

1 **Revealing antimicrobial resistance profile of the novel probiotic candidate**

2 ***Faecalibacterium prausnitzii* DSM 17677**

3

4 Daniela Machado ^{a#}, Joana Barbosa ^{a#}, Melany Domingos ^a, Diana Almeida ^a, José

5 Carlos Andrade ^b, Ana Cristina Freitas ^{a*†}, Ana Maria Gomes ^a

6

7 ^a Universidade Católica Portuguesa, CBQF - Centro de Biotecnologia e Química Fina –

8 Laboratório Associado, Escola Superior de Biotecnologia, Rua Diogo Botelho 1327,

9 4169-005 Porto, Portugal

10 ^b CESPu, Instituto de Investigação e Formação Avançada em Ciências e Tecnologias da

11 Saúde, Rua Central de Gandra, 1317, 4585-116 Gandra PRD, Portugal

12 dmachado@porto.ucp.pt

13 jcbarbosa@porto.ucp.pt

14 melanydomingos@hotmail.com

15 dalmeida@porto.ucp.pt

16 jose.andrade@iucs.cespu.pt

17 †afreitas@porto.ucp.pt (deceased)

18 amgomes@porto.ucp.pt

19 # First Authors with equal contribution

20

21 *Corresponding author at: Universidade Católica Portuguesa, CBQF - Centro de

22 Biotecnologia e Química Fina – Laboratório Associado, Escola Superior de

23 Biotecnologia, Rua Diogo Botelho 1327, 4169-005 Porto, Portugal

24 E-mail address: afreitas@porto.ucp.pt

25 **Abstract**

26 *Faecalibacterium prausnitzii*, a resident anaerobic bacterium commonly found in
27 healthy gut microbiota, has been proposed as a next generation probiotic with high
28 potential for application in food matrices and pharmaceutical formulations. Despite its
29 recognized health benefits, detailed information regarding its antimicrobial
30 susceptibility profile is still lacking. However, this information is crucial to determine
31 its safety, since the absence of acquired antimicrobial resistance is required to qualify a
32 probiotic candidate as safe for human and animal consumption. Herein, the
33 antimicrobial susceptibility profile of *F. prausnitzii* DSM 17677 strain was evaluated by
34 integrating both phenotypic and *in silico* data. Phenotypic antimicrobial susceptibility
35 was evaluated by determining minimum inhibitory concentrations of 9 antimicrobials
36 using broth microdilution and E-test® methods. Also, the whole genome of
37 *F. prausnitzii* DSM 17677 was analysed, using several databases and bioinformatics
38 tools, to identify possible antibiotic resistance genes (ARG), genomic islands (GI) and
39 mobile genetic elements (MGE). With exception of erythromycin, the same
40 classification (susceptible or resistant) was obtained in both broth microdilution and E-
41 test® methods. Phenotypic resistance to ampicillin, gentamycin, kanamycin and
42 streptomycin were detected, which was supported by the genomic context. Other ARG
43 were also identified but they seem not to be expressed under the tested conditions.
44 *Faecalibacterium prausnitzii* DSM 17677 genome contains 24 annotated genes
45 putatively involved in resistance against the following classes of antimicrobials:
46 aminoglycosides (such as gentamycin, kanamycin and streptomycin), macrolides (such
47 as erythromycin), tetracyclines and lincosamides. The presence of putative ARG
48 conferring resistance to β -lactams could only be detected using a broader homology
49 search. The majority of these genes are not encoded within GI or MGE and no plasmids

50 were reported for this strain. Despite the fact that most genes are related with general
51 resistance mechanisms, a streptomycin-specific ARG poses the only potential concern
52 identified. This specific ARG is encoded within a GI and a MGE, meaning that it could
53 have been laterally acquired and might be transferred to other bacteria present in the
54 same environment. Thus, our findings provide relevant insights regarding the
55 phenotypic and genotypic antimicrobial resistance profiles of the probiotic candidate
56 *F. prausnitzii* DSM 17677.

57

58 **Keywords**

59 Antimicrobial resistance genes, *Faecalibacterium prausnitzii*, *in silico* analysis, safety
60 profile, susceptibility testing.

61 **1. Introduction**

62 Globally, antimicrobials remain as one of the main therapeutic approaches indicated for
63 the treatment of several infectious diseases (Aminov, 2017). Nevertheless,
64 overprescription and misuse of such antimicrobial agents has driven the worldwide
65 emergence, evolution and dissemination of antimicrobial resistance (AR) in pathogenic
66 and non-pathogenic bacteria associated with humans, animals and the environment
67 (Hernando-Amado et al., 2019; Sirichoat et al., 2020). Considering the growing
68 emergence of AR, joined with a lack of new antimicrobial agents, AR has been
69 considered a major global public health issue (World Health Organization, 2014). In
70 this context, food microorganism as probiotics are now coming under scrutiny for
71 possible involvement in the spreading of AR (Campedelli et al., 2019; Selvin et al.,
72 2020). The existence of AR in probiotic strains could be advantageous when they are
73 used as adjuvant of antibiotic therapy (Neut et al., 2017). However, if those
74 determinants have an acquired nature, the consumption of probiotics pose a potential
75 risk of dissemination in gut microbiota and in environment (Anisimova and Yarullina,
76 2019; Selvin et al., 2020).

77 Currently, several strains of the intestinal commensal bacterium *Faecalibacterium*
78 *prausnitzii* are proposed as next generation probiotic candidates, mostly due to their
79 specific bioactivities against several intestinal inflammatory conditions, which provoke
80 interest in the application in food matrices and pharmaceutical formulations (Almeida et
81 al., 2020; Andrade et al., 2020). Specifically, the administration of either live
82 *F. prausnitzii* DSM 17677 or its supernatant reduced the severity of induced colitis and
83 counterbalancing the associated dysbiosis, in a mice model (Martín et al., 2014; Sokol
84 et al., 2008). Despite the multiple beneficial effects of these novel probiotic candidates,
85 there is little information regarding their safety profile, namely in terms of antimicrobial

86 susceptibility pattern (Foditsch et al., 2014; Martín et al., 2017). In European countries,
87 the absence of acquired AR is one of the most important selection criteria for a probiotic
88 candidate in order to be stated as safe for human and animal consumption and,
89 consequently, to be included in the list of qualified presumption of safety (EFSA-
90 FEEDAP et al., 2018). For this purpose, phenotypic testing based on determination of a
91 minimum inhibitory concentration (MIC) for a selected group of antimicrobials, as well
92 as *in silico* detection of AR genes in whole genome sequence, should be performed
93 (EFSA-FEEDAP et al., 2018). Notwithstanding, this methodology has not yet been
94 widely implemented in the scope of probiotics (Rozman et al., 2020). In particular,
95 scarce data on phenotypic and *in silico* analyses of AR in *F. prausnitzii* strains was
96 previously reported (Bag et al., 2017; Foditsch et al., 2014; Martín et al., 2017). In this
97 context, the present study aimed to assess antimicrobial susceptibility profile of novel
98 probiotic candidate *F. prausnitzii* DSM 17677, using phenotypic and *in silico* methods.
99

100 **2. Material and methods**

101 **2.1. Bacterial strain and growth conditions**

102 *Faecalibacterium prausnitzii* strain DSM 17677 (Leibniz Institute DSMZ-German
103 Collection of Microorganisms and Cell Cultures, Braunschweig, Germany) was used for
104 this study. It was initially cultured according to the recommended conditions proposed
105 by DSMZ (2020). This strain was kept frozen at -80°C in sBHI [Brain Heart Infusion
106 medium (37g/L; VWR International, Leuven, Belgium) supplemented with yeast extract
107 (5g/L; VWR International), hemin (5mg/L; Alfa Aesar, Kandel, Germany), vitamin K1
108 (5µL/L; Sigma-Aldrich, St. Louis, Missouri, USA) and L-cysteine (2g/L; Alfa Aesar)],
109 as previously used by Maier et al. (2017), with 20% (v/v) of glycerol (Fisher Scientific,
110 Loughborough, United Kingdom). For each assay, a *F. prausnitzii* DSM 17677 glycerol

111 stock was grown in sBHI for 16 h at 37 °C in an anaerobic incubator (Whitley A35
112 HEPA Anaerobic Workstation, Bingley, United Kingdom) under an atmosphere of 85%
113 N₂, 5% H₂ and 10% CO₂. Afterwards, 0.1 mL of the previously grown culture was
114 transferred to 10 mL of sBHI medium, and this bacterial suspension was incubated
115 anaerobically for 9-10 h at 37°C for the antimicrobial susceptibility testing.

116

117 **2.2. Antimicrobial susceptibility testing**

118 Minimum inhibitory concentration of 9 clinically relevant antimicrobials recommended
119 by EFSA-FEEDAP guidelines (2018) were determined by broth microdilution method
120 in accordance with guidelines from Clinical and Laboratory Standards Institute (2020)
121 and Wiegand et al. (2008), with some minor modifications. The recommended
122 antimicrobials were: ampicillin (0.25-2 µg/mL), vancomycin (1-8 µg/mL), gentamicin
123 (1-8 µg/mL), kanamycin (4-32 µg/mL), streptomycin (2-16 µg/mL), erythromycin
124 (0.25-2 µg/mL), clindamycin (1-8 µg/mL), tetracycline (0.5-4 µg/mL) and
125 chloramphenicol (1-8 µg/mL). All antimicrobials were purchased from Sigma, with the
126 exception of chloramphenicol that was purchased from Merck (Darmstadt, Germany).
127 The stock solutions of antimicrobials were prepared according to the manufacturer's
128 instructions. Briefly, *F. prausnitzii* DSM 17677 grown culture in sBHI broth was used
129 to prepare a bacterial suspension with cell concentration of 10⁶ CFU/mL. Afterwards,
130 50 µL of two-fold serial dilutions of each antimicrobial previously prepared in sBHI
131 broth were added to each well of a 96-well round base microtiter plate (Sarstedt,
132 Nümbrecht, Germany). Then, each well was inoculated with 50 µL of bacterial
133 suspension (10⁶ CFU/mL) and the 96-well microtiter plates were incubated at 37 °C
134 during 48 h under anaerobic conditions. Moreover, a negative control, containing only
135 sBHI medium (100 µL/well), and a positive control that contained 50 µL of bacterial

136 suspension (10^6 CFU/mL) with 50 μ L of sBHI medium per well were included. After
137 incubation, the MIC was evaluated and defined as the lowest antimicrobial
138 concentration that inhibited visible bacterial growth. Additionally, MIC for these 9
139 antimicrobials were also determined using the E-test® (Biomérieux, Marcy-l’Etoile,
140 France), in accordance with guidelines recommended by the supplier with some minor
141 changes. In brief, the bacterial suspension prepared in sBHI broth, with a cell
142 concentration of 3×10^8 CFU/mL was spread evenly onto sBHI agar plates using a sterile
143 cotton swab. Then, the E-test® strips were placed onto the surface of the plates that
144 incubated under same conditions as for the broth microdilution method. After 48 h of
145 incubation, MIC values (in μ g/mL) were read where the pointed end of the ellipse
146 intersects the strip. Both antimicrobial susceptibility tests were performed at least three
147 independent times with two replicates in each assay. Using microbiological cut-off
148 values established by EFSA-FEEDAP guideline, *F. prausnitzii* DSM 17677 strain was
149 classified as susceptible or resistant to a certain antimicrobial (EFSA-FEEDAP et al.,
150 2018).

151

152 **2.3. *In silico* analysis of antibiotic resistance genes (ARG)**

153 Genome sequences of *F. prausnitzii* strains were obtained from GenBank (Benson et al.,
154 2018) in August 2020, in a total of 104 sequences. The presence and distribution of
155 ARG in these genomes was investigated using PATRIC – the Pathosystems Resource
156 Integration Center (Davis et al., 2020). PATRIC retrieves genome annotations and other
157 curated data associated with each genome, including laboratory-derived AR phenotypes
158 if available, host organisms, isolation sources and geographical information. For the
159 analysis of the 104 *F. prausnitzii* genome sequences, the annotated “Speciality genes”
160 were filtered to retrieve only the “Antibiotic Resistance” related genes, whose source

161 was PATRIC. The total number of hits was then subdivided into antibiotic related
162 classes and the percentage of genes falling in each class was assessed.
163 PATRIC was also used to retrieve the ARG encoded within *F. prausnitzii* DSM 17677
164 genome (= A2-165; Accession number: NZ_CP022479.1), the strain used in the present
165 study for the phenotypic tests, following the same criteria. Additionally, the same
166 genome was screened using ResFinder 3.1 (Bortolaia et al., 2020) and CARD (version
167 3.1.0) – The Comprehensive Antibiotic Resistance Database (Alcock et al., 2019).
168 Particularly CARD's RGI (Resistance Gene Identifier) tool (version 5.1.1) (Jia et al.,
169 2017) was used, with homology search criteria defined to show not only “Perfect” and
170 “Strict” matches, but also “Loose” similarities.
171 Genomic islands (GI), regions of probable horizontal origin, within *F. prausnitzii* DSM
172 17677 genome were predicted using IslandViewer4 (Bertelli et al., 2017), while
173 possible integrative and conjugative elements (ICE) were detected using ICEberg 2.0
174 (Liu et al., 2019). Prophage-like sequences, putatively acquired from prophage
175 integration within *F. prausnitzii* DSM 17677 genome, were also identified, using
176 Prophage Hunter (Song et al., 2019) and PHASTER (Arndt et al., 2016; Zhou et al.,
177 2011) web tools. The guanine/cytosine (GC) content was evaluated using the genome
178 browser Artemis (version 16.0.0) (Carver et al., 2012). A representation of the full
179 genome, including all the delimited GI, integrative elements, prophage-like sequences,
180 as well as a GC distribution graphic, was constructed using SnapGene Viewer software
181 (from Insightful Science; available at snappgene.com).

182

183 **3. Results and discussion**

184 Antimicrobial resistance in established and potential probiotic strains does not constitute
185 a safety concern *per se*, when the resistance mechanisms responsible for the resistant

186 phenotype is intrinsic (EFSA-FEEDAP et al., 2018; Gueimonde et al., 2013). In fact,
187 EFSA-FEEDAP reported that only the presence of acquired AR genes coding for, or
188 contributing to, resistance to clinically relevant antimicrobials in those microorganisms
189 to be incorporated in foods or feeds should be considered a hazard (EFSA-FEEDAP et
190 al., 2018). In the present study, the antimicrobial susceptibility profile of *F. prausnitzii*
191 DSM 17677 strain was evaluated using phenotypic and *in silico* approaches.

192

193 **3.1. Phenotypic antimicrobial susceptibility profile**

194 The MIC values of 9 clinically relevant antimicrobials determined using the broth
195 microdilution and E-test® methods are displayed in Table 1. Phenotypic antimicrobial
196 susceptibility profile of *F. prausnitzii* DSM 17677 was established according to
197 microbiological cut-off values available in the EFSA-FEEDAP guideline (2018) that
198 covers the characterisation of microorganisms used as feed additives or as production
199 organisms in European Union. Considering the microbiological cut-off values for
200 *Corynebacterium* and other Gram positive bacteria from EFSA-FEEDAP guidelines
201 (2018), *F. prausnitzii* DSM 17677 strain was susceptible to vancomycin, clindamycin,
202 tetracycline and chloramphenicol. In contrast, this strain was resistant to ampicillin,
203 gentamicin, kanamycin and streptomycin. Only erythromycin failed to meet the same
204 classification in both methods, with broth microdilution technique classifying the
205 bacterial strain as resistant (MIC higher than 2 µg/mL) while the E-test® method
206 classified the strain as susceptible (MIC value ranging 0.5-1 µg/mL). Indeed,
207 disagreements in the classification of microbial species using different antimicrobial
208 susceptibility testing methods have been reported in the literature (Kulengowski et al.,
209 2019; Matuschek et al., 2018). However, it is important to note that broth microdilution

210 method is widely accepted as reference method for antimicrobial susceptibility testing
211 (Haeili et al., 2019; Hakvoort et al., 2020; Humphries et al., 2018).

212 Regarding ampicillin E-test®, it was possible to observe a phenotypic heterogeneity in
213 which small colonies grew within the broader inhibition zone, throughout a range of
214 ampicillin concentrations. According to E-test® guidelines, the inner inhibition zone,
215 where no colonies were observed, was considered for the determination of the resistance
216 phenotype (Table 1), although considering either of them would result in the same
217 classification (EFSA-FEEDAP cut-off: 1 µg/mL). This occurrence can be related with
218 the phenomenon of heteroresistance, which describes the ability of a pre-existing
219 subpopulation of cells, within an homogeneous culture, to show increased levels of
220 resistance compared with the main population (Andersson et al., 2019; Band and Weiss,
221 2019). Heteroresistance was already reported for various classes of antimicrobials,
222 including β-lactams. The mechanisms underlying this phenomenon are still unclear, but
223 it is assumed that they are related with differentially expressed, mutated or even
224 duplicated ARG within the same population (El-Halfawy and Valvano, 2015). Although
225 it can sometimes influence the result of an antimicrobial clinical treatment, this is
226 usually a transient characteristic, which is reverted upon removal of the antimicrobial
227 selective pressure (Andersson et al., 2019; Band and Weiss, 2019).

228 To the best of our knowledge, to date, only two studies had evaluated the antimicrobial
229 susceptibility profile of *F. prausnitzii* strains (Foditsch et al., 2014; Martín et al., 2017).

230 In fact, Foditsch et al. (2014) studied the antimicrobial susceptibility profile of several
231 *F. prausnitzii* strains isolated from fecal samples of healthy calves and piglets. Using
232 the E-test® on VTR2RF agar plates, these researchers verified that all isolates were
233 resistant to ciprofloxacin and sulfamethoxazole-trimethoprim, more than 50% of
234 isolates were resistant to tetracycline, cefepime and ceftiofur; 34.5% were resistant to

235 ceftriaxone; 27.6% were resistant to ampicillin. Around 55.2% of the isolates grew in
236 the highest concentration of amikacin tested and only 1 isolate was resistant to
237 gentamicin, being all isolates susceptible to chloramphenicol (Foditsch et al., 2014).
238 Later, Martín et al. (2017) tested the antimicrobial susceptibility to tetracycline,
239 kanamycin, chloramphenicol, linezolid, quinupristin/dalfopristin, trimethoprim,
240 gentamicin, erythromycin, cefpirome, clindamycin, streptomycin, vancomycin, and
241 ampicillin for *F. prausnitzii* DSM 17677 strain and other *F. prausnitzii* strains isolated
242 from human healthy volunteers using the E-test® assay on Wilkins-Chalgren agar.
243 Among the 9 antimicrobials recommended by EFSA-FEEDAP to test, these researchers
244 showed that *F. prausnitzii* DSM 17677 strain was only resistant to streptomycin,
245 contrasting with our results. These discrepancies may be explained by the different
246 growth conditions, specifically growth medium composition (sBHI agar in our work vs
247 Wilkins-Chalgren agar in Martín et al. work) or in incubation atmosphere (85% N₂,
248 10% CO₂ and 5% H₂ in our work vs 90% N₂, 5% CO₂ and 5% H₂ in Martín et al. work).
249 Indeed, growth conditions, such as media composition or CO₂ content in incubation,
250 have been reported as important factors influencing the antimicrobial susceptibility of
251 several bacterial strains (De Vecchi et al., 2008; Loose et al., 2020). Underlining this
252 rationale, EFSA-FEEDAP alerts to potential interference of medium components, test
253 type (broth microdilution vs agar dilution), and culture conditions (pH, temperature,
254 time of incubation) on the antimicrobial susceptibility profile (EFSA-FEEDAP et al.,
255 2018).

256

257 **3.2. Prevalence of ARG within the genomes of *F. prausnitzii* species**

258 Genotypic approaches have been used as a valuable complement to the phenotypic
259 antimicrobial resistance studies for several years (Courvalin, 1991). Homology-based

260 genome searches can be used to predict ARG and infer a possible resistance profile
261 (Brinkac et al., 2017). The major advantage of such approach is that it allows the
262 detection of virtually any known ARG and, the identification of new variants of these
263 genetic determinants. However, this also constitutes the major pitfall of any genotypic
264 approach: only known mechanisms can be detected, while resistance caused by new
265 mechanisms or resulting from gene expression modulation might remain unnoticed.
266 Yet, sequence data can be store indefinitely and be re-analysed when new ARG are
267 phenotypically characterized (Anjum et al., 2017).

268 In this study, the genomic information regarding possible ARG was gathered, starting
269 with a brief overview of the *F. prausnitzii* genomes deposited so far. The 104
270 *F. prausnitzii* genomes (deposited until August 2020) were analysed using PATRIC, to
271 identify ARG. PATRIC integrates metadata publicly available from genomes deposited
272 in GenBank (Benson et al., 2018) and RefSeq (Pruitt et al., 2012). Each entry is also
273 annotated using RAST (Rapid Annotations using Subsystems Technology) (Aziz et al.,
274 2008), while maintaining the original GenBank or RefSeq annotations. Thus, PATRIC
275 allows an integrated and parallel analysis of genes of interest across the various
276 platforms (Wattam et al., 2014). More recently, and given the increasing importance of
277 studying the widespread phenomena of AR, PATRIC has evolved to facilitate the
278 analysis of ARG (Antonopoulos et al., 2019). A total number of 2324 ARG were
279 identified across all the genomes analysed, conferring resistance to 19 classes of
280 antimicrobial compounds (Supplementary Table A.1). On average, each genome
281 possesses 23 annotated resistance genes, but the effective number of ARG ranges from
282 2 (*F. prausnitzii* strain ATCC 27766) to 43 (*F. prausnitzii* strain AHMP21)
283 (Supplementary Table A.2). Regarding the antibiotic classes, the most abundant ARG
284 confer resistance to peptide antibiotics (15%), followed by triclosans (13%) and

285 aminoglycosides (11%) (Fig. 1). The antibiotic classes with only a few ARG detected
286 include β -lactams (4 ARG detected), glycopeptides (4), chloramphenicol (3) and
287 streptogramin (1) (Fig. 1 and Supplementary Table A.3).

288 The human microbiome has been considered as a large reservoir of known and new
289 ARG, including some genes that can be genetically distant from known genes (Rolain,
290 2013; Rolain et al., 2016). Antibiotic production and resistance development are natural
291 mechanisms that provide evolutionary advantages to microorganisms in competitive
292 environments, including the human gut (Rolain, 2013). Just like the gut microbiota, the
293 respective resistome is highly variable, depending on the eating habits and environment
294 exposure. Consequently, strains will have a diverse ARG pattern depending on the
295 source where they were isolated from (Baron et al., 2018). The ARG most frequently
296 reported in the human gut microbiome are those related with resistance to tetracycline,
297 β -lactams, aminoglycosides, glycopeptides, chloramphenicol and macrolides (Baron et
298 al., 2018). Consistently, all of these classes could also be found in *F. prausnitzii*
299 genomes: 41 genomes possess one or more ARG putatively conferring resistance to
300 aminoglycosides; 37 possess one or more tetracycline resistance genes and 19 possess
301 macrolide resistance genes (Supplementary Table A.3). Although less represented, β -
302 lactams (4), chloramphenicol (3) and glycopeptides (2) resistance genes could also be
303 found in a small number of genomes (Supplementary Table A.3), which might indicate
304 that they do not belong to the usual resistome of *F. prausnitzii* species. In this context, it
305 is important to distinguish between intrinsic and acquired ARG. The first is usually
306 encoded within the chromosome and includes non-specific mechanisms that might
307 mediate the resistance phenotypes against several classes of antibiotics, such as non-
308 specific efflux pumps and inactivating enzymes (Peterson and Kaur, 2018). Conversely,
309 acquired ARG are usually associated with horizontal gene transfer (HGT) mechanisms

310 through mobile elements, and involve mechanisms of resistance that specifically target a
311 single antibiotic or class of antibiotics, including enzymes that modify the antibiotic or
312 the antibiotic target (Peterson and Kaur, 2018). Most ARG identified within
313 *F. prausnitzii* genomes are associated with general targets within the species or
314 unspecific mechanisms of resistance (Table 2). Hence, it can be presumed that they
315 consist mainly of intrinsic mechanisms of resistance, associated with the species, rather
316 than resistance genes acquired by HGT.

317 The high incidence of peptide antibiotics resistance genes can be explained by their
318 wide structural and functional diversity, which is most likely due to their ubiquitous
319 production in all domains of life – including prokaryotes, animals and plants –, and as
320 well as being an immune-modulatory component of the human innate defence system in
321 aiding pathogen clearance (Abdi et al., 2019; Mahlapuu et al., 2016). Thus, it is
322 expectable that commensal bacteria develop their own resistance mechanisms, to adapt
323 and successfully establish within the hosts' organism, which otherwise could be harmful
324 to them (Baron et al., 2018). Notably, the mechanisms involved in the resistance to this
325 kind of antibiotics are general resistance mechanisms, such as non-specific efflux
326 pumps or membrane alterations (Table 2). This means that they are not specifically
327 targeting a given compound, which, as mentioned, constitutes an indication of an
328 intrinsic resistance rather than an acquired trait (Peterson and Kaur, 2018).

329 The present *in silico* data lacks phenotypic confirmation, since phenotypic data
330 regarding this species is not available, except for the type-strain *F. prausnitzii* A2-165
331 (= DSM 17677) and a few other isolates (this study and references Foditsch et al.
332 (2014); Martín et al. (2017)).

333

334 **3.3. ARG and putative mobile elements within *F. prausnitzii* DSM 17677 genome**

335 *In silico* analysis is becoming a major tool in antibiotic resistance surveillance, allowing
336 the prediction of the potential resistance profile of a given strain. Indeed, EFSA-
337 FEEDAP guidelines encourage the use of this approach to characterize possible
338 resistance profiles of new candidate strains that are to be applied for human
339 consumption (EFSA-FEEDAP et al., 2018). In this way, alongside the phenotypic
340 characterization, *F. prausnitzii* DSM 17677 genome was also analysed in more detail
341 using PATRIC (Antonopoulos et al., 2019). A total of 24 ARG hits were obtained in
342 PATRIC, including genes that might mediate resistance against aminoglycosides,
343 macrolides, tetracycline and lincosamides (Supplementary Table B.1). Thus, there is
344 genomic evidence supporting the phenotypic observation of resistance to gentamycin,
345 kanamycin and streptomycin (aminoglycosides) and erythromycin (belonging to
346 macrolide class) (Table 1). No phenotypic evidence of resistance was observed for
347 tetracycline and clindamycin (from lincosamides class). Notice that the presence of a
348 specific ARG in a bacterial strain does not necessarily translate into the expression of a
349 resistant phenotype, for various reasons: *i*) the gene itself can be truncated; *ii*) it might
350 not be expressed in the growth conditions; *iii*) the product of a specific gene might only
351 provide a low level of resistance to the strain, thus, not reaching the cut-off values
352 defined by EFSA-FEEDAP to classify it as a resistant strain (EFSA-FEEDAP et al.,
353 2018). Additionally, the whole principle of ARG search and annotation relies on
354 homology-based searches, therefore, it might not be possible to ensure that the product
355 of the detected ARG sequence retains the function of the original query sequence
356 (Brinkac et al., 2017). This variability might also explain the inconsistency observed for
357 erythromycin, between broth microdilution and E-test® methods. Erythromycin related
358 ARG are associated with non-specific resistance mechanisms, and thus, they might be
359 differentially expressed under the test conditions for each method.

360 In addition, ResFinder, which allows the detection of acquired ARG within genomes,
361 using a defined threshold of 98.00% identity, was also used to detect putative ARG.
362 Thus, it detects whole resistance genes, although not providing information regarding
363 their functional integrity and expression (Zankari et al., 2013). From the annotated
364 genes retrieved by PATRIC, only aminoglycoside-encoded resistance was also
365 confirmed by ResFinder (Supplementary Table B.1). This might indicate that resistance
366 to streptomycin could have been acquired via HGT.

367 Interestingly, ARG putatively conferring resistance to β -lactams, which could explain
368 ampicillin-resistance phenotype, were not identified within *F. prausnitzii* DSM 17677
369 genome, using neither PATRIC nor ResFinder (Supplementary Table B.1). To
370 understand if other resistance genes are present within this genome, an homology-based
371 search was performed in CARD database, using Resistance Gene Identifier (RGI) tool
372 (Alcock et al., 2019). CARD contains a large, continuously curated, database of
373 molecular sequence reference data for prediction of AR genotype from genomic data.
374 This includes intrinsic resistance, dedicated resistance genes, and acquisition of
375 resistance via mutation of antimicrobial targets and associated elements (Jia et al.,
376 2017). The newly added RGI software predicts ARG from genomes, including non-
377 annotated sequences, using CARD-deposited ARG as query sequences (Alcock et al.,
378 2019). The RGI software analyses genome sequences under three paradigms: “Perfect”,
379 “Strict”, and “Loose”. The “Perfect” algorithm detects ARG with an exact (100%)
380 match to a CARD reference sequence and it is most often applied to clinical
381 surveillance. The “Strict” algorithm is more flexible, allowing some variation from the
382 CARD reference sequence, as long as the sequence falls within the curated BLAST cut-
383 off, to ensure that the detected variant is likely a functional ARG. This algorithm is
384 useful to detect previously unknown variants of ARG or antibiotic targets altered via

385 mutation. Finally, the “Loose” algorithm works outside the detection model cut-off,
386 thus allowing the detection of new, emergent threats and more distant homologs of
387 ARG, as well as homologous partial sequences that may not have a role in AR. “Loose”
388 algorithm, also known as “Discovery”, when combined with phenotypic screening
389 allows the discovery of novel ARG (Alcock et al., 2019; Jia et al., 2017). Although no
390 “Perfect” or “Strict” matches could be identified, a total of 263 “Loose” hits were
391 retrieved by RGI software upon homology-based search within *F. prausnitzii* DSM
392 17677 genome (Supplementary Table B.2). The absence of “Perfect” or “Strict” hits is
393 not surprising, considering the novelty of the antibiotic resistance profiling in this
394 species. As postulated, the “Loose” algorithm allowed the detection of otherwise
395 unnoticed genes, which might represent new variants of previously known resistance
396 genes. The “Loose” matches comprise the ARG previously identified in PATRIC and
397 other new hits, which include 25 ARG related with β -lactams resistance (Supplementary
398 Table B.2, highlighted). The majority of these β -lactam-related ARG are involved in
399 general resistance mechanisms, such as unspecific efflux pumps or antibiotic target
400 alteration, while only 9 encode putative β -lactamases, responsible for inactivating the
401 antibiotic itself (Supplementary Table B.2) (Docquier and Mangani, 2018). Together
402 with the phenotypic results, it is possible to infer that one, or more, of these mechanisms
403 are active and thus, confer resistance to ampicillin. The presence of more than one
404 mechanism and the possibility of their differential expression, might also explain the
405 heterogeneous resistance phenotype observed in the E-test® method for ampicillin, with
406 the occurrence of isolated colonies within the defined inhibition zone. As β -lactam
407 resistance does not seem to be a characteristic spread among the *F. prausnitzii* species,
408 the emergence of this phenotype has to be carefully considered and further analysed, as
409 it might constitute an acquired trait.

410 Although ARG *per se* might be worrisome under the lights of the worldwide spread
411 antibiotic resistance problematic, the HGT of such ARG constitutes a major issue when
412 trying to control the spreading across bacterial species (McInnes et al., 2020; Sun et al.,
413 2019). In fact, HGT occurrence was already documented between commensal and
414 opportunistic bacteria in human gut. The gut's high bacterial cell density provides an
415 environment favourable to conjugation events, which, in turn, promotes the spreading of
416 plasmids and integrative and conjugative elements (ICE) carrying ARG (McInnes et al.,
417 2020). Indeed, EFSA-FEEDAP guidelines consider acquired resistance of major
418 concern when evaluating the potential application of a novel probiotic for human
419 consumption (EFSA-FEEDAP et al., 2018). No plasmids were reported so far for this
420 strain, thus, excluding that vector as a source of mobilizable ARG. To understand the
421 ARG transferability potential in *F. prausnitzii* DSM 176777, ICEfinder tool from
422 IceBerg 2.0 software, was used to predict putative ICE regions within its genome. ICE
423 regions usually contain recombination and conjugation modules, and might even
424 mobilize other genetic elements encoded nearby such as chromosome-borne integrative
425 and mobilizable elements (IME) (Liu et al., 2019). Seven putative integrative regions
426 were identified within *F. prausnitzii* DSM 176777 genome (Supplementary Table B.3;
427 Fig. 2). Of particular interest is the presence of a β -lactamase-encoding gene embedded
428 in region 2 (IME) and the annotated ANT(6)-Ib, putatively conferring resistance to the
429 aminoglycoside streptomycin, within region 6 (ICE) (Supplementary Table B.3; Fig. 2).
430 While the first might require additional elements to be mobilized, the second might pose
431 some risk of genetic transferability and should be further analysed. This also confirms
432 what was observed using ResFinder, regarding the possible streptomycin acquired
433 resistance.

434 In parallel, the presence of putative GI was also assessed using IslandViewer (Bertelli et
435 al., 2017). Genomic islands are defined as clusters of consecutive genes, differing
436 between closely related strains, likely acquired by HGT in bacterial genomes. This
437 designation refers to genome regions that might be mobile, non-mobile or might have
438 lost their mobility through their evolution. In the first case, several types of mobile
439 genetic elements might be present, which include ICE, among others (Juhas et al.,
440 2009). Genomic Islands usually provide an advantage in a selective environment in
441 terms of bacterial adaptation and fitness. Depending on their origin and evolution, GI
442 usually contain a variation on the GC content when compared with the average GC
443 content of the chromosome where they are included. Also, GI usually include mobility
444 genes and insertion sequences that promote GI integration in the “host’s” core
445 chromosome (Bertelli et al., 2019, 2017; da Silva Filho et al., 2018). Thus,
446 IslandViewer 4.0 webserver integrates complementary GI prediction tools, which are
447 based in the nucleotide bias, the presence of mobility genes and comparative genomics,
448 to provide accurate and integrated GI predictions (Bertelli et al., 2017). IslandViewer
449 predicted a total of 69 GI, some of them overlapping each other (Fig. 2A;
450 Supplementary Table B.4). Notice the number of GI that is embedded into the
451 previously predicted ICE regions, thus, confirming a certain past and/or present
452 mobility of these sequences: they were putatively acquired from other genome(s) and
453 have the potential to be laterally transferred to other bacteria, if ideal conditions are
454 provided. Another evidence of the lateral acquisition of the GI is presented in Fig. 2B,
455 where the GI (orange bars on the in the bottom) coincide with regions of lower GC
456 content, compared with the average GC content of the chromosome. Putative
457 acquired/mobilizable genomic regions can also be transferred due to the action of
458 bacteriophages. Also known as phages, they play a major role in HGT mediation, since

459 they can be found in almost every microbial community (Casjens, 2003; Chiang et al.,
460 2019). Thus, bacteriophages can affect the bacterial gene content, contributing to the
461 genome plasticity through the acquisition and selection of genetic information that
462 might confer some evolutionary advantage, including virulence factors or ARG
463 (Ramisetty and Sudhakari, 2019). A prophage is defined as the latent form of a
464 bacteriophage, in which the viral genes are integrated in the bacterial cell – either in the
465 chromosome or an extrachromosomal plasmid – without causing the disruption of the
466 cell (Fortier and Sekulovic, 2013). If the prophage remains intact, it can be induced, due
467 to environmental stressors or randomly, to excise and replicate, thus undergoing the
468 lytic cycle (Chiang et al., 2019; Ramisetty and Sudhakari, 2019). During this process,
469 the newly formed viral particle might also include genes from the host, thus
470 contributing to HGT of between bacterial genomes (Casjens, 2003; Loh et al., 2020).
471 The occurrence of possible prophage sequences was also assessed in the present work,
472 using several web tools, starting with Prophage Hunter (Song et al., 2019). This online
473 platform uses a machine learning approach that allows the identification of sequences
474 putatively encoding prophages. It also provides a scoring method for categorizing each
475 prophage as “active”, “ambiguous” or “inactive”. Prophages identified as “active”
476 exhibit the complete genomic sequence and thus can be induced to produce phage
477 particles, while “ambiguous” prophages designate sequences that are somehow
478 incomplete, including mutated or truncated genes (Loh et al., 2020; Song et al., 2019).
479 As such, from a total of 81 prophage sequences candidate that were identified, 7 of them
480 were classified as “active” prophages (Supplementary Table B.5). A similar analysis
481 using PHASTER web tool, which is an integrated search and annotation tool to identify
482 and annotate putative prophage gene clusters, as well as to evaluate the completeness of
483 these putative prophages (Zhou et al., 2011). Six putative prophage-like regions were

484 identified, that were already identified using Prophage Hunter (Supplementary Table
485 B.6). Although 3 of them are classified as “active” by Prophage Hunter, only one is
486 expected to be complete according with PHASTER (Fig. 1 and Supplementary Table
487 B.6). A closer analysis revealed the absence of putative ARG embedded in this
488 prophage region. It is important to stress, that a prophage that is classified as “active”
489 or “intact” might not be prone to produce phage particles, as they can be false positives
490 – homology search does not ensure functionality – or they might lack the proper
491 induction factor (Loh et al., 2020). However, even if this event does occur, the absence
492 of embedded ARG reduces the possibility of antimicrobial resistance spreading to other
493 species within the environment.

494 Altogether, it is possible to conclude that ARG conferring resistance to β -lactams are
495 most likely acquired (Supplementary Table B.7). Also, some of the identified putative
496 β -lactamases have the potential to be transferred to other bacteria, as they are encoded
497 within putative ICE (Supplementary Table B.7) and should be regarded with caution in
498 further applications. This is consistent with the low number of *F. prausnitzii* strains that
499 possess β -lactams-related ARG (see previous section). The multiple ARG that can be
500 involved in this resistance phenotype, might also explain the existence of a
501 subpopulation with heterogeneous resistance phenotype in the E-test® method.

502 However, the highest concern is related with the resistance to streptomycin, mediated by
503 a specific mechanism of antibiotic inactivation. In addition, there is a low prevalence of
504 this specific ARG in other *F. prausnitzii* strains, as shown in the previous section. This
505 was also confirmed by a BLAST search against all deposited *F. prausnitzii* genomes
506 (taxid: 853), using either the DNA or the protein sequences, obtained from CARD as
507 query, which retrieved only two hits (data not shown), one of them corresponding to *F.*
508 *prausnitzii* A2-165 (= DSM 17677), as expected. In addition, this streptomycin-related

509 ARG belongs to an ICE and a GI. Associated with the phenotypic outcome of resistance
510 to streptomycin, this poses the only considerable risk to a potential application of *F.*
511 *prausnitzii* DSM 17677 as a probiotic strain, according to EFSA guidelines. However, it
512 is worth mentioning that resistance to streptomycin has already been reported for
513 probiotic candidates *Lactobacillus paragasseri* K7 and *Bifidobacterium animalis* subsp.
514 *lactis* BI-04 (Rozman et al., 2020), as well as in other gut-derived *Bifidobacterium*
515 strains, which belong to species already classified as safe by the qualified presumption
516 of safety list (Duranti et al., 2017).

517 Besides the presence of MGE, the transferability of chromosomal regions containing
518 ARG also depends on physical conditions, such as the occurrence of high bacterial cell
519 density or selective pressure, due to the presence of a specific antibiotic. In fact, it is
520 very unlikely that, without antibiotic selective pressure, ARG will be acquired and,
521 specially, fixed, in a given population, since these processes come with a fitness cost for
522 the acquiring-bacteria (Martínez, 2012). Although the human gut can provide the
523 required proximity between commensal/probiotic strains, possessing putative ARG, and
524 pathogenic bacteria, the transferability of these ARG highly depends on the antibiotic
525 pressure selectivity. Regarding streptomycin, this antibiotic is poorly absorbed by the
526 gastrointestinal tract and usually administered by intramuscular injections (except for
527 the treatment gastrointestinal infections). Also, due to its secondary effects,
528 streptomycin is being gradually replaced for other aminoglycosides with less toxic
529 effects for the patients (Grosset and Singer, 2013). Thus, it seems reasonable to expect
530 that the presence of streptomycin resistance gene encoded within MGE does not pose a
531 significant risk of ARG transferability, given the low probability of this antibiotic to
532 induce this phenomenon.

533

534 **4. Conclusions**

535 Our study provides novel and important data regarding the phenotypic and genotypic
536 AR profiles of probiotic candidate *F. prausnitzii* DSM 17677. Such information is a
537 paramount requirement for acceptance of this bacterial strain as a probiotic, allowing its
538 incorporation in nutraceutical and pharmaceutical products. Despite minor
539 inconsistencies in the results obtained by broth microdilution and E-test® methods, a
540 phenotypic AR profile was established, which could also be directly linked to the
541 genomic context. Furthermore, other ARG were detected that did not result in a
542 resistance phenotype. Nevertheless, most of detected ARG are related to general
543 resistance mechanisms and are not included within GI or MGE that allied with the
544 absence of plasmids, reduce the AR transferability risk. The only potential concern
545 identified relates with the streptomycin-specific ARG, which is encoded within an MGE
546 and belongs to a GI, which increases the likelihoods of acquired resistance. To meet the
547 criteria of the qualified presumption of safety status that requires the absence of
548 acquired ARG, future research on the transferability and the related mechanisms is
549 urgently needed to clarify the effective risk of the presence of streptomycin ARG in *F.*
550 *prausnitzii* DSM 17677. Furthermore, the establishment of a standardized protocol for
551 phenotypic AR profiling would be relevant to enable comparative studies among
552 researchers in food microbiology. Also, further studies involving other antimicrobials
553 clinically relevant for humans mentioned in other important guidelines such as CLSI
554 and EUCAST, together with the employment of quality control strains during
555 phenotypic antimicrobial susceptibility testing, would help deeper safety assessment of
556 *F. prausnitzii* DSM 17677 prior its use in food/feed products.

557

558 **Funding**

559 This work was supported by national funds through FCT/MEC (PIDDAC), project
560 references IF/00588/2015, under the Scientific Employment Stimulus - Individual Call
561 (CEEC Individual) -CEECIND/00520/2017/ CP1404/CT0001, and by Operational
562 Program Competitiveness and Internationalization in its FEDER component and by the
563 budget of the Foundation for Science and Technology, I.P. (FCT, IP) in its OE
564 component, project reference POCI-01-0145-FEDER-031400-PTDC/BAA-
565 AGR/31400/2017. We would also like to thank the scientific collaboration under the
566 FCT project UIDB/50016/2020.

567

568 **Declarations of interest**

569 None.

570

571 **References**

572 Abdi, M., Mirkalantari, S., Amirmozafari, N., 2019. Bacterial resistance to
573 antimicrobial peptides. *J. Pept. Sci.* 25, e3210. <https://doi.org/10.1002/psc.3210>

574 Alcock, B.P., Raphenya, A.R., Lau, T.T., Tsang, K.K., Bouchard, M., Edalatmand, A.,
575 Huynh, W., Nguyen, A.-L. V, Cheng, A.A., Liu, S., Min, S.Y., Miroshnichenko, A.,
576 Tran, H.-K., Werfalli, R.E., Nasir, J.A., Oloni, M., Speicher, D.J., Florescu, A., Singh,
577 B., Faltyn, M., Hernandez-Koutoucheva, A., Sharma, A.N., Bordeleau, E., Pawlowski,
578 A.C., Zubyk, H.L., Dooley, D., Griffiths, E., Maguire, F., Winsor, G.L., Beiko, R.G.,
579 Brinkman, F.S., Hsiao, W.W., Domselaar, G. V, McArthur, A.G., 2019. CARD 2020:
580 antibiotic resistome surveillance with the comprehensive antibiotic resistance database.
581 *Nucleic Acids Res.* 48, D517–D525. <https://doi.org/10.1093/nar/gkz935>

582 Almeida, D., Machado, D., Andrade, J.C., Mendo, S., Gomes, A.M., Freitas, A.C.,
583 2020. Evolving trends in next-generation probiotics: a 5W1H perspective. *Crit. Rev.*

584 Food Sci. Nutr. 60, 1783–1796. <https://doi.org/10.1080/10408398.2019.1599812>

585 Aminov, R., 2017. History of antimicrobial drug discovery: Major classes and health
586 impact. *Biochem. Pharmacol.* 133, 4–19. <https://doi.org/10.1016/j.bcp.2016.10.001>

587 Andersson, D.I., Nicoloff, H., Hjort, K., 2019. Mechanisms and clinical relevance of
588 bacterial heteroresistance. *Nat. Rev. Microbiol.* 17, 479–496.
589 <https://doi.org/10.1038/s41579-019-0218-1>

590 Andrade, J.C., Almeida, D., Domingos, M., Seabra, C.L., Machado, D., Freitas, A.C.,
591 Gomes, A.M., 2020. Commensal Obligate Anaerobic Bacteria and Health: Production,
592 Storage, and Delivery Strategies. *Front. Bioeng. Biotechnol.* 8, 550.
593 <https://doi.org/10.3389/fbioe.2020.00550>

594 Anisimova, E.A., Yarullina, D.R., 2019. Antibiotic Resistance of *LACTOBACILLUS*
595 Strains. *Curr. Microbiol.* 76, 1407–1416. <https://doi.org/10.1007/s00284-019-01769-7>

596 Anjum, M.F., Zankari, E., Hasman, H., 2017. Molecular Methods for Detection of
597 Antimicrobial Resistance, in: *Antimicrobial Resistance in Bacteria from Livestock and*
598 *Companion Animals*. American Society of Microbiology, pp. 33–50.
599 <https://doi.org/10.1128/microbiolspec.arba-0011-2017>

600 Antonopoulos, D.A., Assaf, R., Aziz, R.K., Brettin, T., Bun, C., Conrad, N., Davis, J.J.,
601 Dietrich, E.M., Disz, T., Gerdes, S., Kenyon, R.W., Machi, D., Mao, C., Murphy-Olson,
602 D.E., Nordberg, E.K., Olsen, G.J., Olson, R., Overbeek, R., Parrello, B., Pusch, G.D.,
603 Santerre, J., Shukla, M., Stevens, R.L., VanOeffelen, M., Vonstein, V., Warren, A.S.,
604 Wattam, A.R., Xia, F., Yoo, H., 2019. PATRIC as a unique resource for studying
605 antimicrobial resistance. *Brief. Bioinform.* 20, 1094–1102.
606 <https://doi.org/10.1093/bib/bbx083>

607 Arndt, D., Grant, J.R., Marcu, A., Sajed, T., Pon, A., Liang, Y., Wishart, D.S., 2016.
608 PHASTER: a better, faster version of the PHAST phage search tool. *Nucleic Acids Res.*

609 44, W16–W21. <https://doi.org/10.1093/nar/gkw387>

610 Aziz, R.K., Bartels, D., Best, A., DeJongh, M., Disz, T., Edwards, R.A., Formsma, K.,
611 Gerdes, S., Glass, E.M., Kubal, M., Meyer, F., Olsen, G.J., Olson, R., Osterman, A.L.,
612 Overbeek, R.A., McNeil, L.K., Paarmann, D., Paczian, T., Parrello, B., Pusch, G.D.,
613 Reich, C., Stevens, R., Vassieva, O., Vonstein, V., Wilke, A., Zagnitko, O., 2008. The
614 RAST Server: Rapid annotations using subsystems technology. *BMC Genomics* 9, 75.
615 <https://doi.org/10.1186/1471-2164-9-75>

616 Bag, S., Ghosh, T.S., Das, B., 2017. Complete genome sequence of *Faecalibacterium*
617 *prausnitzii* isolated from the gut of a healthy Indian adult. *Genome Announc.* 5,
618 e01286-17. <https://doi.org/10.1128/genomeA.01286-17>

619 Band, V.I., Weiss, D.S., 2019. Heteroresistance: A cause of unexplained antibiotic
620 treatment failure? *PLoS Pathog.* 15, e1007726.
621 <https://doi.org/10.1371/journal.ppat.1007726>

622 Baron, S.A., Diene, S.M., Rolain, J.M., 2018. Human microbiomes and antibiotic
623 resistance. *Hum. Microbiome J.* <https://doi.org/10.1016/j.humic.2018.08.005>

624 Benson, D.A., Cavanaugh, M., Clark, K., Karsch-Mizrachi, I., Ostell, J., Pruitt, K.D.,
625 Sayers, E.W., 2018. GenBank. *Nucleic Acids Res.* 46, D41–D47.
626 <https://doi.org/10.1093/nar/gkx1094>

627 Bertelli, C., Laird, M.R., Williams, K.P., Lau, B.Y., Hoad, G., Winsor, G.L., Brinkman,
628 F.S.L., 2017. IslandViewer 4: Expanded prediction of genomic islands for larger-scale
629 datasets. *Nucleic Acids Res.* 45, W30–W35. <https://doi.org/10.1093/nar/gkx343>

630 Bertelli, C., Tilley, K.E., Brinkman, F.S.L., 2019. Microbial genomic island discovery,
631 visualization and analysis. *Brief. Bioinform.* <https://doi.org/10.1093/bib/bby042>

632 Bortolaia, V., Kaas, R.S., Ruppe, E., Roberts, M.C., Schwarz, S., Cattoir, V., Philippon,
633 A., Allesoe, R.L., Rebelo, A.R., Florensa, A.F., Fagelhauer, L., Chakraborty, T.,

634 Neumann, B., Werner, G., Bender, J.K., Stingl, K., Nguyen, M., Coppens, J., Xavier,
635 B.B., Malhotra-Kumar, S., Westh, H., Pinholt, M., Anjum, M.F., Duggett, N.A., Kempf,
636 I., Nykäsenoja, S., Olkkola, S., Wiczorek, K., Amaro, A., Clemente, L., Mossong, J.,
637 Losch, S., Ragimbeau, C., Lund, O., Aarestrup, F.M., 2020. ResFinder 4.0 for
638 predictions of phenotypes from genotypes. *J. Antimicrob. Chemother.* 75, 3491–3500.
639 <https://doi.org/10.1093/jac/dkaa345>

640 Brinkac, L., Voorhies, A., Gomez, A., Nelson, K.E., 2017. The Threat of Antimicrobial
641 Resistance on the Human Microbiome. *Microb. Ecol.* 74, 1001–1008.
642 <https://doi.org/10.1007/s00248-017-0985-z>

643 Campedelli, I., Mathur, H., Salvetti, E., Clarke, S., Rea, M.C., Torriani, S., Ross, R.P.,
644 Hill, C., O'Toole, P.W., 2019. Genus-wide assessment of antibiotic resistance in
645 *Lactobacillus* spp. *Appl. Environ. Microbiol.* 85, e01738-18.
646 <https://doi.org/10.1128/AEM.01738-18>

647 Carver, T., Harris, S.R., Berriman, M., Parkhill, J., McQuillan, J.A., 2012. Artemis: An
648 integrated platform for visualization and analysis of high-throughput sequence-based
649 experimental data. *Bioinformatics* 28, 464–469.
650 <https://doi.org/10.1093/bioinformatics/btr703>

651 Casjens, S., 2003. Prophages and bacterial genomics: What have we learned so far?
652 *Mol. Microbiol.* 49, 277–300. <https://doi.org/10.1046/j.1365-2958.2003.03580.x>

653 Chiang, Y.N., Penadés, J.R., Chen, J., 2019. Genetic transduction by phages and
654 chromosomal islands: The new and noncanonical. *PLoS Pathog.* 15.
655 <https://doi.org/10.1371/journal.ppat.1007878>

656 Clinical and Laboratory Standards Institute, 2020. Performance Standards for
657 Antimicrobial Susceptibility Testing, 30th ed., CLSI supplement M100. Clinical and
658 Laboratory Standards Institute, Wayne, PA, USA.

659 Courvalin, P., 1991. Genotypic approach to the study of bacterial resistance to
660 antibiotics. *Antimicrob. Agents Chemother.* 35, 1019–1023.
661 <https://doi.org/10.1128/AAC.35.6.1019>

662 da Silva Filho, A.C., Raittz, R.T., Guizelini, D., De Pierri, C.R., Augusto, D.W., dos
663 Santos-Weiss, I.C.R., Marchaukoski, J.N., 2018. Comparative Analysis of Genomic
664 Island Prediction Tools. *Front. Genet.* 9, 619. <https://doi.org/10.3389/fgene.2018.00619>

665 Davis, J.J., Wattam, A.R., Aziz, R.K., Brettin, T., Butler, R., Butler, R.M., Chlenski, P.,
666 Conrad, N., Dickerman, A., Dietrich, E.M., Gabbard, J.L., Gerdes, S., Guard, A.,
667 Kenyon, R.W., MacHi, D., Mao, C., Murphy-Olson, D., Nguyen, M., Nordberg, E.K.,
668 Olsen, G.J., Olson, R.D., Overbeek, J.C., Overbeek, R., Parrello, B., Pusch, G.D.,
669 Shukla, M., Thomas, C., Vanoeffelen, M., Vonstein, V., Warren, A.S., Xia, F., Xie, D.,
670 Yoo, H., Stevens, R., 2020. The PATRIC Bioinformatics Resource Center: Expanding
671 data and analysis capabilities. *Nucleic Acids Res.* 48, D606–D612.
672 <https://doi.org/10.1093/nar/gkz943>

673 De Vecchi, E., Nicola, L., Larosa, M., Drago, L., 2008. *In vitro* activity of telithromycin
674 against *Haemophilus influenzae* at epithelial lining fluid concentrations. *BMC*
675 *Microbiol.* 8, 23. <https://doi.org/10.1186/1471-2180-8-23>

676 Docquier, J.-D., Mangani, S., 2018. An update on β -lactamase inhibitor discovery and
677 development. *Drug Resist. Updat.* 36, 13–29. <https://doi.org/10.1016/j.drug.2017.11.002>

678 DSMZ, 2020. *Faecalibacterium prausnitzii* DSM17677. DSMZ. Braunschweig. URL
679 <https://www.dsmz.de/collection/catalogue/details/culture/DSM-17677> (accessed
680 11.12.20).

681 Duranti, S., Lugli, G.A., Mancabelli, L., Turrone, F., Milani, C., Mangifesta, M.,
682 Ferrario, C., Anzalone, R., Viappiani, A., van Sinderen, D., Ventura, M., 2017.
683 Prevalence of antibiotic resistance genes among human gut-derived Bifidobacteria.

684 Appl. Environ. Microbiol. 83, e02894-16. <https://doi.org/10.1128/AEM.02894-16>

685 EFSA-FEEDAP, Rychen, G., Aquilina, G., Azimonti, G., Bampidis, V., Bastos, M. de
686 L., Bories, G., Chesson, A., Cocconcelli, P.S., Flachowsky, G., Gropp, J., Kolar, B.,
687 Kouba, M., López-Alonso, M., López Puente, S., Mantovani, A., Mayo, B., Ramos, F.,
688 Saarela, M., Villa, R.E., Wallace, R.J., Wester, P., Glandorf, B., Herman, L.,
689 Kärenlampi, S., Aguilera, J., Anguita, M., Brozzi, R., Galobart, J., 2018. Guidance on
690 the characterisation of microorganisms used as feed additives or as production
691 organisms. EFSA J. 16, e05206. <https://doi.org/10.2903/j.efsa.2018.5206>

692 El-Halfawy, O.M., Valvano, M.A., 2015. Antimicrobial heteroresistance: An emerging
693 field in need of clarity. Clin. Microbiol. Rev. 28, 191–207.
694 <https://doi.org/10.1128/CMR.00058-14>

695 Foditsch, C., Santos, T.M.A., Teixeira, A.G.V., Pereira, R.V.V., Dias, J.M., Gaeta, N.,
696 Bicalho, R.C., 2014. Isolation and characterization of *Faecalibacterium prausnitzii*
697 from calves and piglets. PLoS One 9, e116465.
698 <https://doi.org/10.1371/journal.pone.0116465>

699 Fortier, L.C., Sekulovic, O., 2013. Importance of prophages to evolution and virulence
700 of bacterial pathogens. Virulence 4, 354–365. <https://doi.org/10.4161/viru.24498>

701 Grosset, J.H., Singer, T., 2013. Streptomycin. Brenner's Encycl. Genet. Second Ed. 4,
702 568–569. <https://doi.org/10.1016/B978-0-12-374984-0.01484-4>

703 Gueimonde, M., Sánchez, B., de los Reyes-Gavilán, C.G., Margolles, A., 2013.
704 Antibiotic resistance in probiotic bacteria. Front. Microbiol. 4, 202.
705 <https://doi.org/10.3389/fmicb.2013.00202>

706 Haeili, M., Kafshdouz, M., Pishnian, Z., Feizabadi, M.M., Martinez-Martinez, L., 2019.
707 Comparison of susceptibility testing methods for determining the activity of colistin
708 against Gram-negative bacilli of clinical origin. J. Med. Microbiol. 68, 60–66.

709 <https://doi.org/10.1099/jmm.0.000879>

710 Hakvoort, H., Bovenkamp, E., Greenwood-Quaintance, K.E., Schmidt-Malan, S.M.,
711 Mandrekar, J.N., Schuetz, A.N., Patel, R., 2020. Imipenem-Relebactam Susceptibility
712 Testing of Gram-Negative Bacilli by Agar Dilution, Disk Diffusion, and Gradient Strip
713 Methods Compared with Broth Microdilution. *J. Clin. Microbiol.* 58, e00695-20.
714 <https://doi.org/10.1128/JCM.00695-20>

715 Hernando-Amado, S., Coque, T.M., Baquero, F., Martínez, J.L., 2019. Defining and
716 combating antibiotic resistance from One Health and Global Health perspectives. *Nat.*
717 *Microbiol.* 4, 1432–1442. <https://doi.org/10.1038/s41564-019-0503-9>

718 Humphries, R.M., Ambler, J., Mitchell, S.L., Castanheira, M., Dingle, T., Hindler, J.A.,
719 Koeth, L., Sei, K., 2018. CLSI methods development and standardization working
720 group best practices for evaluation of antimicrobial susceptibility tests. *J. Clin.*
721 *Microbiol.* 56, e01934-17. <https://doi.org/10.1128/JCM.01934-17>

722 Jia, B., Raphenya, A.R., Alcock, B., Waglechner, N., Guo, P., Tsang, K.K., Lago, B.A.,
723 Dave, B.M., Pereira, S., Sharma, A.N., Doshi, S., Courtot, M., Lo, R., Williams, L.E.,
724 Frye, J.G., Elsayegh, T., Sardar, D., Westman, E.L., Pawlowski, A.C., Johnson, T.A.,
725 Brinkman, F.S.L., Wright, G.D., McArthur, A.G., 2017. CARD 2017: expansion and
726 model-centric curation of the comprehensive antibiotic resistance database. *Nucleic*
727 *Acids Res.* 45, D566–D573. <https://doi.org/10.1093/nar/gkw1004>

728 Juhas, M., van der Meer, J.R., Gaillard, M., Harding, R.M., Hood, D.W., Crook, D.W.,
729 2009. Genomic islands: Tools of bacterial horizontal gene transfer and evolution. *FEMS*
730 *Microbiol. Rev.* 33, 376–393. <https://doi.org/10.1111/j.1574-6976.2008.00136.x>

731 Kulengowski, B., Ribes, J.A., Burgess, D.S., 2019. Polymyxin B Etest[®] compared with
732 gold-standard broth microdilution in carbapenem-resistant *Enterobacteriaceae*
733 exhibiting a wide range of polymyxin B MICs. *Clin. Microbiol. Infect.* 25, 92–95.

734 <https://doi.org/10.1016/j.cmi.2018.04.008>

735 Liu, M., Li, X., Xie, Y., Bi, D., Sun, J., Li, J., Tai, C., Deng, Z., Ou, H.Y., 2019.

736 ICEberg 2.0: An updated database of bacterial integrative and conjugative elements.

737 Nucleic Acids Res. 47, D660–D665. <https://doi.org/10.1093/nar/gky1123>

738 Loh, B., Chen, J., Manohar, P., Yu, Y., Hua, X., Leptihn, S., 2020. A Biological

739 Inventory of Prophages in *A. baumannii* Genomes Reveal Distinct Distributions in

740 Classes, Length and Genomic Positions. Front. Microbiol. 11, 2020.10.26.355222.

741 <https://doi.org/10.1101/2020.10.26.355222>

742 Loose, M., Naber, K.G., Coates, A., Wagenlehner, F.M.E., Hu, Y., 2020. Effect of

743 different media on the bactericidal activity of Colistin and on the synergistic

744 combination with Azidothymidine against *mcr-I*-positive colistin-resistant *Escherichia*

745 *coli*. Front. Microbiol. 11, 54. <https://doi.org/10.3389/fmicb.2020.00054>

746 Mahlapuu, M., Håkansson, J., Ringstad, L., Björn, C., 2016. Antimicrobial peptides: An

747 emerging category of therapeutic agents. Front. Cell. Infect. Microbiol. 6, 194.

748 <https://doi.org/10.3389/fcimb.2016.00194>

749 Maier, E., Anderson, R.C., Roy, N.C., 2017. Live *Faecalibacterium prausnitzii* does not

750 enhance epithelial barrier integrity in an apical anaerobic co-culture model of the large

751 intestine. Nutrients 9, 1349. <https://doi.org/10.3390/nu9121349>

752 Martín, R., Chain, F., Miquel, S., Lu, J., Gratadoux, J.-J., Sokol, H., Verdu, E.F.,

753 Bercik, P., Bermúdez-Humarán, L.G., Langella, P., 2014. The Commensal Bacterium

754 *Faecalibacterium prausnitzii* Is Protective in DNBS-induced Chronic Moderate and

755 Severe Colitis Models. Inflamm. Bowel Dis. 20, 417–430.

756 <https://doi.org/10.1097/01.MIB.0000440815.76627.64>

757 Martín, R., Miquel, S., Benevides, L., Bridonneau, C., Robert, V., Hudault, S., Chain,

758 F., Berteau, O., Azevedo, V., Chatel, J.M., Sokol, H., Bermúdez-Humarán, L.G.,

759 Thomas, M., Langella, P., 2017. Functional characterization of novel *Faecalibacterium*
760 *prausnitzii* strains isolated from healthy volunteers: A step forward in the use of *F.*
761 *prausnitzii* as a next-generation probiotic. *Front. Microbiol.* 8, 1226.
762 <https://doi.org/10.3389/fmicb.2017.01226>

763 Martínez, J.L., 2012. Bottlenecks in the Transferability of Antibiotic Resistance from
764 Natural Ecosystems to Human Bacterial Pathogens. *Front. Microbiol.* 2, 265.
765 <https://doi.org/10.3389/fmicb.2011.00265>

766 Matuschek, E., Åhman, J., Webster, C., Kahlmeter, G., 2018. Antimicrobial
767 susceptibility testing of colistin – evaluation of seven commercial MIC products against
768 standard broth microdilution for *Escherichia coli*, *Klebsiella pneumoniae*, *Pseudomonas*
769 *aeruginosa*, and *Acinetobacter* spp. *Clin. Microbiol. Infect.* 24, 865–870.
770 <https://doi.org/10.1016/j.cmi.2017.11.020>

771 McInnes, R.S., McCallum, G.E., Lamberte, L.E., van Schaik, W., 2020. Horizontal
772 transfer of antibiotic resistance genes in the human gut microbiome. *Curr. Opin.*
773 *Microbiol.* 53, 35–43. <https://doi.org/10.1016/j.mib.2020.02.002>

774 Neut, C., Mahieux, S., Dubreuil, L.J., 2017. Antibiotic susceptibility of probiotic
775 strains: Is it reasonable to combine probiotics with antibiotics? *Med. Mal. Infect.* 47,
776 477–483. <https://doi.org/10.1016/j.medmal.2017.07.001>

777 Peterson, E., Kaur, P., 2018. Antibiotic resistance mechanisms in bacteria:
778 Relationships between resistance determinants of antibiotic producers, environmental
779 bacteria, and clinical pathogens. *Front. Microbiol.* 9, 2928.
780 <https://doi.org/10.3389/fmicb.2018.02928>

781 Pruitt, K.D., Tatusova, T., Brown, G.R., Maglott, D.R., 2012. NCBI Reference
782 Sequences (RefSeq): Current status, new features and genome annotation policy.
783 *Nucleic Acids Res.* 40, D130–D135. <https://doi.org/10.1093/nar/gkr1079>

784 Ramisetty, B.C.M., Sudhakari, P.A., 2019. Bacterial “grounded” prophages: Hotspots
785 for genetic renovation and innovation. *Front. Genet.* 10, 12.
786 <https://doi.org/10.3389/fgene.2019.00065>

787 Rolain, J.M., 2013. Food and human gut as reservoirs of transferable antibiotic
788 resistance encoding genes. *Front. Microbiol.* 4, 173.
789 <https://doi.org/10.3389/fmicb.2013.00173>

790 Rolain, J.M., Abat, C., Jimeno, M.T., Fournier, P.E., Raoult, D., 2016. Do we need new
791 antibiotics? *Clin. Microbiol. Infect.* 22, 408–415.
792 <https://doi.org/10.1016/j.cmi.2016.03.012>

793 Rozman, V., Mohar Lorbeg, P., Accetto, T., Bogovič Matijašić, B., 2020.
794 Characterization of antimicrobial resistance in lactobacilli and bifidobacteria used as
795 probiotics or starter cultures based on integration of phenotypic and *in silico* data. *Int. J.*
796 *Food Microbiol.* 314, 108388. <https://doi.org/10.1016/j.ijfoodmicro.2019.108388>

797 Selvin, J., Maity, D., Sajayan, A., Kiran, G.S., 2020. Revealing antibiotic resistance in
798 therapeutic and dietary probiotic supplements. *J. Glob. Antimicrob. Resist.* 22, 202–
799 205. <https://doi.org/10.1016/j.jgar.2020.02.007>

800 Sirichoat, A., Flórez, A.B., Vázquez, L., Buppasiri, P., Panya, M., Lulitanond, V.,
801 Mayo, B., 2020. Antibiotic susceptibility profiles of lactic acid bacteria from the human
802 vagina and genetic basis of acquired resistances. *Int. J. Mol. Sci.* 21, 2594.
803 <https://doi.org/10.3390/ijms21072594>

804 Sokol, H., Pigneur, B., Watterlot, L., Lakhdari, O., Bermúdez-Humarán, L.G.,
805 Gratadoux, J.-J., Blugeon, S., Bridonneau, C., Furet, J.-P., Corthier, G., Grangette, C.,
806 Vasquez, N., Pochart, P., Trugnan, G., Thomas, G., Blottière, H.M., Doré, J., Marteau,
807 P., Seksik, P., Langella, P., 2008. *Faecalibacterium prausnitzii* is an anti-inflammatory
808 commensal bacterium identified by gut microbiota analysis of Crohn disease patients.

809 Proc. Natl. Acad. Sci. 105, 16731–16736. <https://doi.org/10.1073/PNAS.0804812105>

810 Song, W., Sun, H.X., Zhang, C., Cheng, L., Peng, Y., Deng, Z., Wang, D., Wang, Y.,
811 Hu, M., Liu, W., Yang, H., Shen, Y., Li, J., You, L., Xiao, M., 2019. Prophage Hunter:
812 an integrative hunting tool for active prophages. *Nucleic Acids Res.* 47, W74–W80.
813 <https://doi.org/10.1093/nar/gkz380>

814 Sun, D., Jeannot, K., Xiao, Y., Knapp, C.W., 2019. Editorial: Horizontal Gene Transfer
815 Mediated Bacterial Antibiotic Resistance. *Front. Microbiol.* 10, 1933.
816 <https://doi.org/10.3389/fmicb.2019.01933>

817 Wattam, A.R., Abraham, D., Dalay, O., Disz, T.L., Driscoll, T., Gabbard, J.L.,
818 Gillespie, J.J., Gough, R., Hix, D., Kenyon, R., MacHi, D., Mao, C., Nordberg, E.K.,
819 Olson, R., Overbeek, R., Pusch, G.D., Shukla, M., Schulman, J., Stevens, R.L.,
820 Sullivan, D.E., Vonstein, V., Warren, A., Will, R., Wilson, M.J.C., Yoo, H.S., Zhang,
821 C., Zhang, Y., Sobral, B.W., 2014. PATRIC, the bacterial bioinformatics database and
822 analysis resource. *Nucleic Acids Res.* 42, 581–591. <https://doi.org/10.1093/nar/gkt1099>

823 Wiegand, I., Hilpert, K., Hancock, R.E.W., 2008. Agar and broth dilution methods to
824 determine the minimal inhibitory concentration (MIC) of antimicrobial substances. *Nat.*
825 *Protoc.* 3, 163–175. <https://doi.org/10.1038/nprot.2007.521>

826 World Health Organization, 2014. Antimicrobial Resistance Global Report on
827 Surveillance. Geneva.

828 Zankari, E., Hasman, H., Kaas, R.S., Seyfarth, A.M., Agersø, Y., Lund, O., Larsen,
829 M.V., Aarestrup, F.M., 2013. Genotyping using whole-genome sequencing is a realistic
830 alternative to surveillance based on phenotypic antimicrobial susceptibility testing. *J.*
831 *Antimicrob. Chemother.* 68, 771–777. <https://doi.org/10.1093/jac/dks496>

832 Zhou, Y., Liang, Y., Lynch, K.H., Dennis, J.J., Wishart, D.S., 2011. PHAST: A Fast
833 Phage Search Tool. *Nucleic Acids Res.* 39, W347–W352.

834 <https://doi.org/10.1093/nar/gkr485>

835

836 **Figure Captions**

837 Fig. 1. Percentage of antibiotic resistance genes detected in 104 *Faecalibacterium*
838 *prausnitzii* genomes, divided by antibiotic classes.

839

840 Fig. 2. *Faecalibacterium prausnitzii* DSM 17677 genome map. (A) represents the whole
841 chromosome, with ICE (in blue), GI (in red) and Prophage-like region (in green),
842 highlighted in the external circles; possible ARG, identified by CARD are represented
843 inside the chromosome circle: in blue, ARG confirmed with PATRIC/ResFinder; in
844 orange, CARD ARG encoded with an ICE and an GI; in purple, ARG encoded only
845 within the ICE. (B) GI (orange bars) and Prophage-like region (green bar)
846 representation and GC content across the chromosome, with colour coded scheme, to
847 ease the correlation. Images produced in SnapGene Viewer.

848

849 Table 1. Antimicrobial susceptibility profile of *Faecalibacterium prausnitzii* DSM 17677

Antimicrobial	MIC (µg/mL) in broth microdilution	MIC (µg/mL) in E-test®	EFSA-FEEDAP cut-off values (µg/mL)
Ampicillin	> 2	> 3	1
Vancomycin	≤ 1	0.25 - 0.5	4
Gentamicin	> 8	12	4
Kanamycin	> 32	16 - 32	16
Streptomycin	> 16	≥1024	8
Erythromycin	> 2	0.5-1	1
Clindamycin	≤ 1	≤ 0.016	4
Tetracycline	≤ 0.5	0.023-0.047	2
Chloramphenicol	2-4	1-1.5	4

850 Note: Resistances are indicated in bold i.e., MIC exceeding EFSA-FEEDAP cut-off

851 values.

852

853

854

855

856 Table 2. Putative ARG detected within 104 *F. prausnitzii* genomes, divided by their mode of action and antibiotic classes. In **bold** are represented
 857 mechanisms that are specific for a certain antibiotic.

Mode of Action	CARD definition	# Hits	Antibiotics	# Hits
Antibiotic target in susceptible species	Antibiotic-sensitive wild-type bacterial components; might suffer mutations that render them not susceptible	1597	Quinolones	212
			Cycloserine	208
			Fosfomycin	207
			Fusidic acid	128
			Mupirocin	111
			Rifamycin/peptide antibiotics	110
			Fosmidomycin	108
			Peptide antibiotics	107
			Triclosan	105
			Tetracyclines/glycylcyclines	101
Elfamycins	99			

Mode of Action	CARD definition	# Hits	Antibiotics	# Hits
			Aminoglycosides	97
			Sulfonamides	1
			Other	3
Protein altering cell wall charge conferring antibiotic resistance	Cell wall alteration	245	Peptide antibiotics	245
Gene conferring resistance via absence	Deletion of gene or gene product results in resistance. For example, deletion of a porin gene blocks drug from entering the cell.	107	Aminoglycosides	107
Antibiotic inactivation enzyme	Enzyme that catalyses the inactivation of an antibiotic resulting in resistance. Inactivation includes chemical modification, destruction, etc.	80	Aminoglycosides	51
			Nitroimidazoles	19
			Lincosamides	2
			Penams/cephalosporins	2
			Chloramphenicol	3

Mode of Action	CARD definition	# Hits	Antibiotics	# Hits
			β -lactams	2
			Streptogramins	1
Efflux pump conferring antibiotic resistance	Subunits of efflux proteins that pump antibiotic out of a cell to confer resistance.	17	Tetracyclines	16
			Macrolides	1
Protein involved in antibiotic sequestration	A gene whose product inactivates an antibiotic by formation of a complex, preventing interaction of the antibiotic with its target.	101	Triclosan	101
Antibiotic target replacement protein	Alternate proteins that have the same functions as other antibiotic target proteins but are structurally different and thus resistant to antibiotics.	99	Triclosan	99

Mode of Action	CARD definition	# Hits	Antibiotics	# Hits
Antibiotic target protection protein	These proteins confer antibiotic resistance by bind the antibiotic target to prevent antibiotic binding.	47	Tetracyclines	47
Antibiotic target modifying enzyme	Enzymes that confer resistance by modifying (mutational alteration or enzymatic modification) antibiotic targets.	26	Lincosamides/streptogramins/macrolides	26
Regulator modulating expression of antibiotic resistance genes	Mechanism activated by the presence of a specific antibiotic.	5	Glycopeptides	5

859 **Supplementary Information**

860 File: SupInfo A contains:

861 Table A.1. List of ARG detected in all deposited *F. prausnitzii* genomes, using
862 PATRIC. Includes the gene designation (when available), function (when available),
863 gene product and the mechanism of resistance mediated by that product. Notice that the
864 availability of these data depends on the annotation data provided at the time of the
865 genome deposit in the database.

866

867 Table A.2. Number of ARG in each *F. prausnitzii* strain, calculated from
868 Supplementary Table A.1. The average number of ARG/strain is calculated at the
869 bottom of the list; the maximum and minimum ARG-containing genomes are
870 highlighted.

871

872 Table A.3. Number of ARG group by gene product and antibiotic classes to which they
873 might confer resistance. See also Fig. 1.

874

875 File: SupInfo B contains:

876 Table B.1. *Faecalibacterium prausnitzii* DSM 17677 ARG retrieved by PATRIC. This
877 table includes the position of each ARG within the genome, the gene designation and its
878 product, the type mechanism possibly involved in conferring resistance, and the
879 antibiotic class and antibiotics to which they might confer resistance. The identification
880 of the same gene by CARD and ResFinder is also indicated. Genes encoded within
881 putative integrative regions are highlighted.

882

883 Table B.2. *Faecalibacterium prausnitzii* DSM 17677 ARG identified by CARD RGI
884 tool. This table includes the cut-off used for the identification, the position of each ARG
885 within the genome, the gene family, the type mechanism possibly involved in
886 conferring resistance, and the antibiotic class to which they might confer resistance. The
887 identification of the same gene by PATRIC and ResFinder is also indicated. Genes
888 encoding β -lactams resistance-mediating products are highlighted.

889

890 Table B.3. List of putative integrative regions detected by ICEfinder. This table includes
891 the position of each region and the CARD RGI hits encoded within each region with the
892 respective information, from Supplementary Table B.2. The GC percentage for each
893 region is also indicated. See also Fig. 1.

894

895 Table B.4. List of putative genomic islands predicted by IslandViewer. The position of
896 each island within the chromosome is indicated as well as genes that were considered
897 acquired; highlighted in yellow, the genes that were identified by CARD as ARG; the
898 integrative regions are also provided, to facilitate an integrated observation
899 (Supplementary Table B.3). See also Fig. 2.

900

901 Table B.5. List of putative prophage-like regions predicted by Prophage Hunter. The
902 position of each region within the chromosome is indicated as well as the closest
903 homologous phage; highlighted in yellow, prophages that are considered active by this
904 tool; hits that were also retrieved by PHASTER are indicated. See also Fig. 2.

905

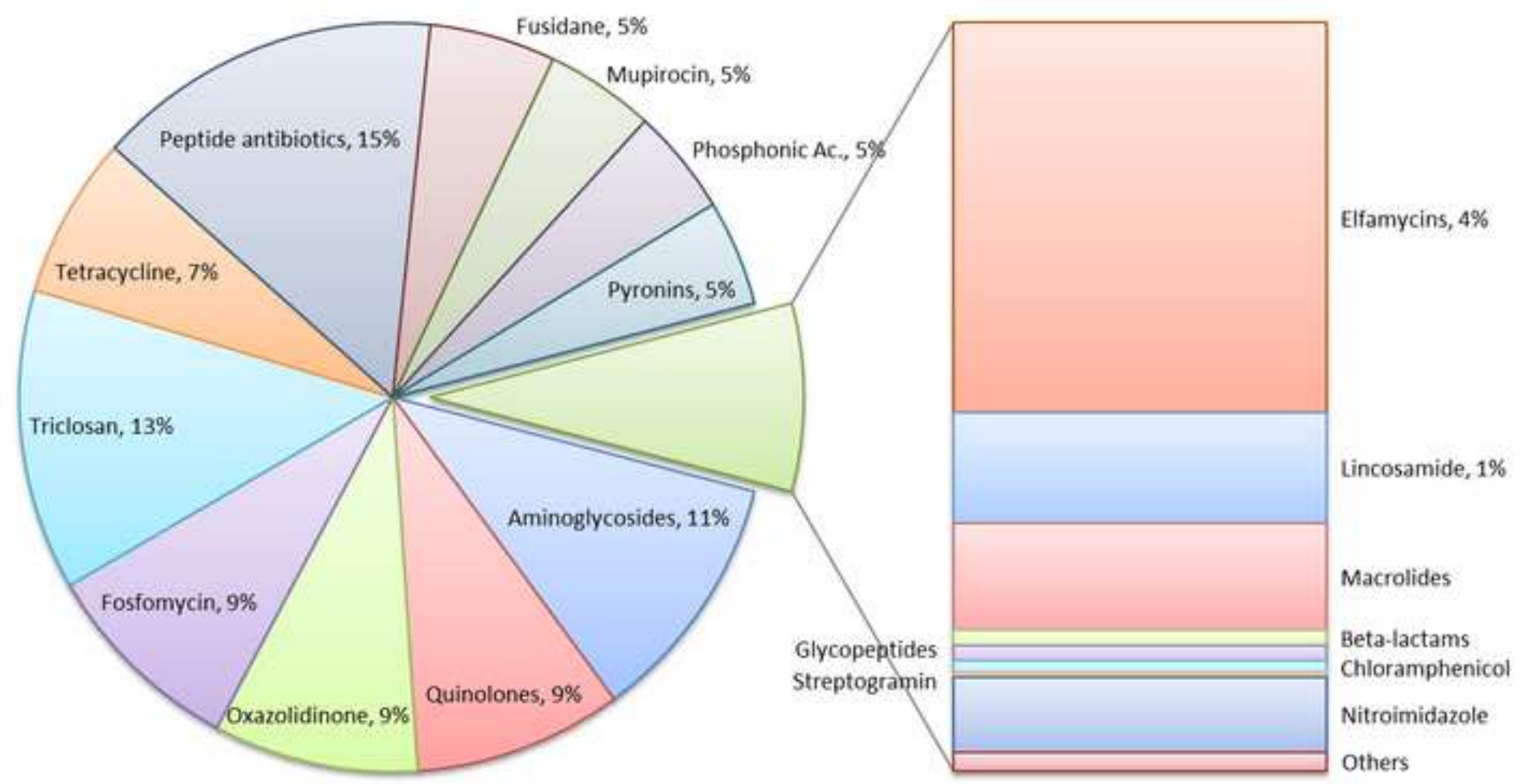
906 Table B.6. List of putative prophage-like regions predicted by PHASTER. The position
907 of each region within the chromosome is indicated as well as the corresponding

908 Prophage Hunter hits; highlighted in yellow and orange, prophages that are considered
909 active and ambiguous by Prophage Hunter, respectively; the integration integrative
910 regions and genomic islands are also provided, to facilitate an integrated observation
911 (Supplementary Table B.3 and Table B.4). Predicted CARD ARG that are embedded
912 within these prophage-like regions are also shown (Supplementary Table B.2). See also
913 Fig. 2.

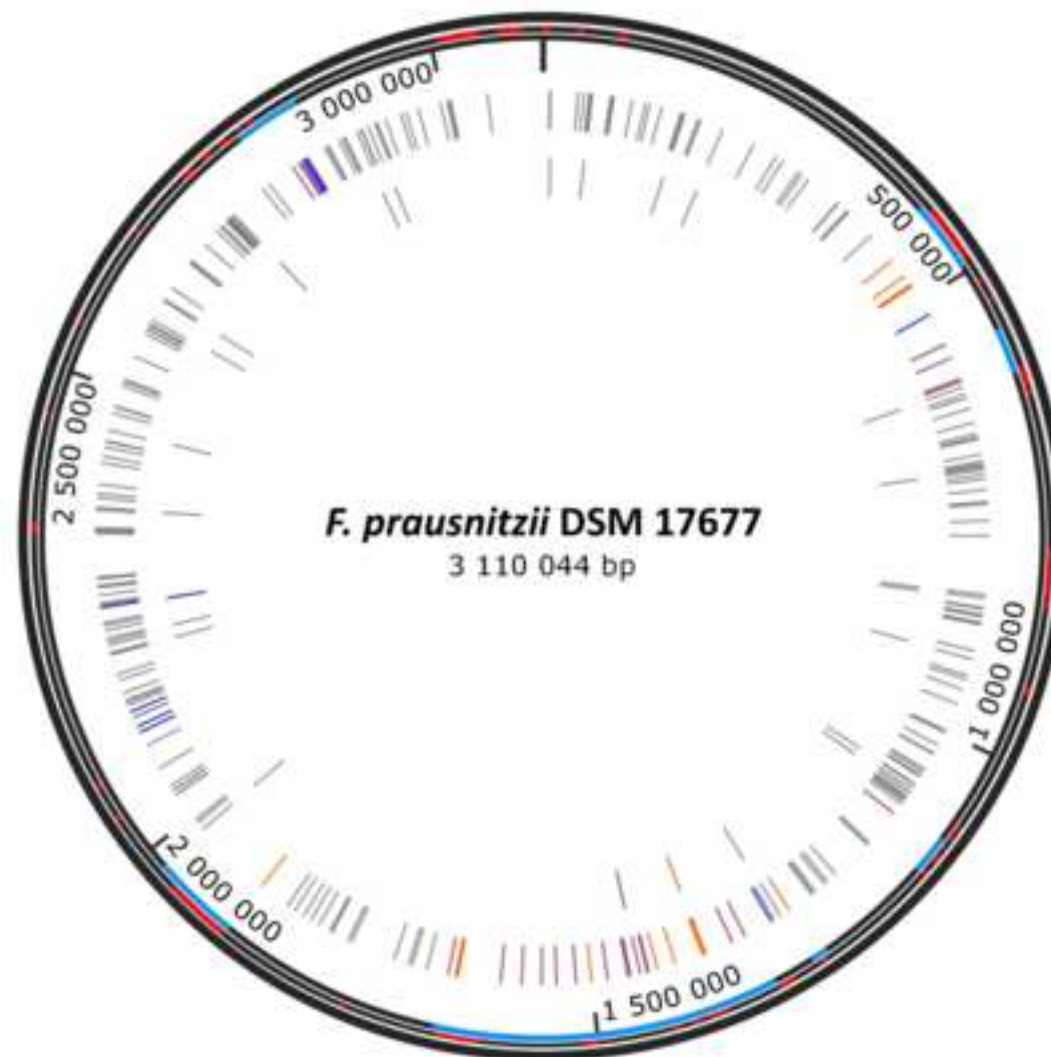
914

915 Table B.7. Summary table integrating the *in silico* data from Supplementary Table B.1
916 to Table B.4, with special emphasis on the most concerning ARG identified in this
917 study.

Figure 1



A



B

