

1 **Profiling grapevine trunk pathogens *in planta*: A case for**
2 **community-targeted DNA metabarcoding**

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21 **ABSTRACT**

22 DNA metabarcoding, commonly used in exploratory microbial ecology studies, is a promising
23 method for the simultaneous *in planta*-detection of multiple pathogens associated with disease
24 complexes, such as the grapevine trunk diseases. Their detection is particularly challenging, due
25 to the presence within an individual wood lesion of multiple co-infecting trunk pathogens and
26 other wood-colonizing fungi, which span a broad range of taxa in the Fungal Kingdom. As such,
27 we designed metabarcoding primers, using as template the ribosomal internal transcribed spacer
28 of grapevine trunk-associated Ascomycete fungi (GTAA) and compared them to two universal
29 primer widely used in microbial ecology. We first performed *in silico* simulations and then tested
30 the primers by high-throughput amplicon sequencing of (i) multiple combinations of mock
31 communities, (ii) time-course experiments with controlled inoculations, and (iii) diseased field
32 samples from vineyards under natural levels of infection. All analyses showed that GTAA had
33 greater affinity and sensitivity, compared to those of the universal primers. Importantly, with
34 GTAA, profiling of mock communities and comparisons with shotgun-sequencing metagenomics
35 of field samples gave an accurate representation of genera of important trunk pathogens, namely
36 *Phaeomoniella*, *Phaeoacremonium*, and *Eutypa*, the abundances of which were greatly over- or
37 under-estimated with universal primers. Overall, our findings not only demonstrate that DNA
38 metabarcoding gives qualitatively and quantitatively accurate results when applied to grapevine
39 trunk diseases, but also that primer customization and testing are crucial to ensure the validity of
40 DNA metabarcoding results.

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43 INTRODUCTION

44 Grapevine trunk diseases affect the longevity and productivity of grapevines (*Vitis vinifera*) in all
45 major growing regions of the world [1-4]. They are caused by numerous species of fungi that infect
46 and damage the wood, causing chronic infections [5-7]. Among the most common grapevine trunk
47 diseases are Eutypa dieback (primarily caused by *Eutypa lata*), Esca (primarily caused by
48 *Phaeoacremonium minimum*, *Phaeomoniella chlamydospora*, and *Fomitiporia* spp.),
49 Botryosphaeria dieback (primarily caused by *Neofusicoccum parvum*, *Diplodia seriata*, among
50 other fungi in the Botryosphaeriaceae family), Phomopsis dieback (primarily caused by *Diaporthe*
51 *ampelina*), and Black foot (caused by *Cylindrocarpon*, *Campylocarpon*, and *Ilyonectria* spp.) [4,
52 8-11]. Because of the characteristic mixed infections, trunk diseases represent a disease complex
53 [12, 13]. In addition to infections of pruning wounds by airborne and splash-dispersed spores,
54 trunk pathogens may be introduced to a healthy vineyard by asymptomatic propagation material.
55 Fungi associated with grapevine trunk diseases have been found in rootstock mother-plants, rooted
56 rootstock cuttings, bench-grafts, and young grafted vines [14-16]. The presence of multiple species
57 in the same vine complicates disease diagnosis and, consequently, proper timing of practices to
58 limit infection in the vineyard and to propagate clean nursery stock.

59 Taxonomic identification of fungi associated with grapevine wood is currently done by the
60 following steps: (i) plating grapevine woody tissue on nutrient-rich agar plates, (ii) hyphal-tip
61 colony isolation to pure cultures, (iii) DNA extraction from fungal mycelium, (iv) PCR
62 amplification of taxonomically informative loci, such as the nuclear ribosomal internal transcribed
63 spacer (ITS), elongation factor, and β -tubulin, and (v) comparisons of amplicon sequences with
64 sequence databases [17-19]. PCR-based diagnostics represent a significant improvement
65 compared to traditional approaches that depend on morphological features for species

66 identification and, thus, require skilled expertise in mycology [20]. However, these approaches
67 still require an initial culturing step, which may limit the detection of slow-growing fungi.
68 Alternatively, with species or genus-specific markers, PCR could be used to determine *in planta*
69 the presence of certain species, thereby skipping the culturing step [21, 22]. One limitation of this
70 approach, however, is that it may not detect all trunk pathogens in a given sample [23, 24]. Indeed,
71 certain combinations of fungi may be important in the severity of symptom expression [25].
72 Because trunk pathogens cause mixed infections, attempts have been made to characterize the
73 composition of the trunk-pathogen community. For example, finger-printing techniques like
74 Automated Ribosomal Intergenic Spacer Analysis (ARISA) [26] and Single-Strand Conformation
75 Polymorphism (SSCP) [27, 28] have been used to compare fungal communities among different
76 samples of grapevine wood, although these do not identify trunk pathogens to the species level. A
77 DNA macroarray system, based on reverse dot-blot hybridization containing oligonucleotides
78 complementary to portions of the β -tubulin locus, was developed for species-level identification,
79 specifically for detection of trunk pathogens that cause Young vine decline [23]. We previously
80 described a strategy, based on untargeted shotgun sequencing of metagenomic DNA and RNA, to
81 detect and quantify trunk pathogens *in planta* simultaneously [13]. Despite clear advantages over
82 other approaches, this method still has its limitations, such as relying on assembled genomes, as
83 well as costly library preparation and computationally intensive analyses.
84 DNA metabarcoding, which has been used extensively for the analysis of microbial communities
85 [29-32], may provide a cheaper and more scalable method for the characterization of trunk-
86 pathogen communities. This approach has already been applied to other pathosystems to address
87 a variety of research objectives. For example, DNA metabarcoding has been used to identify
88 candidate pathogens [33, 34] and potential biocontrol agents [35], to profile putative plant

89 pathogens associated with insects [36], and to diagnose quarantine pathogens as part of national
90 plant-protection programs [37-39]. DNA metabarcoding infers taxonomic composition of complex
91 biological samples by amplifying, sequencing, and analyzing target genomic regions [40, 41]. The
92 ribosomal ITS, which is under low evolutionary pressure and, thus, presents high levels of
93 variation between closely related species, has been commonly used as a barcode for the analysis
94 of fungal biodiversity [42, 43]. ITS is typically amplified by universal primers that anneal to the
95 conserved flanking sequences. The “universality” of the primers, which derives from their ability
96 to amplify a broad range of taxonomically unrelated species across the Fungal Kingdom [44], is
97 exploited in studies that aim to profile fungal communities, typically in exploratory analyses of
98 environmental samples. We hypothesized that although universal primers may capture broad
99 biodiversity in exploratory analyses, they may provide less accurate representation of microbial
100 pathogen communities than primers that are designed and optimized to amplify species known to
101 be associated with those communities, based on prior knowledge of disease etiology. After all,
102 grapevine-trunk diseases are one of the most widely studied disease complexes, in terms of species
103 composition (Lamichane and Venturi, 2015). In this work, we designed and evaluated
104 metabarcoding primers that were optimized to amplify the ITS regions of grapevine trunk
105 pathogens. By a combination of *in silico* simulations, and analyses of ‘mock’ communities,
106 samples from controlled inoculations, and samples from symptomatic vineyards, we demonstrated
107 that community-customized metabarcoding provides greater more qualitatively and quantitevely
108 accurate representation of trunk-pathogen communities than common universal primers.

109

110 **RESULTS**

111 **Primer design, selection, and validation with target species**

112 We designed multiple degenerate primers that target the internal transcribed spacer (ITS) of
113 grapevine trunk-associated ascomycetes (GTAA) using the TrunkDiseaseID as reference database
114 [20]. Primer potential was determined *in silico*, considering the amplicon size and estimating the
115 number of sequence hits to the database, their alignment mismatches, and gap scores. From a total
116 of twenty forward and three reverse degenerate primers, primers GTAA182f and GTAA526r
117 (GTAA, hereinafter) performed the best and were selected for further testing. The GTAA primers
118 target the entire ITS2 region with the forward and reverse primers aligning to the 5.8S ribosomal
119 RNA and the large subunit ribosomal ribonucleic acid (LSU), respectively (**Table 1** and **Figure**
120 **1A**). The primers produced amplicons of approximately 350 bp from isolates of seven trunk
121 pathogens, as expected based on the amplicon size predicted from the 213 ITS sequences of
122 ascomycetes in the TrunkDiseaseID database (301.72 ± 7.53 bp; **Figure 1B**). We obtained a
123 similar amplicon size when the GTAA primers were used to amplify total DNA extracted from
124 naturally infected grapevines with trunk-disease symptoms (**Figure 1C**). Failure to amplify DNA
125 of two negative controls, grape leaves, and of the bacterium *Agrobacterium tumefaciens*, supports
126 their specificity to Fungi (**Figure 1B**). Amplicon sequences matched the correct species, when
127 aligned to the NCBI non-redundant nucleotide database using BLASTn, thereby confirming the
128 ITS region amplified by the GTAA primers is informative for taxonomic assignments (**Additional**
129 **File 1**).

130 GTAA primer precision, sensitivity, efficiency, and usefulness for metabarcoding of grapevine
131 trunk pathogens were compared to those of the BITS [45] and SP primers [46]. The BITS primers
132 are widely used for fungal metabarcoding analysis in vineyards and grape must (e.g.: [47-51]),
133 whereas the SP primers [46] were recently used by [52] for fungal microbial ecology, and
134 implemented in the Earth Microbiome Project (<http://www.earthmicrobiome.org>; **Table 1** and

135 **Figure 1A**). Samples were from DNA extracted from potted grapevines either inoculated with *N.*
136 *parvum* or from non-inoculated controls. By sampling PCR reactions every five cycles, the GTAA
137 amplicon was visible on an agarose gel starting at 20 cycles, whereas those of SP and BITS were
138 visible at 25 and 30 cycles, respectively (**Figure 1D**). Furthermore, SP produced multiple bands,
139 which may be due to non-specific binding and/or chimeric amplicons. Based on qPCR with the
140 same samples, the average Ct values for GTAA were approximately nine cycles lower than those
141 of BITS ($P < 1.85e^{-04}$) and SP ($P < 3.50e^{-04}$) (**Figure 1E**). Overall, our findings suggest a higher
142 affinity of the GTAA primers, when amplifying samples containing grapevine trunk pathogens.

143 ***In silico* simulation of amplification and taxonomic identification**

144 We then carried out an *in silico* simulation that compared the potential amplification bias and
145 taxonomic usefulness of GTAA, BITS, and SP primers using a comprehensive dataset of fungi
146 associated with trunk diseases. We compiled a custom database of 521 full-length ITS sequences
147 across 17 genera (**Figure 2A, Additional File 2**). We included only full-length ITS sequences to
148 be able to compare primers that amplify different regions of the ITS (**Figure 1A**). *In silico*
149 amplification of each sequence in the custom database was carried out considering all alternative
150 sequences of degenerated primers and allowing a series of mismatches between primer and
151 template sequences. *In silico* amplification was carried out testing all possible combinations of
152 allowed mismatches, from 0 to 5 mismatches in the first five nucleotides of the primer (head) and
153 0 to 2 mismatches in the last two nucleotides of the primer (tail). GTAA primers amplified a
154 higher number of sequences than BITS and SP primers, for every parameter tested (**Figure 2B**).
155 When no mismatches between primer and target were allowed, GTAA primers amplified 85.80%,
156 SP primers amplified 13.63%, and BITS primers were predicted to amplify none of the sequences
157 in the database. When at least two mismatches were allowed in the tail of the primer, BITS and SP

158 primers amplified only 16.70% and 30.33% of target sequences, respectively, whereas GTAA
159 primers amplified 86.75%. With the most permissive parameters, GTAA primers amplified
160 98.08% of the sequences, and BITS and SP primers amplified 97.89% and 25.91%, respectively.
161 The requirement of multiple mismatches for BITS primers to achieve a similar number of
162 sequences as GTAA primers is consistent with the cycle-sampling results (**Figure 1D**), and
163 suggests that GTAA primers are more efficient than BITS at amplifying the ITS of grapevine trunk
164 pathogens.

165 To determine if amplicons generated by GTAA primers are informative for taxonomic assignment,
166 we analyzed with Mothur [53] the amplicons that were generated by the simulation. By comparing
167 the assigned genera (observed) with the expected genera for each primer set we assessed false
168 positive (FP; i.e, erroneously assigned), false negative (FN; i.e, not amplified or not assigned), and
169 true positive (TP; i.e, correctly assigned, **Figure 2C**) rates. GTAA primers had the highest
170 sensitivity ($TP/(TP+FN)*100 = 89.50 \pm 6.45\%$), followed by BITS ($54.25 \pm 47.86\%$), and SP
171 ($20.50 \pm 2.53\%$). SP and GTAA primers displayed similar precision (SP: $TP/(TP+FP)*100 = 97.50$
172 $\pm 1.00\%$; GTAA: $97.00 \pm 0.00\%$), which was higher than that of BITS primers ($72.25 \pm 48.18\%$).
173 The different performance of the three primer sets in the simulation appeared to be mostly due to
174 amplification bias against certain genera (**Figure 2D**). GTAA primers amplified and correctly
175 assigned to the proper genera a larger fraction of sequences than the other two primer sets for 14
176 out of 17 genera tested. This was the case for the following widely distributed trunk pathogens:
177 *Eutypa* (GTAA: $98.0 \pm 4.0\%$, BITS: $53.8 \pm 53.8\%$, and SP: $44.0 \pm 4.0\%$), *Diaporthe* (GTAA: 96.5
178 $\pm 1.3\%$, BITS: $51.3 \pm 53.6\%$, and SP: $18.7 \pm 9.3\%$), and *Phaeoacremonium* (GTAA: $95.5 \pm 3.0\%$,
179 BITS: $51.5 \pm 56.1\%$, and SP: $18.7 \pm 9.0\%$). BITS primers correctly assigned more sequences for
180 *Lasiodiplodia* (GTAA: $71.0 \pm 8.0\%$, BITS: $75.0 \pm 50.0\%$, and SP: $0.0 \pm 0.0\%$) and *Cylindrocarpon*

181 (GTAA: $35.5 \pm 41.0\%$, BITS: $46.3 \pm 33.54\%$, and SP: $0.0 \pm 0.0\%$). SP primers correctly assigned
182 more sequences for *Campylocarpon* (GTAA: $50.0 \pm 57.7\%$, BITS: $50.0 \pm 57.7\%$, and SP: $75.0 \pm$
183 50.0%). Overall, this simulation predicted that, unlike the two universal primer sets, GTAA
184 primers amplify ITS of more trunk pathogens and allow taxonomic assignment with greater
185 sensitivity (i.e., higher true positive rate) and specificity (i.e., lower false negative rate). SP primers
186 were not included in further experiments, due to their poor performance in these early stages.

187 **Analysis of mock communities and infection time course**

188 To evaluate the primers for characterizing the species composition of mixed infections, we first
189 started by sequencing with an Illumina MiSeq and analyzing mock communities (**Figure 3A**).
190 Although DNA was extracted from stems with no symptoms of trunk disease to be used as a pure
191 source of grape DNA, both primer sets detected fungi, mostly belonging to the genera
192 *Campylocarpon* and *Phaeoacremonium* (**Figure 3A**). When grape DNA was mixed with DNAs
193 of *Phaeoa. minimum* and *Phaeom. chlamydospora* both primer sets identified the correct taxa, with
194 small relative deviation from expected values (GTAA $\delta = 11.02 \pm 7.0\%$; BITS $\delta = 16.68 \pm$
195 11.39%). For mock communities including *Eutypa*, GTAA primers detected this trunk pathogen
196 in similar amounts to the expected abundance ($\delta = 9.74 \pm 1.10\%$), whereas BITS primers greatly
197 underestimated its abundance ($\delta = 88.87 \pm 1.27\%$). In mock communities with equal
198 concentrations of DNA from *E. lata*, *Phaeoa. minimum*, *Phaeom. chlamydospora*, *N. parvum*, *D.*
199 *seriata*, and *D. ampelina*, there was underrepresentation of *Eutypa* ($\delta = 16.70 \pm 0.12\%$), and
200 *Phaeoacremonium* ($\delta = 13.10 \pm 0.63\%$), and overrepresentation of *Phaeomoniella* ($\delta = 24.34 \pm$
201 1.56%) by BITS primers (**Figure 3A**). The correlation of expected and observed abundances in
202 these mock communities was greater for GTAA ($R = 0.92$) than BITS ($R = 0.67$; **Figure 3B**).
203 Because DNA was mixed in equal amounts, the expected relative abundance of each genus was

204 16.6%. GTAA primers detected *Eutypa* at $16.42 \pm 2.41\%$, whereas BITS primers detected this
205 trunk pathogen at $0.05 \pm 0.03\%$. In the case of *Diplodia*, GTAA primers estimated the abundance
206 of the genus at $3.13 \pm 0.48\%$ and BITS primers at $28.9 \pm 0.8\%$. Interestingly, neither primer set
207 was able to detect properly *Diaporthe*, reporting only $0.58 \pm 0.23\%$ and $0.87 \pm 0.15\%$ for GTAA
208 and BITS primers, respectively. Nonetheless, GTAA primers provide a better qualitative and
209 quantitative representation of important trunk pathogens.

210 We then tested the two primers using grape samples collected at different time points after
211 controlled inoculation with a trunk pathogen. The objective of this analysis was to determine if the
212 metabarcoding approach could detect quantitative differences between samples at early and late
213 stages of infection. Vines were inoculated with *N. parvum* and stem samples were collected at 24
214 hours, 2 weeks, and 6 weeks post-inoculation. Plants non-inoculated wounded (NIW) and non-
215 inoculated non-wounded (NINW) were included as controls. As expected, *Neofusicoccum* was
216 predominant in the inoculated wounded (IW) samples, but absent from the controls (**Figure 4**),
217 except for a single NIW sample, possibly due to cross-contamination during wounding or from
218 contamination of the propagation material. Both primer sets revealed a five-fold increase in the
219 average percentage of *Neofusicoccum* between 24 hours and 6 weeks post-inoculation.

220 **Analysis of field samples and comparison with reference-based shotgun metagenome** 221 **sequencing**

222 We then tested the primers on naturally infected grapevines. We used the same 28 field samples
223 described in [13], which allowed us to compare the metabarcoding approach with the quantitative
224 taxonomic profiles obtained by a reference-based shotgun metagenome sequencing. The samples
225 were grouped according to symptoms into *Eutypa* dieback (ED), Esca (ES), wood canker without
226 foliar symptoms (WC), and apparently-healthy (AH). All 28 field samples were amplified with

227 both GTAA and BITS primers, with SP primers used for a subset. Taxonomy assignment based
228 on amplicon metabarcoding detected 14 genera, in addition to those with genomes in the
229 multispecies reference, with abundances $> 0.05\%$ in one or more samples (**Additional File 3**).
230 Both GTAA and BITS primer sets identified *Alternaria*, *Cyphellophora*, and *Penicillium*, whereas
231 *Cladosporium*, *Aureobasidium*, *Gibberella*, and *Cryptovalsa* were only identified by GTAA
232 primers, and *Angustimassarina*, *Exophiala*, *Erysiphe*, *Meyerozyma*, *Acremonium*, and
233 *Vishniacozyma* by BITS primers. GTAA primers revealed species abundances at very similar
234 levels to those obtained by metagenomics analysis (**Figure 5A**), with a strong linear correlation
235 between the two approaches ($R = 0.95$; **Figure 5B**), which was higher than those of both BITS (R
236 $= 0.63$) and SP primers ($R = 0.27$). In agreement with the other results described above, BITS
237 primers underestimated *Eutypa* in *Eutypa*-dieback samples and overrepresented *Phaeoconiella* in
238 *Esca* samples. The even weaker correlation obtained with SP primers was due to the strong bias
239 against *Eutypa* and *Diaporthe*. GTAA primers showed stronger correlations across all genera of
240 trunk pathogens ($0.89 < R < 0.99$, **Figure 5C**) compared to those of BITS ($0.58 < R < 0.75$). Both
241 primer sets showed a low correlation for *Neofusicoccum*, likely due to the low abundance of this
242 genus in the samples assayed. Overall, our findings confirm the universal primers have a
243 significant bias against important taxa and were outperformed by our GTAA primers for trunk
244 pathogens.

245 **DISCUSSION**

246 In this study, we tested the application of DNA metabarcoding to profile the fungal taxa associated
247 with grapevine trunk diseases. We show that DNA metabarcoding of ribosomal ITS amplified with
248 commonly-adopted universal primers consistently misrepresented the abundance of important
249 trunk pathogen species, such as *Eutypa* and *Phaeoconiella*. The customization of primer design

250 using trunk pathogen sequences as template led to improved the results with greater sensitivity.
251 This was likely due to greater homology between the GTAA primers and the ITS of the grapevine
252 trunk pathogens they target. On average the sequence identity of the grapevine trunk pathogen
253 targets was significantly greater with the GTAA primers ($97.4 \pm 5.5\%$; $P < 2e-16$) than with the
254 other universal primers (BITS: $90.2 \pm 7.1\%$; SP: $83.3 \pm 0.2.2\%$) used in the study. Amplification
255 bias of universal ITS primers due to higher levels of mismatches for certain taxonomic group were
256 observed previously using *in silico* PCR [54, 55]. Importantly, we also showed that the GTAA
257 primers had higher sensitivity while maintaining a precision threshold for taxonomic assignment
258 of 97%, suggesting that the customization of the target region also played a role in improving the
259 DNA metabarcoding for these organisms. We should stress out that BITS and SP primers are not
260 the only available universal primers and the goal of this study was not to provide a comprehensive
261 survey of all universal ITS metabarcoding primers. BITS and SP were selected, because they are
262 both widely used DNA barcoding primers, including in studies conducted on vineyard and wine
263 must samples [46-51]. We cannot rule out that other universal primers that were not tested in this
264 study may have performed differently. However, the results presented in this study show that
265 universal primers may not be always appropriate to study a fungal community and, when fungal
266 community composition is available, researchers should consider customizing their DNA
267 metabarcoding primers. In addition, we illustrate the value of assessing both the amplification and
268 taxonomy usefulness of the metabarcoding primers *in silico* prior to downstream wet lab
269 evaluations.

270 In addition to customization of primers, the inclusion of other DNA barcodes should help
271 overcome some of the limitations associated with the ITS region, such as copy number variation
272 between and within species and low resolution in separating some phylogenetically closely related

273 fungal species [42, 56]. For example, the ITS region does not accurately identify species of plant-
274 pathogenic fungi like *Alternaria*, *Botryosphaeria*, and *Diaporthe* [57]. The genera for which the
275 GTAA primers consistently underestimated abundances like *Lasiodiplodia*, *Botryosphaeria*,
276 *Diplodia*, and *Diaporthe* are known to be difficult to be resolved with the ITS region alone [57].
277 The high correlation between metagenomics and metabarcoding results using the GTAA primers
278 suggest that copy number variation of the ITS region is not an overwhelming issue for the
279 grapevine trunk pathogens present in the field samples. Nonetheless, we expect that the inclusion
280 of additional barcodes, such as β -tubulin and elongation factor 1- α , will help increase accuracy of
281 taxonomic identification at the species level and help measure those genera for which the ITS is
282 known not to be effective [20, 23, 58, 59].

283 CONCLUSIONS

284 As trunk diseases are complex diseases caused by mixed infections, DNA metabarcoding should
285 provide a rapid and effective method for high-throughput multispecies identification overcoming
286 the limitations of currently applied diagnostic methods. Universal primers are advantageous in
287 exploratory analysis where *a priori* knowledge on the taxonomic composition of the samples is
288 limited or not available. However, a more targeted approach should be used when the objective is
289 to study a more defined group of microorganisms, like the grapevine trunk pathogens which
290 symptoms have been consistently associated with certain fungal species [4, 20, 60]. Overall, the
291 results presented here demonstrated that DNA metabarcoding can be applied to grapevine trunk
292 diseases. With further improvement of taxonomic identification by combining multiple barcoding
293 loci and of quantification by measurement of direct correlation between fungal biomass and PCR
294 amplification cycles, we envision DNA metabarcoding to be routinely applied in trunk pathogen
295 research and diagnostics. DNA metabarcoding provides multiple advantages to methods employed

296 in the past. Namely, there is no need of fungal isolation, it allows high number of samples to be
297 analyzed at the same time given the multiplexing potential of the technology, and takes advantage
298 of the constantly improving high-throughput sequencing technologies. Since wood pathogens may
299 remain asymptomatic in young, non-stressed vines, propagation material may contain latent fungal
300 infections and may become symptomatic after planting and serve as a source of inoculum for
301 further infections of potentially clean plants. Methods of virus detection and eradication have been
302 crucial in ensuring that the material in germplasm repositories and clean plant programs is free of
303 known viruses. By allowing the rapid testing of large number of wood samples from mother plants
304 in foundation blocks and propagation material in nurseries, we expect that the applications of
305 metabarcoding to trunk pathogen diagnostics will help reduce the amount of trunk pathogens
306 introduced into vineyards at planting as well as the incidence of young vine decline. Our results
307 also demonstrated that primer customization and testing are crucial to ensure the validity of DNA
308 metabarcoding results.

309

310 **METHODS**

311 **Metabarcoding primers targeting grapevine trunk-associated Ascomycetes (GTAA)**

312 Ribosomal Internal transcribed spacer (ITS) sequences of trunk pathogens and other wood-
313 colonizing fungi of grape, specifically in the Division Ascomycota, were retrieved from the
314 TrunkDiseaseID.org database [20]. Sequences were aligned using ClustalW2 (v2.1; [61]) to
315 identify conserved regions. Sequence alignment was used as input for the metabarcoding primer
316 design software Primer Prospector v1.0.1 [62], using a sensitivity threshold of 80% and an initial
317 primer seed size of 5 bp. The ITS sequence of *E. lata* (GeneBank KU721859.1) was used as a
318 ‘reference’. Primers were selected based on median amplicon size, and mismatches, gaps, and

319 numbers of matches to the sequences in the database. The base pairs ‘AG’ were used as a linker
320 between the primer and an eight-nucleotide barcodes on the 5’ region of the forward primer
321 sequence. Barcode sequences were as described in [63]. A list of barcoded forward GTAA primers
322 is listed in **Additional File 4**.

323 A custom database was compiled with full length ITS sequences of species in the following genera:
324 *Botryosphaeria*, *Diplodia*, *Dothiorella*, *Lasiodiplodia*, *Neofusicoccum*, *Phaeomoniella*,
325 *Diaporthe*, *Phaeoacremonium*, *Diatrype*, *Diatrypella*, *Eutypa*, *Xylaria*, *Cylindrocarpon*,
326 *Campylocarpon*, *Dactylonectria*, *Ilyonectria*, and *Neonectria*. Sequences were retrieved from the
327 NCBI GenBank repository. Completeness of the ITS sequences was validated using the hidden
328 Markov models-based software ITSx [64]. Only sequences spanning the entire ITS region (ITS1,
329 5.8S, and ITS2) were kept for downstream analysis. Species and GenBank accessions of the
330 complete ITS sequences included in the custom database are listed in **Additional File 2**. To reduce
331 redundancy and identify outliers, the complete ITS sequences were clustered using the UCLUST
332 algorithm [65] integrated in Qiime (v1.9.1; [66]) with 97% identity. The longest representative
333 sequence of each cluster was selected, using the Qiime ‘pick_rep_set.py’ function. All
334 representative sequences were aligned using Mafft v7.271 [67]) with the ‘--auto’ argument and
335 1,000 iterations. Sequences clustering outside the expected family were removed from the final
336 custom database.

337 The program Degenerate In-Silico PCR (dispr, <https://github.com/douglasgscfield/dispr>) was
338 used to predict and evaluate the amplification of sequences of the custom ITS database, using our
339 GTAA primers, and universal BITS [45], and SP [46] primers. Dispr allowed an amplicon size of
340 100 to 400 bp, all combinations up to five mismatches in the head of the primer (‘H’ or 5’-most
341 region), and all combinations up to two mismatches in the tail of the primer (‘T’ or the remaining

342 3'-portion of the primer). The resulting amplicons produced *in silico* were then used for taxonomy
343 assignment with 80% confidence, using Mothur (v1.39.5; [53]), as it is integrated in Qiime
344 (v1.9.1). The UNITE database v7.2 [68] was used as taxonomic reference. True positives were
345 defined as sequences that were assigned to the expected genus, false positives were sequences
346 assigned to a different genus, and false negatives were sequences not assigned to any genus or
347 were not amplified by disp.

348 To generate mock communities, we combined (i) DNA from a healthy grapevine with DNA from
349 pure cultures of three trunk pathogens at different concentrations, or (ii) equal concentrations of
350 DNA from pure cultures of six trunk pathogens. For the former, grape and fungal DNA were
351 combined as follows: 90% grape with 10% *E. lata* isolate Napa209 [69], *Phaeoa. minimum* isolate
352 1119 [70], or *Phaeom. chlamydospora* isolate C42 [71]; 80% grape with 10% of each of two fungal
353 isolates (in all three pair-wise combinations of the three isolates); and 70% grape DNA with 10%
354 of each of the three fungal isolates. For the latter, equal concentrations of DNA were combined
355 from the same three trunk pathogens and three additional species: *N. parvum* isolate UCD646So
356 [11], *Dia. ampelina* isolate Wolf911 [72], and *Diplodia seriata* isolate SBen831 [73]. Grape DNA
357 was extracted from the leaves of a non-inoculated, non-wounded plant; this DNA template came
358 from a previous experiment [74]. The mock communities containing grape DNA were amplified
359 and sequenced independently three times, whereas the mock community of six fungal DNAs was
360 amplified and sequenced independently five times. Prior to DNA extraction on Potato Dextrose
361 Agar (PDA; Difco laboratories, Detroit, MI). DNA was extracted as described in [73] and
362 measured by Qubit (Life technologies).

363 To test *in planta* detection of a trunk pathogen at variable levels of infection (i.e., from low to high
364 concentrations of fungal biomass over time), DNAs for the infection time course of *N. parvum*

365 were extracted from the same samples described in [74]. Briefly, 1-year-old potted *V. vinifera*
366 ‘Cabernet Sauvignon’ FPS 19 plants were inoculated with isolate UCD646So mycelia. Woody
367 stems were collected at seven time points: 0 hpi, 3 hpi, 24 hpi, 2 wpi, 6 wpi, 8 wpi, and 12 wpi.
368 Wood samples from 1 cm below the inoculation site were collected using flame-sterilized forceps
369 and immediately placed in liquid nitrogen for nucleic acids extraction. Infections were confirmed
370 by positive recovery of the pathogen after 5-day growth on PDA.

371 To test *in planta* detection of multiple trunk pathogens in mixed infection (i.e., to characterize the
372 species composition of a naturally established trunk-pathogen community), DNA from the same
373 28 field samples described in [13] was used to make cross-technology comparisons. These field
374 samples were collected from mature vines (> 8 years-old) showing a variety of the most common
375 symptoms associated with trunk diseases. Wood samples were collected from distinct plants with
376 the following combinations of symptoms: Eutypa dieback foliar and wood symptoms, Esca foliar
377 and wood symptoms, wood symptoms and no foliar symptoms, and apparently healthy plants with
378 no foliar or wood symptoms.

379 **High throughput sequencing libraries**

380 Each sample was amplified using the unique 8-nt barcode forward primer sets for GTAA and
381 BITS, to enable sample multiplexing. The 25- μ l PCR reaction mix contained 2 ng of DNA
382 template, 1X Colorless GoTaq flexi buffer (Promega Corporation, Madison, WI), 1.5 mM MgCl₂,
383 0.1 mg/ml BSA, 0.2 mM dNTPs, 0.4 μ M of each primer, and 1.25 units of GoTaq Flexi DNA
384 polymerase (Promega Corporation, Madison, WI). PCR program (Veriti thermal cycler, Applied
385 Biosystems) was as follows: initial denaturation at 94°C for 3 min, followed by 35 cycles at 94°C
386 for 45 seconds, 55°C for 1 min., and 72°C for 1 min., and a final extension at 72°C for 10 min. In
387 the experiments to assess primer affinity, reactions were stopped after 5, 10, 15, 20, 25, 30, and 35

388 cycles. Following PCR, amplicon size and uniqueness were verified using gel electrophoresis, and
389 bands were cleaned using Ampure XP magnetic beads (Agencourt, Beckman Coulter). DNA
390 concentration was determined for each purified amplicon using Qubit (Life technologies). For the
391 single isolate validation, amplicons were sequenced with Sanger (DNA Sequencing Facility,
392 University of California, Davis).

393 For high-throughput sequencing, equimolar amounts of all barcoded amplicons were pooled into
394 a single sample, the total concentration of which was determined by Qubit. Five hundred
395 nanograms of pooled DNA were then end-repaired, A-tailed and single-index adapter ligated
396 (Kapa LTP library prep kit, Kapa Biosystems). After adapter ligation, the sample was size-selected
397 with two consecutive 1X bead-based cleanups; concentration and size distribution were
398 determined with Qubit and Bioanalyzer (Agilent Technologies), respectively. DNA libraries were
399 submitted for sequencing in 250-bp paired-end mode on an Illumina MiSeq (UCDavis Genome
400 Center DNA technologies Core). All FASTQ files with the amplicon sequences separated by
401 barcode were deposited in the NCBI Sequence Read Archive (BioProject: PRJNA485180; SRA
402 accession: SRP156804).

403 **Amplicon sequencing community analysis**

404 Adapter-trimming was carried out using BBDuk (BBMap v.35.82; [http://jgi.doe.gov/data-and-](http://jgi.doe.gov/data-and-tools/bb-tools/)
405 [tools/bb-tools/](http://jgi.doe.gov/data-and-tools/bb-tools/)) in paired-end mode with sequence “AGATCGGAAG” and the following
406 parameters: ktrim=r, k=10, mink=6, edist=2, ordered=t, qtrim=f and minlen=150. Adapter-
407 trimmed FASTQ files were then quality-filtered using Trimmomatic v0.36 [75] with paired-end
408 mode, phred33, a sliding windows of 4:19, and a minimum length of 150 bp. Sequencing data
409 were then processed in the Qiime environment v1.9.1 [66]. Barcodes were extracted from the
410 FASTQ files using the “extract_barcodes.py” function with the “-a” argument that attempts read

411 orientation and a barcode length of eight base pairs. The resulting sequences and barcodes were
412 used to tag the reads with “split_libraries_fastq.py”, a threshold quality score of 20, and a barcode
413 size of eight basepairs. Operational taxonomic units (OTUs) were identified with a 99% similarity
414 threshold using the UCLUST algorithm [65] with the reverse strand match enabled (“-z”), and the
415 longest sequence of each OUT was chosen as representative sequence. Taxonomy assignment was
416 carried out using Mothur (v1.39.5; [53] with the UNITE database v7.2 [68] as reference and a 80%
417 confidence threshold. For each sample, sequences were randomly sampled with the function
418 “single_rarefaction.py” from the OTU tables to obtained a total number of sequences per sample
419 equal to the lowest number of reads across GTAA, BITS, and SP datasets. Taxonomy tables at the
420 genus level were then created using “summarize_taxa.py”.

421

422 **DECLARATIONS**

423 **Ethics approval and consent to participate**

424 Not applicable

425 **Consent for publication**

426 Not applicable

427 **Availability of data and material**

428 All FASTQ files with the amplicon sequences separated by barcode were deposited in the NCBI
429 Sequence Read Archive (BioProject: PRJNA485180; SRA accession: SRP156804).

430 **Competing interests**

431 The authors declare that they have no competing interests

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439 sciences, and disciplines with engineering-related applications and methods.

440 **Authors' contributions**

441 Conceived and designed the experiments: AMC and DC. Performed the lab experiments: RFB,
442 AMC, ET, and JD. Performed field sampling: PER. Performed bioinformatic analysis: AMC.
443 Wrote the manuscript: AMC, KB, and DC. All authors read and approved the final manuscript.

444

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676

677 **FIGURE LEGENDS**

678 **Figure 1. Primer design and testing.** (A) Schematic representation of the annealing sites of
679 forward and reverse GTAA, BITS, and SP primers in the fungal ribosomal ITS. Reported amplicon
680 sizes were calculated based on the ITS sequence of *Eutypa lata* (KU320617.1). (A) & (B):
681 Bioanalyzer electropherograms showing amplicons generated using GTAA primers and (A) DNA
682 from purified organisms as template and (C) field samples with different trunk disease symptoms.

683 *Agrobacterium tumefaciens*, *V. vinifera* and nuclease-free water were included as controls. AH:
684 apparently healthy, ED: Eutypa Dieback, ES: Esca and WC: wood cankers (no leaf symptoms).
685 (D) PCR products from grapevines inoculated with *N. parvum* at six weeks post-inoculation, which
686 were visible on agarose gels at five-cycle intervals, when amplified with primers GTAA, BITS, or
687 SP. L: 100 bp Ladder. (E) Cycle thresholds (Ct) measured by qPCR of the same reactions shown
688 in (D).

689 **Figure 2. *In silico* simulation of amplification and taxonomic assignment.** (A) Neighbor-
690 joining tree of the full-length ITS sequences included in the custom database used in the
691 simulation. (B) Barplots showing the number of sequences predicted to be amplified by each
692 primer set at different combinations of mismatches. H: primer head. T: primer tail. Numbers
693 correspond to the number of mismatches either in H or T. (C) Barplots showing the number of
694 false negative (FN), false positive (FP), and true positive (TP) sequences with each primer set. (D)
695 Percentage of sequences per genus correctly assigned with each primer set at the different
696 mismatch combinations.

697 **Figure 3. Results of DNA metabarcoding of mock communities.** (A) Stacked barplots showing
698 the relative abundance of genera in the mock communities identified using the GTAA and BITS
699 primers. Eu: *Eutypa*, Pa: *Phaeoacremonium*, Pm: *Phaeomoniella*, Di: *Diaporthe*, Dp: *Diplodia*,
700 Np: *Neofusicoccum*, and Vv: *Vitis vinifera*. (B) Linear correlations between observed and expected
701 abundances for each genus contained in the mock community.

702 **Figure 4. Results of DNA metabarcoding of an infection time-course.** Stacked barplots show
703 the relative abundance of the genera detected in a time course experiment after inoculation with
704 *N. parvum*. IW: wound-inoculated with *N. parvum*; NIW: non-inoculated non-wounded; NINW:
705 non-inoculated non-wounded controls.

706 **Figure 5. Results of DNA metabarcoding of field samples and comparisons with shotgun**
707 **whole-genome metagenomics (WGS).** (A) Stacked barplots showing the relative abundance of
708 the genera detected by WGS and DNA metabarcoding with GTAA, BITS, and SP primers. (B)
709 Scatter plots showing the correlations of the relative abundance obtained by DNA metabarcoding
710 and WGS. (C) Scatter plots showing the correlations of the relative abundance separately for each
711 genus obtained by DNA metabarcoding using the GTAA primers and WGS.

712 TABLES

713 **Table 1. Metabarcoding primer sequences targeting the ITS region used in this study**

Primer name	Direction	Primer sequence 5' to 3'	Citation
GTAA	Forward	AAAAC TTCAACAACGGATC	this study
GTAA	Reverse	TYCCTACCTGATCCGAGGTC	this study
BITS	Forward	CTACCTGCGGARGGATCA	[45]
BITS	Reverse	GAGATCCRTTGYTRAAAGTT	[45]
SP	Forward	CTTGGTCATTTAGAGGAAGTAA	[46]
SP	Reverse	GCTGCGTTCCTCATCGATGC	[46]

714

715 ADDITIONAL FILES

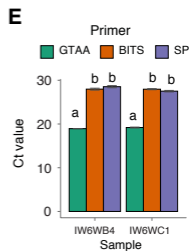
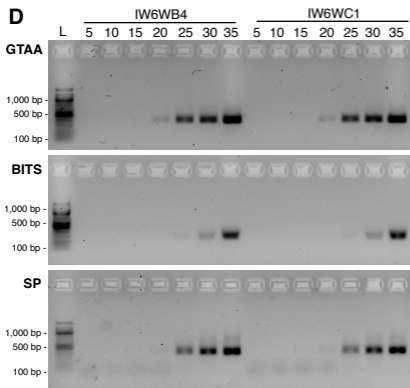
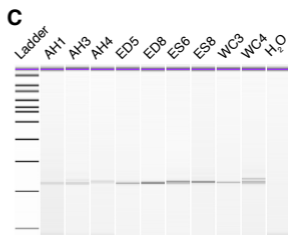
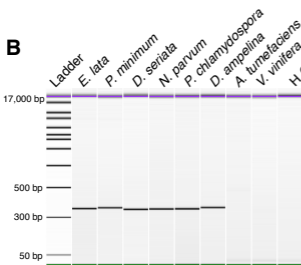
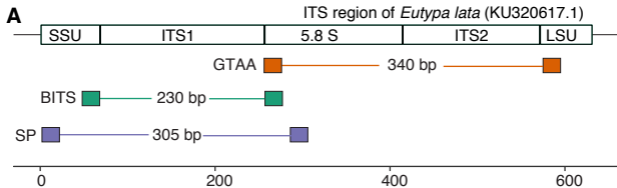
716 **Additional File 1: Text S1 (FASTA).** Assembled amplicon sequences produced by the GTAA
717 primers and sequenced with Sanger.

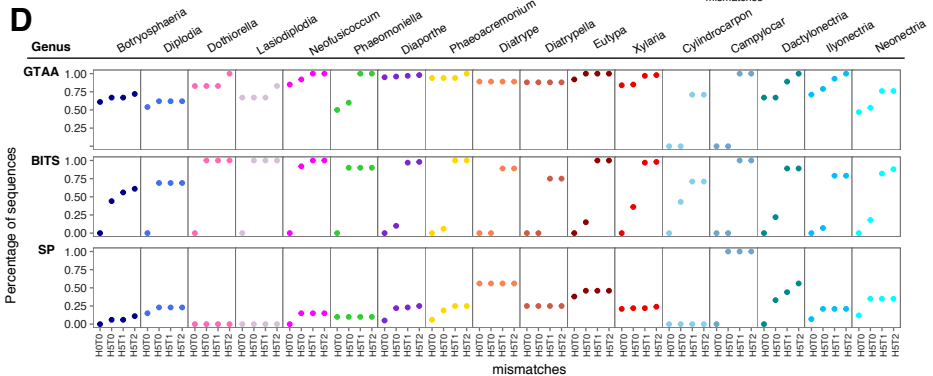
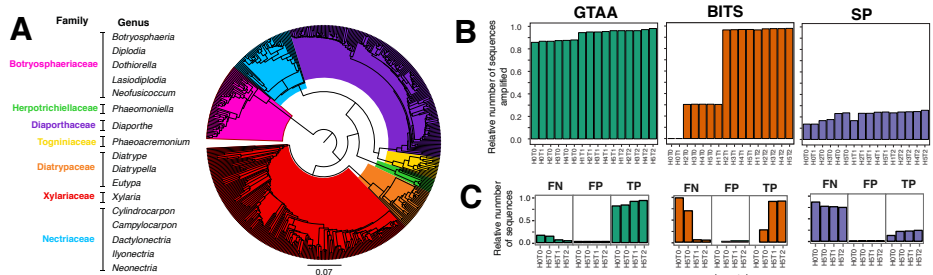
718 **Additional File 2: Table S1 (XLSX).** List of filtered sequences retrieved from the NCBI used as
719 a database for primer testing and test results per primer. TP: True Positive, FP: False Positive and
720 FN: False Negative.

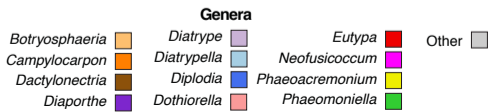
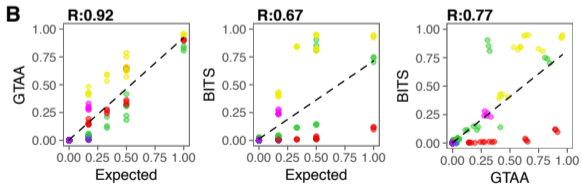
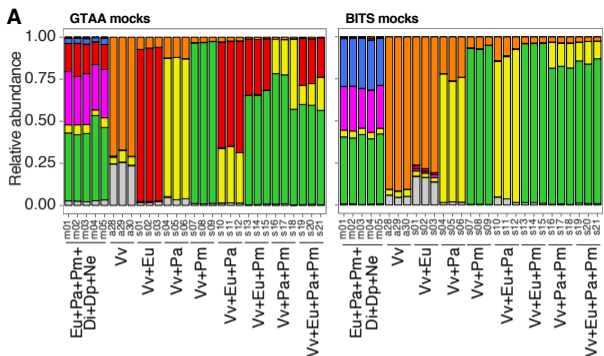
721 **Additional File 3: Figure S1 (PDF).** Venn diagram of genera detected by the GTAA and BITS
722 primers from the field samples detected with more than 0.05% abundance per sample.

723 **Additional File 4: Table S2 (XLSX).** List of forward GTAA primers with linker and barcodes

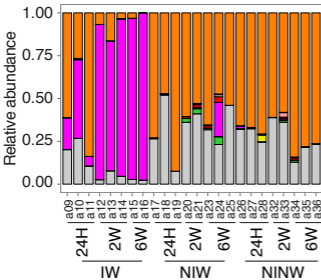
ITS region of *Eutypa lata* (KU320617.1)



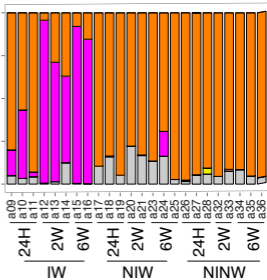




GTAA *N. parvum* experiment



BITS *N. parvum* experiment



Genera

