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4	Impact of DNA polymerase choice on assessment of bacterial
5	communities by a <i>Legionella</i> genus-specific next-generation sequencing
6	approach
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30 ABSTRACT

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32 The library preparation step is a major source of bias in NGS-based studies. Several PCR-33 related factors might negatively influence the application of NGS tools in environmental 34 studies and diagnostics. Among the most understudied factors are DNA polymerases. In 35 our study, we evaluated the effect of DNA polymerase type on the characterisation of 36 bacterial communities, more precisely *Legionella*, using a genus-specific NGS approach. 37 The assay with proof-reading high fidelity KAPA HiFi showed better amplification yield than the one with widely used non-proofreading HotStarTaq. Legionella community 38 39 richness metrics were significantly overestimated with HotStarTaq. However, the choice 40 of DNA polymerase did not significantly change the community profiling and composition. These results substantiate the use of proof-reading high fidelity DNA 41 polymerases in NGS assays and highlight the need of considering the impact of different 42 43 DNA polymerases in comparative studies and future guidelines for NGS-based diagnostic 44 tools.

46 INTRODUCTION

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Since its arrival, next-generation sequencing (NGS) has undergone fast and 48 49 continuous progress in terms of technology, allowing enhanced performance, throughput 50 and accuracy in a shorter time [1]. NGS methodologies have been widely applied in 51 environmental microbiology to provide a broader characterisation of bacterial community 52 structure, a better comprehension of the influence of environmental and anthropogenic 53 drivers in both the emergence and persistence of bacterial pathogens; and for the detection and characterisation of clinically and environmentally relevant organisms [2–4]. One of 54 55 the most environmentally and clinically important water-based bacterial taxa worth studying is *Legionella*. Legionella species are persistently and abundantly found in a 56 highly diverse set of aquatic environments and more than 25 species have been associated 57 with disease [5]. Legionellosis outbreaks are frequently reported and have been mostly 58 59 linked to hot water settings and foremost cooling towers, suggesting the need of a 60 continuous environmental surveillance. NGS assays have already been demonstrated to 61 have the capacity to provide a detailed understanding of the spatio-temporal dynamics 62 and diversity of *Legionella* species and be used as reliable and precise monitoring tools 63 of pathogenic species, such as L. pneumophila [6].

NGS has the potential to be an upgrade to the current accepted methodologies, and consequently be translated into active frameworks for routine surveillance and investigation of man-made freshwater systems. However, thorough *in vitro* and *in situ* validation studies are needed to evaluate if NGS-based methodologies are precise and reliable methodologies that are worth to be widely implemented and applied. Guidelines have already been presented and implemented for the use of NGS in clinical diagnostics and genetics screening [7–9]. Yet, no recommendations, guidelines or regulatory

71 approaches currently exist for validation and establishment of the assays and 72 interpretation of the results in environmental settings. Among the factors to consider in 73 the validation of a NGS assay are the potential bias introduced by library preparation and 74 sequencing steps, which can affect not only sensitivity and specificity of the molecular 75 assay developed but also inter-laboratory comparison studies. The potential of NGS as an 76 accurate and sensitive tool for environmental research and diagnostics can be affected by the choice of the primers, targeted taxonomic group, PCR thermo-cycling conditions, 77 template concentration and DNA polymerase [10–13]. 78

In this study, our aim was to evaluate the impact of the choice of the DNA polymerase have on assessment of microbial diversity, more specifically on the alphaand beta-diversity of *Legionella* community in freshwater systems, with a genus-specific 16S rDNA-based NGS approach.

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84 MATERIALS AND METHODS

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To study the effect of the DNA polymerase on Legionella libraries community 86 87 diversity metrics and structure, 8 freshwater samples, i.e. 2 cold drinking water (A and 88 B), 2 hot drinking water (C and D) and 4 cooling tower water (E to H) samples, were 89 collected at the campus of the Helmholtz Centre for Infection Research (HZI), in Braunschweig, Germany, and processed as previously described [6,14,15]. These 90 91 freshwater samples were then analysed by a genus-specific NGS approach using two optimised and independent assays with two hot-start DNA polymerases, HotStarTaq 92 93 (Qiagen, Hilden, Germany) and proofreading KAPA HiFi (KAPA Biosystems, Wilmington, MA, USA). A 16S rRNA gene fragment, with a length of 421 bp, comprising 94 the V3-V4 hypervariable regions was amplified with Legionella genus-specific primer 95

96 pair Lgsp17F 5'-GGCCTACCAAGGCGACGATCG-3'/ Lgsp28R 5'97 CACCGGAAATTCCACTACCCTCTC-3' [16]. Detailed descriptions of the used
98 Illumina MiSeq-based approach with both DNA polymerases as well as of the 16S rDNA
99 data processing and taxonomic classification by the bioinformatics pipeline of SILVA
100 project are given elsewhere [17,18].

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102 **RESULTS**

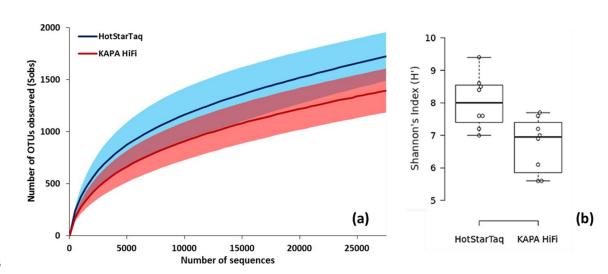
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104 Before evaluating the effect of the DNA polymerase on community diversity 105 metrics, we compared the yield of the two optimised assays by equivolume pooling of all 106 sample libraries and sequencing in a single Illumina MiSeq 250 bp paired-end run. The 107 data revealed a statistically significant difference between the sample sets concerning the 108 number of retrieved 16S rRNA Legionella gene sequences per sample, after quality 109 filtering and taxonomic assignment (Welch's t-test, P<0.05). The mean number of Legionella 16S rRNA gene sequences per sample in the sample set amplified by KAPA 110 HiFi was significantly higher $(52,385 \pm 12,907)$ than in the one amplified by HotStarTaq 111 112 $(37,343 \pm 8,995)$. Of the 8 samples, 7 had a higher number of sequences when amplified 113 with KAPA HiFi, representing, on average, an increase of 45% when compared with 114 HotStarTaq. This discrepancy does not seem to be an outcome of poor sequence quality 115 as the sample sets presented a similar percentage of rejected sequences during quality 116 filtering (1%). These results suggest that the optimised assay with KAPA HiFi polymerase provided a higher Legionella amplification yield than the one with 117 118 HotStarTaq polymerase, regardless of the observed inferior specificity to Legionella genus ($86\% \pm 11\%$ against 97% $\pm 2\%$). Nonetheless, rarefaction analyses revealed 119

120 Operational Taxonomic Units (OTU) values levelling off markedly for all samples (data

121 not shown).

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Fig. 1. Alpha-diversity assessment of Legionella communities in 8 water samples 124 125 with HotStarTaq and KAPA HiFi DNA polymerases. (a) Rarefaction curves showing progress of number of OTUs observed with increasing number of sequences analysed. 126 Number of OTUs observed were calculated after normalisation to the sample with the 127 128 lowest number of sequences. Lines indicate mean values with corresponding coloured 129 shaded standard deviation for HotStarTaq DNA polymerase (blue) and KAPA HiFi DNA polymerase (red). (b) Boxplots of Shannon's diversity index (H') of 8 water samples 130 amplified with both DNA polymerases. Bars within boxes indicate the median value of 131 diversity. Whiskers extend to data points that are less than 1.5 x the interquartile range 132 away from $1^{st}/3^{rd}$ quartile. 133

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Alpha-diversity metrics, i.e. observed OTU richness (OTU_{obs}) and Shannon's diversity index (H'), were calculated, using the software package Explicet [19], for the *Legionella* community of the 8 water samples amplified with both KAPA HiFi and HotStarTaq DNA polymerases, after clustering of NGS reads into OTUs using an identity criterion of 98%. (**Fig. 1**). The use of HotStarTaq, comparatively to KAPA HiFi, showed a steeper slope and exhibited a higher *Legionella* richness with a statistically significant increase of the number of OTUs observed (Welch's t-test, p<0.05) regardless of the

sample (Fig. 1(a)). The sample set when amplified by HotStarTaq had an averaged 142 143 observed richness of 2389 ± 851 OTUs. However, when amplification performed with 144 KAPA HiFi, the sample set showed an averaged observed richness of 1846 ± 469 OTUs, representing a decrease of 23% to the values of Legionella observed richness shown with 145 146 HotStarTaq. Also, the number of singletons and doubletons were reduced by an average of 20% and 22%, respectively, with KAPA HiFi. Similar findings were observed when 147 148 separately amplifying technical replicates (n=3) of a single strain (L. pneumophila ATCC 33152^T), with a total reduction of 38% in singletons and doubletons detected when using 149 150 KAPA HiFi (data not shown).

Moreover, the comparison of the two sample sets showed a statistically significant difference in the Shannon's diversity index (Welch's t-test, P<0.05) with higher values observed in samples amplified with HotStarTaq (**Fig. 1(b**)). Shannon's diversity index on the samples amplified with HotStarTaq ranged from 7.06 to 9.94 (mean: 8.12 ± 0.97) whereas on the samples amplified with KAPA HiFi the index values ranged from 5.64 to 7.99 (mean: 6.86 ± 0.88).

To assess beta-diversity, we characterised the relative abundance and composition 157 158 of the Legionella phylotypes in the freshwater samples amplified with each DNA 159 polymerase (HotStarTaq and KAPA HiFi) and calculated the Bray-Curtis similarity, using PRIMER (Version 7.0.7) [20], to compare the level of community structure 160 161 similarity between different samples and between the same sample amplified with the two 162 different DNA polymerases (Fig. 2). The term *Legionella* phylotype is used according to what has been previously described [17]. Briefly, NGS sequences were assigned to a 163 164 species when sequence identity $\geq 97\%$. The term phylotype englobes the acknowledged Legionella species as well as other defined sequence clusters with sequence identity 165 <97% to a known species. 166

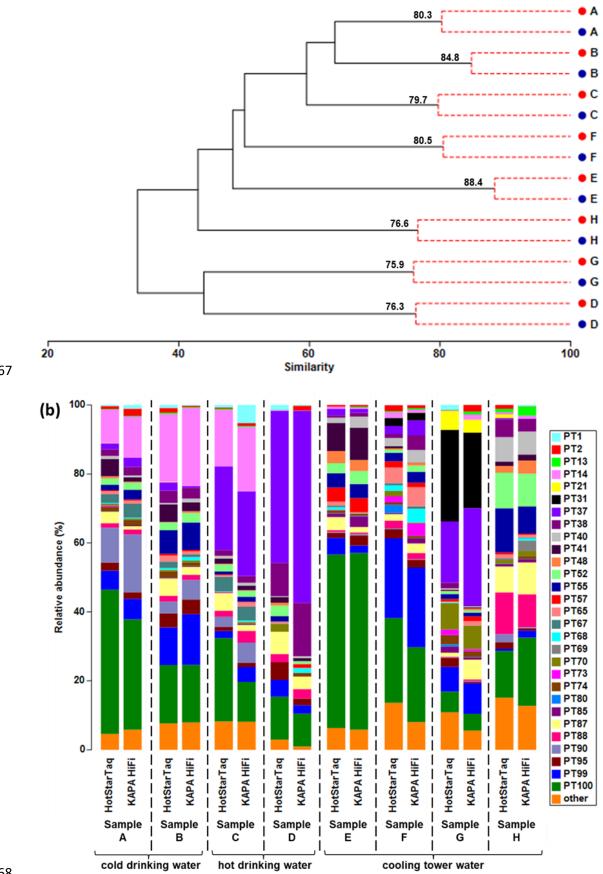


Fig. 2. Effect of DNA polymerases on assessment of Legionella communities 169 170 profiling. (a) Dendogram showing group-average hierarchical clustering of Legionella 171 communities amplified in 8 water samples with both HotStarTaq (blue circle) and KAPA 172 HiFi (red circle) DNA polymerases, using Bray-Curtis similarity, SIMPROF test was performed and red dashed lines in cluster analysis represent groups of samples that do 173 174 not significantly differ in their Legionella community (P>0.05). Values on branches indicate Bray-Curtis similarity index. Sample A, cold drinking water March 2009; sample 175 176 B, cold drinking water April 2009; sample C, hot water March 2009; sample D, hot water 177 January 2014; sample E, cooling tower water April 2013; sample F, cooling tower water 178 June 2013; sample G, cooling tower water August 2013; sample H, cooling tower water 179 January 2015. (b) Bar graph representing relative abundances of *Legionella* phylotypes 180 in 8 water samples amplified with both HotStarTaq and KAPA HiFi DNA polymerase. 181 Phylotypes labelled from 1 to 52 correspond to described Legionella species. PT1, L. 182 adelaidensis; PT2, L. anisa; PT13, L. drozanskii; PT14, L. dumofii; PT21, L. gratiana; 183 PT31, L. maceachernii; PT37, L. pneumophila; PT38, L. quateirensis; PT40, L. rowbothamii; PT41, L. rubrilucens; PT48, L. tucsonensis; PT52, L. worsleiensis. 184 185 Different freshwater samples are separated by a dashed line.

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Based on SIMPROF analysis (**Fig. 2(a**)) [21], we observed that the difference introduced in the *Legionella* community structure by the use of the two different enzymes was not statistically significant in the sample set used (SIMPROF, P>0.05).

190 Nonetheless, shifts in the Legionella community of every sample studied were observed (Fig. 2(b)), with the similarity levels ranging from 75.9 (Sample G) to 88.4 191 192 (Sample E), with an averaged value of 80.3 ± 4.4 . Clear clusters for every sample, independently of the type of water, were formed and unambiguously separated the 8 193 194 different water samples tested. Furthermore, analysis of the composition of the Legionella 195 community in each sample amplified with the two DNA polymerases, revealed a substantial overlap of Legionella phylotypes. When the composition of the communities 196 amplified by the two enzymes were compared, 7 out of 8 samples shared more than 85% 197

of the phylotypes (data not shown). Sample D showed the lowest percentage of sharedphylotypes (60.1%).

200

201 **DISCUSSION**

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203 Currently, NGS technologies are widely applied in the characterisation of 204 communities, discovery of novel microorganisms and pathogen detection [22]. For NGS 205 assays to succeed on these fields, a high amplification efficiency during library 206 preparation and target enrichment is of critical importance to downstream sequencing, especially when targeting specific low-abundant taxa in complex environmental samples. 207 208 The amplification efficiency is influenced by several PCR-related factors, including the 209 choice of DNA polymerase. Our results, despite a better performance of KAPA HiFi, 210 confirmed that the two different polymerases had a very good amplification performance 211 allowing a subsequent appropriate sequencing depth of the target. This was in a certain 212 amount expected due to the hot-start nature of the two enzymes. Plus, our findings 213 highlight that the choice of DNA polymerase, as well as optimisation of the cycling 214 parameters in PCR-based NGS assays, is important to find an appropriate balance 215 between non-specific amplification and yield, in order to maximise detection/sensitivity. 216 In addition, several studies have reported the impact of different amplification and 217 sequencing-related factors on high-throughput studies of microbial communities and how 218 these might distort diversity estimates and community profiling [23–25]. Yet, only a few studies have evaluated the effects of DNA polymerases on the study of communities [26-219 220 29].

Past NGS studies have revealed increasing richness values with increasing
sequencing depth [17,30]. Yet, our results showed that the freshwater samples amplified

by HotStarTaq DNA polymerase, despite a lower sequencing depth, had significantly
higher OTU-based alpha-diversity metrics than KAPA HiFi DNA polymerase, indicating
that alpha-diversity measurements are enzyme-dependent, as previously suggested [26–
226 29].

227 A recent study by Pereira et al. [17], using the same approach, revealed that when 228 comparing the sequence accuracy and quality of the generated NGS Legionella libraries, 229 KAPA HiFi had a significantly higher fidelity, with a mean error rate of 0.38% against 230 0.45% and 65% of error-free sequences against 39%. All data considered, the significant 231 increase in bacterial community richness and diversity, plus the significant higher number 232 of singletons and doubletons observed after amplification of the same 16S rRNA region 233 with HotStarTaq DNA polymerase, is almost certainly linked to a higher generation of 234 erroneous sequences with an enzyme without proofreading activity. This leads to an 235 increased number of OTUs and consequently an artificial overestimation of the richness 236 and diversity estimators [23,31,32]. Similar inflated community richness with 237 polymerases presenting lower fidelity was observed in bacterial libraries by Wu et al. [26] 238 and fungal libraries by Oliver et al. [28]. Yet, conflicting reports by Ahn et al. [27] and 239 Qiu et al. [29] have detected an overestimation of richness with a high-fidelity enzyme 240 instead, attributing it to a higher frequency of chimeric sequences. In our study, as we are 241 targeting a specific taxa, low diversity libraries are generated, not allowing to precisely 242 quantify the potential amount of chimeric sequences in the water samples. However, 243 analysis of Legionella mock communities with the same NGS approach has revealed a small representation of these spurious sequences with KAPA HiFi [17]. 244

When comparing beta-diversity metrics after amplification with both DNA polymerases, the impact of the enzymes on the *Legionella* community composition and structure (beta-diversity) was not significant for the samples studied, contrarily to what

was indicated by Wu et al. [26], when comparing the DNA polymerases TAKARA ExTaq 248 249 with PfuUltra II Fusion HS. Though HotStarTag and KAPAHiFi have different 250 properties, kinetics and slightly different polymerase cocktails were used, the two enzymes seem to have an alike amplification profile of the different Legionella 251 252 phylotypes. However, differences in Bray-Curtis similarity between the samples 253 amplified by two distinct DNA polymerases were observed (mean BC: 80.3 ± 4.4). These 254 discrepancies on Legionella community profiling are slightly higher to the ones 255 previously observed between technical replicates (mean BC: 86.8 ± 4.9) [17].

In summary, our data highlights the potential advantageous effects of the use of proof-reading high-fidelity enzymes such as KAPA HiFi on NGS library preparation methodologies. As well, this study emphasises the influence the choice of DNA polymerase has on the characterisation of microbial communities, especially in alphadiversity metrics, and the critical importance of taking this factor in consideration in comparative studies and in the future use of these high-throughput technologies for pathogen detection and quantification.

263

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267

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271 CONFLICTS OF INTEREST

272 The authors declare no conflict of interest.

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