplification 67 of shorter fragments amplified by PCR in NGS, recent studies described a more reliable community 68 of fungi using shorter Internal Transcribed Spacer amplicons (ITS) of the nuclear ribosomal RNA 69 (rRNA) (Bokulich and Mills, 2013; Ihrmark et al., 2012). 70

The ITS region is considered the universal barcode for identification of fungi and includes the ITS1 and ITS2 regions, separated by the 5.8S gene. These two regions (ITS1-2) are characterized by high evolutionary rates and are edged by highly conserved regions with suitable target sites for universal primers (Begerow et al., 2010). However, the complete ITS region located between the 18S and 28S genes in the nuclear ribosomal RNA is considered too long for 454

sequencing or other NGS (Bellemain and Carlsen, 2010). Therefore, various primers are used to 76 77 amplify parts of the ITS region. In this study, we selected the primer ITS3ngs that targets a site in the ITS1 and the degenerate reverse primer ITS4 which targets an ITS-flanking site in the ribosomal 78 large subunit (LSU) encoding regions (White TJ, Bruns T, Lee S, 1990) based on their ability to 79 amplify fungal species through in silico analysis (Bellemain and Carlsen, 2010; Tedersoo et al., 80 81 2015). Nevertheless, the nuclear rRNA large subunit (LSU/28S/26S) and small subunit (SSU/18S) 82 genes have also been often used to address fungal diversity (Bonanomi et al., 2016; David et al., 2014; De Filippis et al., 2017b, 2017a; Garofalo et al., 2015; Stellato et al., 2015; Wang et al., 83 2015). To bring an overall perspective, most yeasts have been identified from sequence divergence 84 in the D1/D2 domain of LSU rRNA (Kurtzman and Robnett, 1997). Despite the great resolution to 85 recognized yeast species through 26S rRNA sequencing reactions, little is known about the 86 87 potential uses and bias that can be introduced when using this target region in NGS. In this context, 88 it is necessary for ecological studies to compare different targeting regions to describe the most accurate and reliable ecological populations in a food system. Given the nature of current 89 90 challenges, the selection of a suitable genetic marker for the identification of fungi will help researchers to clear current issues insight into the selection of primer sets. 91

92 The main focus of this study is to address sequencing target regions and primer biases on 93 one of the dominating taxonomic groups of fungi in the Dikarya, Ascomycota, which represents 53 % of the described species of true Fungi (Koljalg et al., 2013). This phylum is important in the food 94 95 industry and serves as a source for biomass production, but also includes known human and plant pathogens (Bekatorou et al., 2006; Berbee, 2001). The present research focused on the assessment 96 97 of two different targeting sites for amplicon-based Illumina NGS studies. We tested the 26S primer 98 set, delivering high coverage and accurate taxonomic assignment of short (~ 400 bp) fungal 99 amplicon versus the performance with the ITS2 region. This research intends to bring new insights 100 in the field of taxonomic assignment, validation and resolution of uncertainties on using ampliconsequencing approaches for fungi identification by using mock samples as well as fermented 101

samples. Attention was paid for monitoring fungi in mock communities and biological samples,
where taxonomic assurance of the technique, and mapping and monitoring fungi dynamics are
investigated for food applications.

105 2. Materials and methods

106 2.1. Primer selection and in silico analysis

107 Primer pairs targeting the ITS2 region (Tedersoo et al., 2015), and the D1 domain of 26S rRNA gene (Cocolin et al., 2000), were selected and reported in Table 1. For the amplification of 108 the D1 domain of the 26S, we modified the LS2-MF primer sequence position from reverse to 109 110 forward, corresponding to nucleotide position 266 of Saccharomyces cerevisiae 26S gene as described by Cocolin et al., (2000) and a reverse primer NL4 (Jespersen et al., 2005). The Illumina 111 112 overhang adapter sequences were added to locus-specific sequences. The D1 region from the 26S gene was amplified in silico to compare primer specificity and taxonomic coverage of both LS2-MF 113 and NL4 by using Primer Prospector (Walters et al., 2011) against the constructed 26S databases 114 115 and SILVA's database.

116 2.2. Mock community preparation, DNA extraction, and PCR amplification

117 Strains of yeast and filamentous fungi listed in Table 2 (DISAFA collection, Torino) were used and cultured on Malt Extract Agar (Oxoid, Milan, Italy) plus 25 mg l⁻¹ streptomycin (Sigma, 118 Milan, Italy) incubated at 28 °C for 72 h for yeast and 7 to 10 days for fungi. DNA extraction from 119 120 yeast was carried out from a loopful of grown culture while 250 mg of mycelium was scraped from the plate for filamentous fungi. DNA extraction was carried out as described by Cocolin et al., 121 (2000). DNA from each strain was quantified by using the Qubit dsDNA assay kit (Thermo Fisher 122 Scientific, Milan, Italy) and standardized at 5 ng/µl. A pool (Mock-DNA) containing each of the 123 standardized strain DNA was then obtained and subject to amplification of the ITS2 and the 26S 124 regions. PCR was carried out for the two target regions using a PCR mixture prepared with 12.5 µl 125

of the 2X Kapa HiFi HotStart ReadyMix Taq (Roche, Milan, Italy), 1 μ M each primer, 2.5 μ l of DNA template, and PCR-grade water. Each PCR were subject to the following amplification conditions: thirty cycles of 30 s of denaturation (95 °C), 30 s of primer annealing (55 °C), and 30 s of primer elongation (72 °C), followed by a final elongation step (72 °C) of 10 min.

The amplification of each fungal strain was carried out by using the same couple of primers, each amplicon was then purified using the Agencourt AMPure XP beads (Beckman Coulter Genomics) and quantified using Qubit dsDNA assay kit. Based on the amplicon size of DNA assessed by using a Biorad experion workstation (Biorad, Milan, Italy), amplicons concentration was determined. Amplicons were diluted at 20 mM and aliquots of 10 µl were pooled together to construct a Mock-Amp. In total, two independent Mock-DNA and Mock-Amp were obtained by two independent DNA extraction, quantification and pooling procedure.

137 2.3. DNA extraction and PCR amplification of fermented cocoa beans

A total of twenty fermented cocoa beans samples were collected and DNA extracted as 138 following original study (Mota-Gutierrez et al., 2018). Samples were collected during a 139 fermentation period of 0, 48, 96 and 120 h. Detailed information of samples is reported in Mota-140 Gutierrez et al., 2018 and in supplementary table S1. Briefly, total DNA was extracted from the 141 142 pellet of cocoa matrix by using the MasterPure Complete DNA & RNA Purification kit (Illumina Inc, San Diego, CA) following the manufacturer's instructions. DNA was quantified by using the 143 Qubit dsDNA assay kit (Thermo Fisher Scientific), standardized at 5 ng/µl and subject to 144 amplification of the two target regions using primers and procedure as described above. 145

146 *2.4. Library preparation and sequencing*

147 Sequencing was performed for the two target regions and for the three target samples 148 (Mock-DNA, Mock-Amp and cocoa samples). After the first purification step following the 149 Illumina sample preparation procedure, the library was combined with the sequencing adapters and dual indices using the Nextera XT Index Kit (Illumina, San Diego, USA), obtaining the multiplexed paired-end libraries. Individual libraries concentration in nM were calculated based on the size of amplicons by using a Biorad Experion workstation (Biorad) and diluted to 4 nM, denaturated with 0.2 N NaOH and spiked with 20 % (v/v) of PhiX. The combination of pool library and PhiX were diluted to 12 pM and paired-end sequencing was performed on the MiSeq platform, using MiSeq Reagent Kit V3 (2 x 250bp) (Illumina, San Diego, USA), following the standard Illumina sequencing protocol.

157 2.5. Constructed 26S rRNA sequence database

The construction database of fungal rRNA gene sequence of the 26S gene was used to select 158 primers, which amplify the D1 region of a broad fungal taxa. The sequences were downloaded from 159 the Nucleotide database of the National Center for Biotechnology Information (NCBI; 160 https://www.ncbi.nlm.nih.gov/nucleotide/; accessed March 07, 2018). The database was constructed 161 162 using the large subunit rRNA gene sequences, 23.381 sequences were downloaded using diverse taxonomic ID and the query word "26S rRNA". The final constructed database consisted of 4 phyla, 163 27 classes, 172 families, and 442 fungal strains. Incomplete sequences or sequences with absent 164 taxonomies were removed. Duplicate sequences and sequences that clustered together at 99 % of 165 166 similarity were discarded by using Prinseq and USEARCH respectively (Schmieder and Edwards, 2011). A taxonomy file, matching exactly seven taxonomic levels (root, subphylum, class, order, 167 family, genus and species) was generated from the corresponding taxonomy strings to be 168 169 compatible with implementation in the NGS analysis pipeline QIIME. Both files were manually curated for accuracy and consistency. Sequences obtained from the constructed 26S database from 170 171 biological and mock samples were compared using SILVA database. All sequences identified by D1 domain of 26S rRNA sequence analysis from biological samples and mock communities were 172 173 compared with our constructed database.

174 *2.6. Bioinformatics*

Paired-end reads (2x250 bp) were first merged using the FLASH software (Magoč and Salzberg, 175 2011), with default parameters. Joint reads were further quality filtered (Phred < Q20) using the 176 QIIME 1.9.0 software (Caporaso et al., 2011). Chimeras were then removed with the adopted 177 USEARCH version 8.1 software. Lastly, OTUs were picked at 99 % of similarity by means of 178 UCLUST clustering methods (Edgar, 2010) and representative sequences of each cluster were used 179 to assign taxonomy. For the 26S data, each cluster was used to assign taxonomy using the 180 181 Constructed 26S rRNA gene database and SILVA, while for the ITS dataset the UNITE rRNA ITS database version 2012, by means of the RDP Classifier. Sequences were double-checked using the 182 BlastN search tool (http://www.ncbi.nlm.nih.gov/blast/) to confirm the taxonomy assignment. 183 Cocoa samples datasets (ITS and 26S) were rarefied at 10,018 reads after raw read quality filtering, 184 and OTU tables were filtered for OTUs occurring at 1 % of the relative abundance in at least 2 185 186 samples. While for mock community reads from the two target regions were rarefied at 17,313 reads. 187

188 2.7. Statistical analyses

Statistics and plotting were carried out in the R environment (www.r-project.org). Alpha 189 diversity indices were calculated using the diversity function of the vegan package (Dixon, 2007). 190 191 OTUs table were used to find differences between target regions by Anosim statistical test in R environment. A two-sided permutation test with 999 permutations was performed to compare the 192 OTUs distribution and alpha diversity between the two datasets. Pairwise Kruskal-Wallis Wilcoxon 193 194 test or one way- ANOVA coupled with the Duncan honestly significant difference (HSD) test were used as appropriate to determine significant differences in alpha diversity or OTU abundance from 195 196 mock communities and biological samples. Statistical analysis was acquired through the function aov through the stats package and principal component analysis were plotted using the function 197 *dudi.pca* through the *made4* package using R version 3.3.2 198

199 2.8. Accession numbers

The ITS and 26S rRNA gene sequences are available at the Sequence Read Archive of the National Center for Biotechnology Information (NCBI), under the SRA accession number SRP126081 (fermented cocoa samples ITS) and SRP150401 (fermented cocoa samples 26S and mock sequences data).

204

205 **3. Results**

206 *3.1. In silico performance of 26S primers*

207 We performed an *in-silico* analysis of the 26S primer set against our constructed database and SILVA using Primer Prospector. LS2-MF primer showed the lowest weighted score (Fig. S1A) 208 209 indicating higher coverage across the database sequences and lower number of mismatches if 210 compared with NL4 (Fig. S1B). Comparing the taxonomic coverage of LS2-MF and NL4 against Zygomycota, Glomeromycota, Ascomycota and Basidiomycota sequence, LS2-MF showed the best 211 performance with a coverage higher than 80 % for all the phyla except for *Glomeromycota* (Fig. 212 213 S2A), while NL4 account for the 20 % of the coverage against our constructed database (Fig. S2B). Regarding the performance of the primer sets against the SILVA's database, the score of the 214 primers was higher compared with our database (data not shown). 215

216 *3.2. Performance of primers by mock community analysis*

A mock community containing twenty fungal species (Table 2) was prepared to validate the performance of the two target regions. The possible effect of the bias introduced by PCR (Mock-DNA) and that of sequencing (Mock-Amp) was then evaluated. Amplicon length of the single species showed little variation when the 26S gene was amplified (461 ± 30 bp) while for ITS2 we observed greater dispersion in size (445 ± 55 bp) (Table 2). Significant difference in mycobiota composition (Anosim statistical test, P < 0.05) by using the two target regions or mock communities (DNA or AMP) was observed by Principal Component Analysis (Fig. 1). 224 In both samples (Mock-DNA and Mock-Amp), the target region ITS2 showed similar abundances with respect to the theoretical value for two fungal species, namely Torulaspora 225 delbrueckii and Plectosphaerella cucumerina (Table 3). Similarly, with the 26S target gene, for six 226 species, abundances retrieved from both Mock-DNA and Mock-Amp samples were comparable to 227 the theoretical values (Aspergillus fumigatus, Pichia membranifaciens, Pichia kudriavzevii, 228 229 Penicillium glabrum, Penicillium brevicompactum and Starmerella bacillaris). Furthermore, for the 230 26S region, the species Alternaria alternata, Aspergillus flavus and Fusarium oxysporum rendered different abundances in the Mock-DNA and the Mock-Amp but in both samples the values were 231 comparable to the theoretical. 232

For 18 out of the 20 fungal species tested, the ITS2 region resulted in underestimation or 233 overestimation with respect to the theoretical value in the Mock-DNA, Mock-Amp or both (P <234 235 0.05). Four species were significantly overestimated (A. fumigatus (439bp), F. verticillioides (415bp), K. marxianus (521bp) and P. brevicompactum (428bp)) while other 4 were significantly 236 underestimated (Galactomyces geotrichum (324bp), Hanseniaspora opuntiae 237 (484bp), 238 Schizosaccharomyces pombe (562bp) and Starmerella bacillaris (361bp)), in both samples. Interestingly, G. geotrichum and S. pombe were not detected in any of the two samples. Nine more 239 species resulted to be significantly different from the theoretical value (either higher or lower) in the 240 241 Mock-DNA or Mock-Amp sample only (Table 3).

When the 26S region was targeted, 11 out of the 20 species were either underestimated or overestimated in the Mock-DNA, Mock-Amp or both. More specifically, 4 species were underestimated in both types of samples (*Candida sake* (467bp), *H. opuntiae* (435 bp), *Kluyveromyces marxianus* (427 bp), *T. delbrueckii* (461 bp)) and only one (*G. geotrichum* (502bp)) was overestimated. Five more species were significantly different from the theoretical value (either higher or lower) in the Mock-DNA or Mock-Amp sample only (Table 2 and 3). It also should be 248 pointed out that we did not observe a clear correlation between amplicon size and over or 249 underestimation.

In the Mock-Amp samples, correct relative quantification was obtained for 13 out of the 20 species targeting the 26S region and for 10 out of 20 species with the ITS2 region. In the Mock-DNA samples, correct relative quantification was obtained for 10 out of 20 species in the 26S region and for 4 out of 20 species targeting the ITS2 (Table 3).

Remarkably, *G. geotrichum* and *S. pombe* were only detected when the 26S region was targeted while *H. opuntiae* was the only species that was consistently underestimated, independently from the target region or sample. Overall, 26S sequencing data aligned better to theoretical abundance values for the fungal species tested than did the ITS sequencing data.

258 *3.3. Mycobiota in biological samples*

Sequencing of twenty fermented cocoa beans samples collected during a previous 259 experiment, after amplification with the primers ITS2 and 26S showed a mean sequence length of 260 261 412 and 390 bp. respectively and an estimated sample coverage of 97.73 and 95.87 %, respectively 262 (See Table S1). The 26S target region revealed greater OTU richness compared to the ITS2 region (P < 0.05) as shown in Fig. 2. Overall, 20 and 37 fungal OTUs were identified during the 263 fermentations using the primer set ITS2 and 26S, respectively. In addition, we observed differences 264 in length distribution across the two target genes. Histogram of reads length of 26S showed that the 265 266 higher reads proportion were around 380 bp while for ITS2 we observed a varied distribution of the reads length around 370bp, 400bp, 420bp, and 450bp (See Fig. S3). Eleven OTUs, namely Candida 267 268 jaroonii, Candida tallmaniae, Fusarium, Hanseniaspora, H. opuntiae, Hanseniaspora uvarum, K. 269 marxianus, S. cerevisiae, Saccharomycopsis crataegens, T. delbrueckii and Pichia pijperi were detected by both targeting regions (Fig. 3). The relative abundance of several fungal species was 270 significantly different according to the type of amplicon used (P < 0.05, Fig. 4), in which 271 272 significantly higher relative abundance was found for S. cerevisiae, P. pijperi, and H. uvarum (P < 273 0.05) using ITS2 target gene, while *Hanseniaspora* showed higher abundance when using the 26S 274 (P < 0.05, Fig. 4).

275 3.4. Performance of the new constructed 26S database against SILVA

To validate the new 26S database, biological samples and mock communities identified by 276 D1 domain of 26S rRNA sequence analysis were compared with SILVA's database (Ouast et al., 277 2013). Significant difference in mycobiota composition of the two databases was observed in mock 278 279 communities (Anosim statistical test, P < 0.05). In detail, the constructed 26S database assigned successfully the twenty fungal species, while SILVA's database assigned only ten (A. alternata, H. 280 osmophila, K. marxianus, P. kudriavzevii, P. membranifaciens, S. ludwigii, S. pombe, S. bacillaris, 281 T. delbrueckii, Table 4). Interestingly, S. cerevisiae was not detected by using the SILVA's 282 database from the mock communities. In contrast, no significant differences in the OTU 283 distributions and the alpha diversity calculations were observed between the two datasets from 284 biological samples. In depth, the constructed 26S database assigned 37 fungal species, described 285 above, while SILVA's database assigned 35 (data not shown). However, we observed that S. 286 cerevisiae and H. uvarum were not detected by using the SILVA's database. 287

288 4. Discussion

289 New tools and molecular techniques have been used to detect microbial ecology in the past decades. Recently, the interest in the use of amplicon sequencing to identify taxonomically relevant 290 291 taxa in food has increased. However, this approach has potential biases as previously described 292 (Bowers et al., 2015; Fouhy et al., 2016) where primer selection is considered one of the most 293 important sources of biases (Bokulich et al., 2014; Bonanomi et al., 2016; David et al., 2014; De 294 Filippis et al., 2017a; Ercolini, 2013; Garofalo et al., 2015; Stellato et al., 2015; Stielow et al., 2015; Wang et al., 2015). The 26S region (D1 domain) of the rRNA encoding gene and the ITS2 region 295 have been proposed as good candidates for identifying fungal species when using NGS technologies 296 due to the high taxonomic resolution (Tedersoo et al., 2015). In this study, we performed a 297

comparative evaluation of two regions as amplicon sequencing targets for the identification of fungiand it also describes the mycobiota community in food matrices.

Recently the 26S region has been studied using the Roche 454 technology (De Filippis et al., 300 301 2017b). However, this platform has been shown to result in high sequencing errors due to A and T 302 rich homopolymers (Luo et al., 2012), while Illumina does not present this sequencing error (Erlich et al., 2008). Our results and a previous study (De Filippis et al., 2017b) reveal that 26S gene is a 303 reliable target site for both NGS technologies (Roche 454 and Illumina) for eukaryotic species. In 304 order to evaluate the effect of the target gene used we compared the sequencing results of both 305 306 DNA samples and mock communities. In our study, different relative abundances were obtained for both mock communities and biological samples and these differences were based on the PCR target 307 308 used. We observed in such cases that ITS2 target region led to underestimations of species with 309 longer fragments (S. pombe, and H. opuntiae), while an overestimation of shorter fragments occurred (F. vercillioides, A. fumigatus and P. brevicompactum). However, it should be pointed out 310 that this correlation between amplicon length and estimation is not valid for all the species 311 analysed. Apart from amplification length, other parameters that influence relative abundance 312 calculations of taxa within samples could be considered. Sampling errors, different primers 313 alignment efficiencies during PCR amplification, performance of degenerate primers used during 314 PCR amplification, result in wrong representation in terms of relative abundance of microbial 315 populations (Ihrmark et al., 2012; Polz and Cavanaugh, 1998). 316

In addition to underestimation of abundances, "identification bias" is also common to amplicon-based analyses, where minor groups are poorly represented (Koljalg et al., 2013). The lack of updated reliable public reference data set and the discrepancies to refer to fungal species have been recently demonstrated for the ITS sequences (Koljalg et al., 2013). This is also in accordance with our results, suggesting that our new database for the 26S, validated by the widely used SILVA, proved to be a curated and rich database to be used. Differences between the two 323 databases regarding taxonomic classification of sequences were obtained. The newly constructed 26S database delivered a more precise taxonomic assignment of the sequences. This could be due to 324 the fact that SILVA database comprised also non-microbial sequences, incomplete sequences and 325 sequences with unassigned taxonomy. In contrast, each taxonomy in our database was double 326 checked to get the higher taxonomic resolution, obtaining clearly more robustness results in terms 327 of taxonomic assignment from the biological samples. Special attention must be paid on the miss-328 329 identification of fungal strain on the current available database. This current issue is pointed out in this study, in which S. cerevisiae and H. uvarum were misidentified from fermented samples, using 330 SILVA's database. 331

Given the intricate nature of PCR, the amplification of biological samples has been 332 problematic (Polz and Cavanaugh, 1998). Our results exhibited high proportion of fungal coverage 333 334 (98 - 96 %) by both primer pair sets, which suggested that fungi account for roughly the complete eukarvote rRNA in the studied fermented cocoa beans on average. The results also highlight a 335 lower biodiversity of fungal communities for ITS2 compared with the 26S region in fermented 336 337 cocoa beans, which contradicts with previous studies where ITS region has been used in NGS 338 studies as the universal primer set for fungi (Bokulich et al., 2014, 2012; Tedersoo et al., 2015). Such discrepancies between outcomes of different studies may arise on account of the biased 339 340 quantification of relative abundance of taxa due to the uneven length of ITS fragments (Bellemain and Carlsen, 2010), the preferential amplification of rRNA genes for certain taxa by PCR, 341 sequencing bias due to unequal amplification of the target gene or due to inaccurate taxonomic 342 classification of reference databases (Simon and Daniel, 2011). Despite these challenges, the greater 343 recovery trends in the community composition in the 26S target region observed here, have been 344 345 supported from previous studies, where higher discrimination power of species identification in 346 early diverging lineages of LSU compared with ITS was reported (Schoch et al., 2012).

347 Our study suggested that the 26S as a target showed greater biodiversity in biological samples compared with the universal primer ITS. However, it should be noted that the present study 348 shows the performance of a new pair of primers targeting the 26S region for fungal strains. 349 Therefore, the novelty of these primer sets is also our limitation, that can be successfully overcome 350 through future research focusing on the use of small fragments of the LSU region to target fungal 351 352 species, that could support our observations. Therefore, the combination of both target genes, where 353 species identification can be performed applying ITS and phylogenetic analysis with 26S, is highly recommended and the use of both will depend on the purpose of taxa investigation (Klaubauf et al., 354 2010; Schoch et al., 2012). From a molecular microbial ecology perspective, in terms of 355 classification of marker-gene sequences, there is evidently a need for more extensive testing of 356 primers targeting different genes and *loci*, to support and identify all fungal species in NGS studies. 357 358 Clearly, the benefits of characterizing fermented microbial diversity may bring important advancements to the food industry, such as discrimination of starter culture to improve food quality 359 or to accelerate processes. This study provides new insight into the selection of better primers and 360 361 taxonomical assignment to study fungal ecology, which should enable food research to gain better view of the microbial diversity present in a range of fermentations avoiding biases. One also notes, 362 the limited availability of updated databases to assess ecological populations. 363

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506 Table legend

507

| 508 | Table 1. | Primers | used for | Illumina | MiSeq | sequencing |
|-----|----------|---------|----------|----------|-------|------------|
|-----|----------|---------|----------|----------|-------|------------|

- **Table 2.** Fungal strains used for sequencing analysis and respective amplicon length
- **Table 3**. Relative abundance (%) of the fungal species identified in mock communities amplified
- using two different target regions. The expected concentration is referred to as theoretical

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515 Figure legend

516 **Fig 1.** Principal component analysis (PCA) based on mock mycobiota composition.

Fig 2. Boxplots describe α -diversity measures (Chao1, Shannon index and number of observed species) of fermented cocoa bean samples. Individual points and brackets represent the richness estimate and the theoretical standard error range, respectively.

- Fig 3. Distribution of OTUs in fermented cocoa bean samples in the amplicon datasets divided into
 26S (upper figure) and ITS (lower figure). Only OTUs with an incidence above 1 % in at least 2
 samples are shown.
- **Fig 4.** Boxplots describe statistically different species detected in fermented cocoa bean samples analysed with two different target genes. Individual points and brackets represent the relative abundance and the theoretical standard error range, respectively.
- 526

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| 530 | Supplementary table |
|-----|---|
| 531 | Table S1. Estimation of sample coverage of fungal community of fermented cocoa beans using 26S |
| 532 | target gene (A). Estimation of sample coverage of fungal community of fermented cocoa beans |
| 533 | using ITS2 target region (B) |
| 534 | Table S2. Relative abundance of mock communities identified by two different fungal databases |
| 535 | |
| 536 | Supplementary figure |
| 537 | S1. Bar chart showing the evaluation of primer efficiency. Overall matches and weight score of 26S |
| 538 | primer pair against our constructed database 26S. Assessment of LS2-MF forward primer and NL4 |
| 539 | reversed primer |
| 540 | S2. Predictive taxonomic coverage of 26S primers. Numeric values above bins represent total |
| 541 | sequence counts for each set, LS2-MF forward primer and NL4 reversed primer |
| 542 | S3. Histograms of read length of fungal communities using 26S target region and ITS2 target region |
| 543 | |
| 544 | |

Table 1

| Primer | Features | Primer sequence | Target region | Amplicon length | Reference |
|----------------|----------|-----------------------------------|---------------|--------------------|---------------------------|
| ITS3tagmix1 | Fwd | 5'-CTAGACTCGTCACCGATGAAGAACGCAG-3 | ITS2 | 385 | Tedersoo et al., 2015 |
| ITS4ngs | Rev | 5'- TTCCTSCGCTTATTGATATGC-3' | ITS2 | 565 | Tedersoo et al., 2015 |
| LS2-MF NL-4 | Fwd | 5'-GAGTCGAGTTGTTTGGGAAT-3' | LSU D1 | | This study |
| | Rev | 5'-GGTCCGTGTTTCAAGACGG-3' | LSU D1 | 369 | Jespersen et al., 2005 |

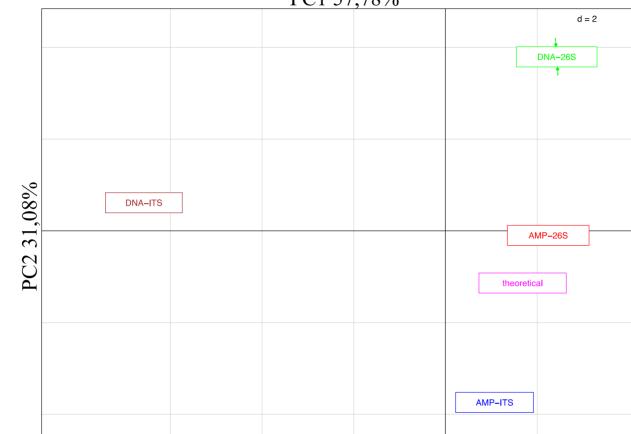
Table 2

| Fungal species | Size bp | Size bp | |
|-----------------------------|---------|---------|--|
| | (26S) | (ITS2) | |
| Alternaria alternata | 454 | 414 | |
| Aspergillus flavus | 455 | 430 | |
| Aspergillus fumigatus | 459 | 439 | |
| Candida sake | 467 | 393 | |
| Fusarium oxysporum | 462 | 405 | |
| Fusarium verticillioides | 458 | 415 | |
| Galactomyces geotrichum | 502 | 324 | |
| Hanseniaspora opuntiae | 435 | 484 | |
| Hanseniaspora osmophila | 462 | 520 | |
| Kluyveromyces marxianus | 427 | 521 | |
| Penicillium brevicompactum | 456 | 428 | |
| Penicillium glabrum | 458 | 426 | |
| Pichia kudriavzevii | 472 | 431 | |
| Pichia membranifaciens | 466 | 398 | |
| Plectosphaerella cucumerina | 457 | 441 | |
| Saccharomyces cerevisiae | 460 | 496 | |
| Saccharomycodes ludwigii | 452 | 470 | |
| Schizosaccharomyces pombe | 488 | 562 | |
| Starmerella bacillaris | 389 | 361 | |
| Torulaspora delbrueckii | 461 | 518 | |

552 **Table 3**

| OTU | Theoretical | 26 | is | Theoretical | ITS | 2 553 |
|-----------------------------|-----------------|--------------------------|--------------------------|----------------|--------------------------|--------------------------|
| | = | DNA | AMP | _ | DNA | AMP |
| Alternaria alternata | 5 ^{ab} | 3.62 ± 0.34^{b} | 5.71 ± 0.47^{a} | 5 ª | 7.67 ± 0.94^{b} | 6.63 ± 0.09^{ab} |
| Aspergillus flavus | 5 ^{ab} | 5.78 ± 0.46^{a} | 4.03 ± 0.07^{b} | 5 ª | 9.39 ± 0.17^{b} | 7.19 ± 0.26^{a} |
| Aspergillus fumigatus | 5 ª | 4.82 ± 1.34^{a} | 3.71 ± 0.09^{a} | 5 ª | 8.41 ± 0.20^{b} | $6.14 \pm 0.01^{\circ}$ |
| Candida sake | 5 ª | 0.24 ± 0.20^{b} | 0.06 ± 0.02^{b} | 5 ^a | 4.57 ± 0.38 ^a | 6.19 ± 0.03^{b} |
| Fusarium oxysporum | 5 ^{ab} | 3.99 ± 0.25^{b} | 5.08 ± 0.22^{a} | 5 ª | 8.11 ± 0.00^{b} | 6.33 ± 0.04 ^a |
| Fusarium verticillioides | 5 ª | $6.27 \pm 0.04^{\circ}$ | 4.63 ± 0.02^{b} | 5 ª | $10.44 \pm 0.12^{\circ}$ | 6.60 ± 0.27^{b} |
| Galactomyces geotrichum | 5 ª | $13.39 \pm 0.61^{\circ}$ | 6.94 ± 0.11^{b} | 5 ^b | 0.00 ± 0.00^{a} | 0.00 ± 0.00^{a} |
| Hanseniaspora opuntiae | 5 ^b | 2.94 ± 0.16^{a} | 3.46 ± 0.33^{a} | 5 ° | 0.70 ± 0.07^{a} | 3.67 ± 0.01^{b} |
| Hanseniaspora osmophila | 5 ª | 10.63 ± 0.56^{b} | 4.73 ± 0.13 ^a | 5 ^b | 1.63 ± 0.34 ª | 4.78 ± 0.08^{b} |
| Kluyveromyces marxianus | 5 ^b | 3.68 ± 0.08^{a} | 3.33 ± 0.16^{a} | 5 ª | 8.55 ± 0.33^{b} | 6.71 ± 0.08^{b} |
| Penicillium brevicompactum | 5 ^a | 6.00 ± 0.92^{a} | 3.55 ± 0.29 ^a | 5 ª | $14.39 \pm 0.09^{\circ}$ | 9.04 ± 0.08^{b} |
| Penicillium glabrum | 5 ^a | 4.86 ± 0.19^{a} | 4.88 ± 0.22^{a} | 5 ^b | 7.08 ± 0.11 ^a | 5.69 ± 0.07^{b} |
| Pichia kudriavzevii | 5 ^a | 7.24 ± 1.89^{a} | 10.76 ± 0.45^{a} | 5 ^b | 0.12 ± 0.14 ª | 2.11 ± 0.16^{b} |
| Pichia membranifaciens | 5 ^a | 6.63 ± 1.83^{a} | 4.82 ± 0.28^{a} | 5 ^b | 2.58 ± 0.39^{a} | 3.57 ± 0.13^{b} |
| Plectosphaerella cucumerina | 5 ^b | 1.11 ± 0.13^{a} | 4.33 ± 0.02^{b} | 5 ^b | 3.51 ± 0.62 ^a | 4.45 ± 0.26^{b} |
| Saccharomyces cerevisiae | 5 ^b | 1.86 ± 0.09^{a} | 4.20 ± 0.49^{b} | 5 ^a | 4.74 ± 0.03 ° | 5.44 ± 0.10^{b} |
| Saccharomycodes ludwigii | 5 ^b | 3.51 ± 0.00^{a} | 4.49 ± 0.00^{b} | 5 ^b | 0.68 ± 0.03^{a} | 4.66 ± 0.03^{b} |
| Schizosaccharomyces pombe | 5 ^b | 5.00 ± 0.11^{b} | 3.71 ± 0.19^{a} | 5 ^b | 0.01 ± 0.00^{a} | $0.00 \pm 0.00^{\circ}$ |
| Starmerella bacillaris | 5 ª | 4.21 ± 1.68 ª | 5.91 ± 0.11 ^a | 5 ^b | 0.16 ± 0.07^{a} | 0.19 ± 0.10^{b} |
| Forulaspora delbrueckii | 5° | 2.46 ± 0.02^{a} | 3.90 ± 0.17^{b} | 5 ª | 5.38 ± 0.11 ª | 5.30 ± 0.52 ° |

Values are expressed as the mean from duplicate determinations (%). Different letters indicate statistical difference related to relative abundances of mock communities using least significant difference test (P < 0.05). *P*-values were adjusted using Bonferroni's method. Different colour showed no difference (grey), underestimation (light green) or overestimation (light blue) between mock samples and theoretical data.



PC1 57,78%

Figure 1.

