

Development of target sequence capture and estimation of genomic relatedness in a mixed oak stand

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9 Abstract

10 Anticipating the evolutionary responses of long-lived organisms, such as trees, to environmental changes, requires the assessment of genetic variation of adaptive traits in natural populations. To this 11 12 end, high-density markers are needed to calculate genomic relatedness between individuals allowing 13 to estimate the genetic variance of traits in wild populations. We designed a targeted capture-based, next-generation sequencing assay based on the highly heterozygous pedunculate oak (Quercus robur) 14 15 reference genome, for the sequencing of 3Mb of genic and intergenic regions. Using a mixed stand of 293 Q. robur and Q. petraea genotypes we successfully captured over 97% of the target sequences, 16 corresponding to 0.39% of the oak genome, with sufficient depth (97X) for the detection of about 17 18 190 thousand SNPs evenly spread over the targeted regions. We validated the technique by evaluating its reproducibility, and comparing the genomic relatedness of trees with their known 19 20 pedigree relationship. We explored the use of the technique on other related species and highlighted the advantages and limitations of this approach. We found that 92.07% of target sequences in Q. 21 22 suber and 70.36% of sequences in Fagus sylvatica were captured. We used this SNP resource to estimate genetic relatedness in the mixed oak stand. Mean pairwise genetic relatedness was low 23 24 within each species with a few values exceeding 0.25 (half sibs) or 0.5 (fulls sibs). Finally we applied the technique to a long standing issue in population genetics of trees regarding the relationship 25 between inbreeding and components of fitness. We found very weak signals for inbreeding 26 27 depression for reproductive success and no signal for growth within both species.

29 **1** Introduction

30 Predicting the evolutionary potential of natural populations is a major goal in many biological domains (e.g evolutionary biology, landscape ecology, conservation biology) given the global 31 32 changes currently faced by organisms and populations. From an evolutionary perspective, the principal challenge is predicting the evolutionary changes required to track ongoing environmental 33 34 changes and to identify key traits likely to respond to ongoing natural selection. These concerns are 35 particularly important in the case of forest trees, which have long generation times. Their evolutionary response must therefore occur within a very small number of generations. The 36 37 prediction of evolutionary responses requires the estimation of essential genetic parameters, such as 38 selection gradients, heritability and evolvability, in situ, at the site at which selection is acting 39 (Conner et al., 2003; Kruuk and Hill, 2008). Trait heritability can be estimated in situations in which 40 the phenotypic similarity between individuals can be compared to their genetic similarity or 41 relatedness (Ritland, 2000).

42 In animals, such as mammals and birds, such studies are generally performed on pedigreed populations (Kruuk, 2004). However, for trees, it is almost impossible to obtain pedigrees extending 43 over more than two generations, at least over the lifetime of the scientist. Fortunately, recent 44 45 developments in genomics, and the use of NGS sequencing have made it possible to measure the realized relatedness between individuals based on a large number of genetic markers, as it has been 46 47 shown that the realized proportion of the genome identical by descent is more precisely estimated 48 with a large number of molecular markers than with pedigree relationships (Kardos et al., 2015), 49 These new methods thus open up new possibilities for the estimation of heritability and genetic 50 variances in situ (Bérénos et al., 2014). Such approaches have already been implemented in trees 51 (Castellanos et al., 2015). We addressed the aforementioned evolutionary questions, by identifying a large number of unlinked SNP markers in species of the Fagaceae family. These markers are widely 52 53 distributed across the genome, encompassing genes and regions of biological interest, as well as regions assumed to be neutral. 54

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56 Whole-genome shotgun sequencing is an easy way to sequence a genome randomly and to identify 57 large numbers of molecular markers suitable for our objectives. However, shotgun sequencing may

constrain marker development in highly repetitive genomes, such as that of oaks, which consists of
52% transposable elements, as reported by Plomion et al. (2018).

Targeted sequence capture coupled with NGS constitutes an efficient alternative approach to the exploration of genetic diversity in a very large number of genomic regions and specimens. The use of sequence capture techniques provides evolutionary biologists with easy access to nucleotide diversity, for addressing various research questions, as already demonstrated in in arable crops (Zhou et al., 2012), fruit (Tennessen et al., 2013) and forest trees (Holliday et al., 2016; Fahrenkrog et al., 2017).

66 Furthermore, these techniques provide high sequence coverage for a small set of target sequences, making it possible to multiplex several samples, thereby reducing the cost of large-scale applications, 67 68 for population genetics studies, for example. Sequence capture techniques require access to a 69 reference genome, but provide highly reproducible SNPs and markers with greater transferability across species than for other pangenomic marker systems (e.g RADseq or GBS) (Harvey et al., 70 71 2016). Intra- and interspecific reproducibility is a prerequisite for comparative studies across 72 populations or related species, even if sampling and molecular analysis are performed at different 73 times. For example, George et al. (2011) developed a genomic capture approach in humans that 74 successfully captured about 96% of coding sequences in monkeys (George et al., 2011). Similarly, in gymnosperms, a common capture design established for spruce and lodgepole pine (Suren et al., 75 76 2016) successfully captured more than 50% of the targeted bases with a coverage of at least 10X.

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Our main objective here was to develop a large number of SNPs for estimating the genetic relatedness and inbreeding coefficient in a mixed oak stand containing two sister species: *Quercus petraea* and *Quercus robur*. We thus developed a targeted sequence enrichment strategy, explored its transferability to related species and applied the detected markers to a long-standing question in tree population genetics: the relationship between inbreeding and fitness components.

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84 2 Materials and Methods

- 86 2.1 Target sequence capture
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88 2.1.1 Plant material and DNA extraction

Leaves were collected from 278 adult oak trees and 15 siblings (8 Q. petraea and 7 Q. robur) in a 89 90 mixed oak stand (Quercus petraea - Quercus robur) located in the Petite Charnie State Forest in western France (latitude: 48.086°N; longitude: 0.168°W). This population corresponds to cohort #1b 91 92 described by Truffaut et al. (2017). The trees were all cut between 1989 and 1993, but were grafted 93 and maintained in a common garden in a nursery located in Guéméné (latitude: 47.63°N; longitude: -94 1.89°W). Leaves were sampled from these grafted plants for DNA extraction. The 15 siblings were 95 sampled during the natural regeneration of the adult trees in the Petite Charnie Forest and are part of cohort #2 described by Truffaut et al. (2017). The parents of the siblings were identified by molecular 96 97 parentage analysis in a previous study (Truffaut et al., 2017), and pedigree relationships were inferred 98 between the parents and their offspring, and between the offspring (Figure 1).

We also collected leaves from two adult beech trees (*Fagus sylvatica*) from St Symphorien, on a tributary of the Ciron river, in south-west France (latitude: 44.25°N; longitude: 0.29°W) and two adult cork oak trees (*Quercus suber*) growing at the INRA Research Station at Pierroton in southwest France (latitude: 44.44°N; longitude: 0.46°W). We considered a total of 300 samples in all, as three adult trees from the Petite Charnie forest were sampled twice.

For DNA extraction, leaves were frozen and stored at -80°C. DNA was extracted with the QIAGEN DNeasy Plant Maxi Kit and DNA quality and quantity were assessed with a spectrophotometer (NanoDrop Thermo Fisher Scientific, Waltham, USA) and a fluorometer (Tecan Infinite F200, Männedorf, Switzerland) with a Broad Range Quant-it dsDNA kit (Thermo Fisher Scientific, Waltham, USA). For each replicated individual, DNA was extracted independently from each of the two samples, independent libraries were constructed and replicates were sequenced in separate proton sequencing runs.

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112 2.1.2 Target sequence selection and probe design

113 We used in-solution hybridization-based sequence capture technology, based on the results of 114 Mamanova et al. (2010). These authors compared the performance of several target-enrichment

techniques, assessed on the basis of several criteria: percentage of target sequences captured, proportion of sequencing reads on target, variability of sequencing coverage across target regions, reproducibility, cost, ease of use and minimum amount of DNA required. Given the number of samples studied, the target size imposed by our resources (2.9 Mb) and the relatively large proportion of repetitive sequences, hybridization-based sequence capture appeared to be the most relevant method in our case.

121 The haploid version of the Quercus robur genome (haplome V2.3), available from http://www.oakgenome.fr/ and described by Plomion et al. 2018, was used for probe design (Plomion 122 123 et al., 2018). The oak genome consists of 25,808 predicted protein-coding genes spread over 1,409 124 scaffolds. The oak genome is highly repetitive. We therefore limited the length of target sequences to 125 150 bp, when necessary, to avoid repetitive sequences. Target sequences were selected on the basis 126 of previous results for genetic diversity and the expression of genes of ecological and physiological 127 relevance. Indeed, over the last 10 years, various genetic surveys have been conducted to identify 128 expressed candidate genes, outlier genes displaying species or population genetic differentiation, or 129 genes displaying significant genotype-phenotype or genotype-environment associations. We 130 reviewed all these surveys and used relaxed thresholds of selection to identify candidate sequences 131 for genomic capture (Table 1). As our resources were limited to a total sequence length of 2.9 Mb for 132 capture, we could not consider entire genes as targets for probe design. We therefore selected target 133 sequences within each gene, depending on its length. For genes of less than 1.5 kb in length, we 134 identified a single 150 bp target sequence located in an exon. Longer genes were artificially 135 subdivided into three regions, and we selected two 150 bp target sequences located in two extreme 136 regions of the gene: one within an exon, and the other within an intron-exon transition (Supplemental 137 file 1). In total, our capture experiment included 9,748 candidate genes. We completed the selection 138 and design of target sequences for genomic capture, by including 150 bp sequences located in 139 intergenic regions. These sequences were selected with a 100 kb sliding window. We examined 140 8,936 windows, and retained a 150 bp sequence at the beginning of the window only if no other 141 target sequence had previously been identified in the window (Table 1). If the target sequence 142 colocalized with a transposable element (TE), it was shifted 150 bp further along in the genome.

Once target sequences had been identified, we retained only those with a GC content between 30% and 60%, as suggested by Chilamakuri et al. (2014). We avoided repetitive regions of the genome by aligning candidate target sequences against the oak genome with BLAT v.35x1, using default

- parameters (Kent, 2002), and we retained target sequences with fewer than 10 alignments on the oakgenome that were distant from TEs.
- 148 Following this strategy, we identified 15,623 candidate target sequences, which were sent to Agilent
- 149 Technologies (Agilent Technologies, Santa Clara, California, USA) for the design of 120 bp probes.
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151 2.1.3 Library preparation

152 Our targeted enrichment procedure was based on Agilent's SureSelect target enrichment system for 153 Ion Torrent Proton sequencing (Thermo Fisher Scientific, Waltham, MA, USA). We randomly 154 assigned the 300 DNA samples to 20 groups, each corresponding to a proton sequencing run. The 15 155 samples in each run were labeled (indexed 1 to 15). We assessed reproducibility, by duplicating three samples corresponding to three individuals. For the three duplicated samples, DNA was extracted 156 157 separately from the two samples, independent libraries were constructed and sequencing was 158 performed in separate runs. A pre-capture library was prepared for each sample, using the 159 NEBNext[®] Fast DNA Library Prep Set for Ion Torrent[™] from New England Biolabs (Ipswich, MA, 160 USA) according to the manufacturer's instructions: 400ng of genomic DNA was sheared, with an 161 M220 focused ultrasonicator (Covaris, Inc., Woburn, MA, USA), to yield 200 bp fragments. Each 162 sheared DNA sample was subjected to end repair and ligated to barcoded adapters. We then selected 163 DNA fragments of 300 bp in size by two consecutive Agencourt® AMPure® XP steps (Beckman 164 Coulter, Inc., Brea, CA, USA): 0.7X then 0.15X. The libraries were subjected to 11 cycles of 165 amplification. Each library was quantified with a Qubit Fluorometer, with the Qubit[™] dsDNA HS 166 Assay Kit (Thermo Fisher Scientific, Waltham, MA, USA). Then equimolar pools of three libraries 167 were prepared (250 ng for each library) for target enrichment.

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169 2.1.4 Target enrichment

The size of the DNA library was limited by the use of in-solution capture, which requires an excess of probe over template. Hybridization to the probes was carried out for 24 h at 65°C, according to the Agilent protocol, in a thermocycler, with 750 ng of library. Following the hybridization and washing steps, the recovered targeted DNA fragments were amplified in KAPA HiFi HotStart ReadyMix (Kapa Biosystems, Wilmington, MA, USA) for 40 s at 98°C, followed by eight cycles of 30 s at

175 98°C, 30 s at 62°C, and 30 s at 72°C, with a final extension for 5 min at 72°C. The captured library 176 pools were quantified by qPCR on a LightCycler® 480 System (Roche Molecular Diagnostics), with 177 the Ion Library TaqMan[™] Quantification kit. In total, 20 pools of 15 libraries were used in 178 equimolar amounts, with a final concentration of the pooled samples of 5 pM for sequencing on an 179 Ion Proton System (Thermo Fisher Scientific, Waltham, MA, USA).

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181 2.1.5 Sequence enrichment

182 For each sample, high-quality Ion Torrent proton reads were demultiplexed and subjected to quality 183 control with Torrent suite V5.0.5 (Thermo Fisher Scientific). Reads were independently aligned with 184 the oak genome, using the Torrent Mapping Alignment Program (TMAP, Thermo Fisher Scientific) and the default parameters for the Torrent suite. We estimated target enrichment by quantifying the 185 186 proportion of sequencing reads correctly aligned with the target sequences. For each sample, this 187 "on-target" set of reads was considered for further analysis. We investigated the coverage of target sequences and calculated the percentage of the length of the target covered by at least one read. 188 189 These analyses were performed with custom scripts developed in Python V2.7.2.

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191 **2.2** SNP detection and population genetics analyses

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193 2.2.1 SNP detection and filtering

194 For each sample (including cork oak and beech, which were used to test the transferability of the 195 capture probes to related species), SNPs were independently called, first with the *mpileup* function of 196 samtools V1.3.1, and then with the bcftools function V1.1-60-g3d5d3d9 (Li et al., 2009). We 197 considered only diallelic variants with a coverage of more than 10X. The minimum allele frequency 198 (MAF, upper case used at the individual level) within an individual, calculated on the basis of all the 199 reads containing the SNP, was set to 30%. A nucleotide polymorphism was considered to be an SNP, 200 if at least one individual was found to be heterozygous at the position concerned within the whole 201 population of 300 samples. For studies of relatedness between individuals, we considered only the 202 293 oak trees from the Petite Charnie forest (278 adults + 15 siblings). The SNP detection pipeline is 203 described in Supplemental file 2. We performed multiple controls and filtering steps in the Petite

Charnie population (*i.e.* 293 trees). We removed all trees for which more than 20% of the SNPs were missing. Similarly, SNPs scored in less than 95% of the trees were removed from the dataset, together with SNPs located on the 538 unanchored scaffolds of the oak genome (Plomion et al., 207 2018).

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209 2.2.2 Assignment of individuals to species

210 For the assignment of each individual to a species, we retained markers in Hardy-Weinberg 211 equilibrium located at least 1,000 bp apart, to avoid a redundancy of marker information due to 212 linkage disequilibrium. We assigned each individual to a species (i.e. cluster) with the 213 fastSTRUCTURE V1.0 algorithm (Raj et al., 2014). We allowed one to five clusters, with default 214 parameters, and the DISTRUCT algorithm was run over assignments based on cluster numbers of 215 two to five, to determine the most likely number of clusters. We assigned individuals strictly to one 216 species (Q. robur or Q. petraea) excluding admixed individuals on the basis of the posterior 217 probability of each individual belonging to one of the clusters.

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219 2.2.3 Estimation of genomic relatedness and inbreeding

220 We investigated the genetic relatedness between trees, by removing markers in linkage disequilibrium ($r^2 > 0.4$) with their neighbors, using the indep-pairphase function of PLINK 221 222 v1.90b3.34 (Purcell et al., 2007) (window size of 50 markers). We performed a Hardy-Weinberg 223 equilibrium exact test (Wigginton et al., 2005) with the -hardy function of PLINK, and p-values 224 were adjusted according to the FDR method of Benjamini & Hochberg (1995), with the R function p. adjust (Benjamini, Y. and Hochberg, Y., 1995). Only markers with a P-value greater than 0.05 225 226 after correction were retained. From these markers, we computed the Fst for each marker common to 227 both populations (Q. petraea and Q. robur) with the function Fst from the R package pegas (Paradis 228 2010). Finally we considered six sets of markers defined on the basis of minimum allele frequency, 229 considered here at the population level (maf, in lower case, for population level): we selected markers 230 with a maf exceeding a threshold of 0.01, 0.05, 0.1, 0.15, 0.3 or 0.4 (Supplemental file 2).

For each species and each set of SNPs, the genomic relatedness matrix (G) between individuals was estimated as:

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$$G = \frac{(M-P) \cdot (M-P)}{2\Sigma pi \cdot (1-pi)}$$

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where M is an n*m matrix of genotypes scored as -1, 0 or 1 for homozygote, heterozygote, alternative homozygote, P is a n*m matrix of allele frequencies computed as 2 (pi - 0.5), p_i is the maf at locus i, n is the number of individuals and m is the number of markers, as described by Van Raden (2008), with the kin function of the R package synbreed (VanRaden, 2008; Wimmer et al., 2012).

241 As indicated above, 15 offspring from the Petite Charnie stand were previously genotyped for 82 242 SNPs, and their parents were identified by parentage analysis (Truffaut et al., 2017). The 15 siblings were either full-sibs or half-sibs from 13 different adult trees, resulting in a total of 54 pairwise-243 244 related individuals. Eight siblings were the offspring of six adult Q. petraea trees, whereas seven 245 were the offspring of seven adult Q. robur trees. Four different pedigree relationships were identified 246 among these 54 pairs of trees: parent-offspring selfed, parent-offspring, full sib-full sib, half sib-half 247 sib. These relationships corresponded to three different expected coefficients of relationship: 1, 0.5, 248 0.25 (Figure 1). For the 54 pairs of trees, we compared genomic relatedness (G) with the expected 249 pedigree relatedness. Finally, we also calculated the genomic relatedness based on the 82 SNPs 250 obtained in a previous study (Truffaut et al., 2017). In the genomic relatedness matrix (G), diagonal elements (G_{ii}) correspond to the relatedness of each individual i to itself relative to population allelic 251 frequencies. In a theoretical population, at equilibrium, with no inbreeding, each individual should 252 have a G_{ii} of 1. Inbreeding is thus assessed as G_{ii} -1 (Van Raden, 2008). The deviation from 0 is 253 254 interpreted as the individual level of inbreeding relative to the population: the coefficient of genomic 255 inbreeding can be positive (*i.e.* individuals are more homozygous than expected from population 256 allelic frequencies) or negative (*i.e.* individuals are less homozygous than expected from population 257 allelic frequencies).

259 2.2.4 Correlation between genomic inbreeding and fitness

We used two traits as proxies for fitness: (i) the reproductive success of each adult tree, as assessed by the parentage analysis of 2,500 offspring and the adult trees, and (ii) the growth of each tree, as assessed by measuring stem circumference at breast height when the trees were cut. The method used to assess reproductive success has been described elsewhere (Truffaut et al., 2017). For each species we used the glm function of R to generate a generalized linear model with the number of offspring regressed against environmental variables and the inbreeding level, according to the formula:

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$$\mathbf{g}(\mathbf{F}_i) = \alpha + \beta_1 \mathbf{X}_{i1} + \gamma \mathbf{I}_i + \varepsilon_i$$

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269 where F_i is the reproductive success of individual i, α is the intercept, β_i is the regression coefficient 270 associated with the first axis of principal component analysis (PCA) on the five environmental variables (i.e. elevation, pH, soil moisture, C/N ratio, organic matter content, see Truffaut et al., 2017 271 272 for details), X is the first PC value extracted from this PCA, Ii is the inbreeding coefficient of 273 individual i associated with the regression coefficient γ , ε_i is the residual error and g is a log-link 274 function associated with the Poisson distribution data. Independent variables were centered such that 275 the intercept of the model corresponded to the phenotypic mean for the population. This 276 transformation had no effect on the regression coefficient values, their standard error or the 277 associated *P*-values. We applied a similar approach to the circumference, except that we used a linear model, as circumference is a normally distributed quantitative variable, and we added the age at 278 279 which each tree was cut as an independent variable (range: 78 to 102 years).

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281 3 Results

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283 *3.1* Target sequence capture

285 3.1.1 Agilent probe design

286 The 3P Quercus robur reference genome was used for probe design (Plomion et al., 2018).

287 The mean size of the target sequences was 150 bp and the probes were 120 bp long. One or two non-288 overlapping probes were therefore designed per target sequence, resulting in a total of 33,931 120 bp 289 probes designed with SureSelect eArray software (Agilent Technologies, Santa Clara, California, 290 USA). These probes covered a total of 2,897,647 bp (i.e. 0.39% of the estimated haploid genome 291 size). In total, 23,704 probes targeted 11,446 (44.35%) of the 25,808 predicted protein-coding genes 292 and 10,227 probes targeted intergenic regions (Table 2). In total, 11,120 probes (46.91%) targeted 293 exons, whereas 6,731 (28.40%) targeted intronic regions and 5,853 (24.69%) targeted exon-intron 294 regions.

The probes designed successfully avoided repeated regions within the genome, as fewer than 10 alignments with the oak genome were identified for 97.36% of the probes (33,034 probes).

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298 3.1.2 Target sequences identification

299 We selected a total of 15,623 genomic regions for capture (*i.e.* 2,914,160 bp), as described in Table 300 1. Agilent Technologies successfully designed 33,931 probes for 15,477 target sequences (99.07%). 301 Among the target sequences, 4,031 (26.05%) corresponded to intergenic regions and 11,446 302 (73.95%) corresponded to genes (Table 2). In total, 4,960 (43.33%) sequences corresponded to 303 exons, whereas 2,991 (26.13%) sequences were located in intronic regions and 3,495 (30.54%) were 304 located in exon-intron regions. The 4,031 intergenic target sequences were distributed as follows: an 305 initial set of 1,796 intergenic target sequences (Table 1), with 2,235 sequences of 150 bp in length 306 used as putative selectively neutral control regions for population genetic analyses. These control 307 regions were evenly distributed over the genome.

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309 3.1.3 Efficiency of target enrichment

The probes designed captured 15,477 target sequences, corresponding to 2,897,647 bp of *Q. robur* DNA. In total, 20 pools of 15 individuals each were independently sequenced with the Ion Torrent Proton sequencing system (*i.e.* 300 samples), with three samples sequenced twice. Each sequencing run produced between 65,426,948 and 134,977,869 reads (Supplemental file 3). Target enrichment

314 was assessed by aligning the reads with the oak genome: on average, for each run, 25.20% of the 315 reads captured 97.19% (i.e. 15,042) of the target sequences (Supplemental file 3). On average, 316 95.47% of the length of the target sequences was captured, and the mean coverage depth over all 317 samples was 96.81X, (range: 48.39X to 161.67X). Coverage length was 95.82% and 98.24X 318 coverage was achieved for the set of Q. robur and Q. petraea samples from the Petite Charnie stand 319 (*i.e.* 296 samples). The size of the sequenced reads ranged from 140 bp to 190 bp (mean: 174 bp). 320 The length of the sequencing reads was significantly positively correlated with the percentage of ontarget sequences (adjusted $R^2=0.2892$, *P*-value=2.2e-16) (Supplemental file 4). 321

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323 **3.1.4** SNP calling

We identified 191,281 polymorphic sites in one of the 297 trees, distributed between 13,572 target 324 325 sequences (87.69%). The number of SNPs in target sequences ranged from 1 to 603 (Figure 2A). 326 Most of the target sequences displaying polymorphism (10,419, 67.32%) contained between one and 327 20 SNPs. The SNPs were, thus, evenly spread over most of the target sequences. We classified these 328 SNPs into genic and intergenic sites on the basis of the oak gene model (Plomion et al., 2018). There 329 were 191,281 SNPs in total: 92,002 (48.10%) were located in intergenic regions and 99,279 330 (51.90%) were located within genes. In total, 51,536 SNPs (51.91%) were exonic, 43,075 SNPs (43.39%) were intronic and 4,668 SNPs (4.70%) were located in UTR regions (2,131 in the 5'UTR 331 332 and 2,537 in the 3'UTR) (Figure 2 B). On average, 7.28 and 10.49 SNPs were detected every 100 bp 333 in genic and intergenic regions, respectively. Intergenic regions were much less covered than genic regions, with a median sequencing depth of 97 and 63 in genic and intergenic regions, respectively. 334 335 Finally, we detected a mean of 13,219 SNPs per tree within the Petite Charnie population.

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337 3.1.5 Reproducibility

Mean sequencing depth differed considerably between proton sequencing runs (Supplemental file 5), even though number of samples per pool in the sequencing runs was identical (15). Four variables were correlated, to some extent, with sequencing depth: the percentage of reads on target (r^2 =0.182753, Figure 3 A), the number of SNPs detected (r^2 = 1.165e+01, Figure 3 B), the number of captured target sequences (r^2 = 6.175e-01, Figure 3 C) and the mean length of the captured sequences (r^2 = 7.401e-03, Figure 3 D).

344 The genomic capture assay was repeated twice for three oak genotypes of the Petite Charnie 345 population. For each genotype, the number of captured targets and the length of the capture sequence 346 were similar (Table 3A). Given the different sequencing depths of the different runs and the stringent 347 filters applied for SNP detection (intra-individual MAF = 30%, depth ≥ 10), for each individual, we did not capture the entire set of targeted SNPs (Table 3A) (80.31% for tree #049, 72.32% for tree 348 349 #288, 60.90% for tree #402). Nevertheless, when captured in both replicates, the same alleles were 350 almost systematically correctly retrieved (Table 3A) (99.9% similarity). When considering all sites 351 (polymorphic sites and monomorphic sites covered by at least by 20X), the percentage of genotype 352 similarity among replicates was 99.86%, 99.65% and 99.27% for tree #049, tree #288 and tree #402, 353 respectively. As expected, decreasing the intra-individual minimum allele frequency (MAF) for SNP 354 detection from 30% to 10% increased the number of SNPs detected. This also made it possible to 355 increase the proportion of targeted SNPs detected for all samples (80% to 84% for tree #049, 60% to 356 68% for tree #402 and 72% to 78% for tree #288). Again, when variants were detected in both 357 replicates, allele similarity was maintained (99.9%). For all samples, sequencing depth exceeded 10X 358 for most of the SNPs detected in only one of the two replicates (Table 3B). We conclude that the 359 individuals were monomorphic at these loci. However, increasing the sequencing depth threshold 360 from 10X to 20X should significantly increase the number of SNPs detected in both replicates.

Finally, we were also able to test for SNP reproducibility, as 25 SNPs identified by SNP calling were 361 362 included in an earlier SNP scoring method used in a previous study of the same trees (Truffaut et al., 363 2017). Indeed, the 278 adult oak trees of La Petite Charnie had already been scored for 82 SNPs for a parentage analysis, with a MassARRAY® System 16 and iPLEX® 17 chemistry (Agena Bioscience, 364 365 San Diego, CA, USA) and 25 of these SNPs were also used in this study. The SNPs identified by the two methods were similar for the two methods except for two trees, for which differences were 366 367 observed at multiple SNPs. We suspect that these differences result from labeling errors, given that 368 the two analyses were conducted three years apart, with different DNA extracts. These two trees were 369 therefore removed from subsequent analyses. A total of 25 SNPs and 250 individuals was scored 370 with both methods (sequence capture and sequenome) giving two sets of 6,250 genotypes. We thus 371 compared the two sets, and over the 6,250 repeated genotypes, 97.67% was concordant (i.e. similar) 372 between the two methods.

374 3.1.6 Transferability

375 We studied the transferability of the targeted sequence capture technology to other species, by 376 including two cork oak (Q. suber) and two beech (F. sylvatica) samples in our study. An alignment 377 of cork oak reads against the 3P oak genome showed a significant level of target enrichment: on 378 average, for both samples, 15.86% of the reads captured 92.07% (i.e. 14,283 and 14,217) of the target 379 sequences (Table 4). When captured, target sequences were covered over 87.18% of their length on 380 average, and the mean depth of coverage over the two samples was 56.03X. Lower values were 381 obtained for the two beech specimens. On average, 8.93% of the reads captured 70.63% (i.e. 10,851 382 and 11,014) of the targeted sequences. Length coverage was only 51,60%, and sequencing coverage 383 was significantly lower, at 26.30X.

When considering *Q. robur* and *Q. petraea* trees only (*i.e.* 293 trees), we identified 13,219 SNPs per sample, on average (Table 4). Smaller numbers of SNPs were detected in the other two species: 9,093 and 3,000 SNPs in cork oak and beech, respectively.

When considering all 297 trees studied here (including 2 *Q. suber* and 2 *F. sylvatica* genotypes), we identified a total of 191,281 polymorphic sites heterozygous in at least one of these trees (Figure 4). In total, 177,232 polymorphic sites were identified in *Q. robur* and *Q. petraea*, and 13,354 and 4,295 sites were identified in *Q. suber* and *F. sylvatica*, respectively. A set of 36 SNPs was found to be common to all three species, as 10,181 SNPs were specific to *Q. suber* (*i.e.* 76% of the *Q. suber* SNPs) and 3,836 SNPs were specific to *F. sylvatica* (*i.e.* 89.31% of the *F. sylvatica* SNPs). As expected, more SNPs were shared between the *Quercus* sp. than between *Quercus* and *Fagus*.

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395 3.2 Population genetics in the Petite Charnie forest stand

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397 3.2.1 Species assignment and interspecific differentiation

According to fastSTRUCTURE, the most probable number of clusters was 2, consistent with the findings of a previous analysis performed on oak trees in the same forest, with 82 SNPs (Truffaut et al., 2017). Individual trees were assigned to the two species according to the value of the admixture coefficient (q) obtained with fastSTRUCTURE software. Trees were assigned to three groups on the basis of threshold values of q: *Q. petraea* purebreds (q \geq 0.9), admixed trees (q 0.1–0.9) and *Q. robur*

403 purebreds ($q \le 0.1$), as described in Truffaut et al. (2017). The results of the fastSTRUCTURE 404 assignment were similar to of the published results obtained with STRUCTURE (Truffaut et al. 2017), except for two individuals assigned to Q. robur by Truffaut et al. but considered admixed in 405 406 our study. These two trees had admixture values very close to the q threshold values in study of 407 Truffaut et al. (2017). Population maf values and heterozygosity distribution within species are 408 presented in Supplemental file 6. Of the 45,429 SNPs detected in Q. petraea and the 51,886 SNPs 409 detected in Q. robur, 21,331 were common to these two species. F_{st} values for all the 21,331 markers common to *Q. petraea* and *Q. robur* showed an L-curve distribution, with a large number of SNPs 410 displaying very low levels of interspecific differentiation (Figure 5). The mean and median F_{st} values 411 412 between the two species were 0.069 and 0.019, respectively, suggesting that these two species 413 display no clear differentiation over a large part of their genome.

414

415 3.2.2 SNP detection and filtering

416 Successive filtering steps on the 191,281 polymorphic sites resulted in various numbers of markers. 417 The final filtering step based on population maf resulted in the lowest number of markers for maf 418 =0.4 and the highest for maf = 0.01, with 1,561 to 33,131 usable markers for *Q. robur* and 1,454 to 419 32,047 for *Q. petraea*, respectively (see Supplemental file 2 for details).

420

421 3.2.3 Genomic relatedness

422 We first compared the expected relationship coefficient derived from pedigree relationships and 423 realized genomic relatedness in the two parent-offspring groups of known pedigree relationships, for 424 54 individual pairs (Figure 1). Considering only genomic relatedness estimated by genomic capture, 425 very minor differences in mean values were observed for numbers of markers between 32,500 426 markers (maf =0.01) and 1,500 markers (maf= 0.4). However, this difference in the number of 427 markers had a slight impact on precision, as the variance of the estimate was lower for larger numbers of markers (Figure 6), a finding supported by the overall distribution of relatedness between 428 429 individuals (Figure 7). Thus, the use of numerous rare alleles has no major effect on the prediction of 430 genomic relatedness. Realized genomic relatedness was slightly lower than expected, in both species (Figure 6). Conversely, when estimated with 82 SNPs only, genomic relatedness was scattered 431

432 around the expected value (Figure 6). The fact of including non-neutral markers (located in exons) in433 the SNPs sets had no impact on the genomic relatedness estimation (not shown here).

At population level, relatedness coefficients were distributed around a mean value of 0 (Figure 7), as expected, given the method used to calculate relatedness. However, we can consider overall mean genetic relatedness to be low within natural populations of *Q. petraea* and *Q. robur*. Among parents (*i.e.* excluding the 15 offspring) with a population maf=0.05, only 20 (*Q. robur*) and 40 (*Q. petraea*) pairs of trees had a genomic relatedness of more than 0.25 (expected for first-cousin relationship or half-sibs) and only three (*Q. robur*) and two (*Q. petraea*) pairs had a genomic relatedness of more than 0.5 (expected for full-sibs), among 8,151 (*Q. robur*) and 10,150 (*Q.petraea*) pairwise estimates.

441

442 3.2.4 Correlation between inbreeding and fitness related traits

443 Genomic inbreeding coefficients were estimated separately for each species from the G matrix 444 calculated with the 1% population maf threshold and markers common to the two species. Overall 445 rates of inbreeding within the two oak species were low (Supplemental file 7). However, one Q. petraea tree had a very high inbreeding value (0.58), and was discarded from the analysis. Overall, 446 447 the individuals of *Q. robur* were more inbred (mean inbreeding = 0.068, SD=0.030) than the individuals of *Q. petraea* (mean inbreeding = 0.037, SD=0.056). GLM analysis showed the number 448 449 of offspring to be significantly negatively correlated with inbreeding level in Q. petraea (coefficient=-3.62, *P*-value=6.06e-3), whereas this relationship was not significant in *Q. robur* 450 451 (coefficient=-1.81, *P*-value=0.114) (Figure 8b). There was no significant correlation between 452 genomic inbreeding and circumference at breast height (Q. petraea: coefficient=-25.39, P-value 453 =0.83; *Q. robur*: coefficient=-36.14, *P*-value =0.58 (Figure 8a)). These results was slightly modified 454 when the G matrix was computed with the markers selected with a maf threshold of 5% : the 455 relationship between number of offspring and inbreeding in Q. petraea became positive while 456 remaining non-significant. Thus, whatever the significance and sign of the relationship, inbreeding 457 depression signals were found to be very weak for both traits, within both species (Figure 8). Finally, 458 when G matrix is computed over all the individuals without subdividing by species, inbreeding had a 459 significant negative effect on both growth (coefficient = -107.41, *P*-value = 0.02) and reproductive 460 success (coefficient = -1.94, *P*-value = 0.02).

462 **4 Discussion**

463

464 4.1 Targeted sequence capture is a reliable, reproducible and transferable marker technique 465 for population genetics studies in oaks and beyond

466

467 Using targeted sequence capture, we successfully sequenced a large number of target genomic regions in a single assay. We obtained robust and reproductible target-enrichment results over several 468 469 hundred samples, despite the use of only one Q. robur individual to design the capture probes. We 470 evaluated the performance of target enrichment according to several parameters (number of captured 471 targets, number of reads on target, length of targeted sequences, sequencing depth). Two of these 472 parameters varied considerably between experiments, providing a cause of concern, at first sight, for 473 SNP detection. First, as observed in other studies (Fu et al., 2010; Albert et al., 2007) about 25% of 474 the sequencing reads mapped to the targeted regions. A low proportion of the reads would be 475 expected to be on-target for complex genomes, such as those of plants, which consist largely of 476 repetitive sequences and transposable elements (52% of the Quercus robur genome), making it 477 difficult to design highly specific capture probes. The duplicated nature of many of the genes in most 478 plants, particularly in trees (Plomion et al. 2018), adds another layer of difficulty in terms of 479 specificity. In this study, the length of sequencing reads was positively correlated with the percentage 480 of on-target sequences. As the length of sequencing reads was increased to 190 bp, the proportion of 481 reads correctly aligned with their targets increased significantly (Supplemental file 4). This may 482 reflect the relatively large size of the Agilent probes (120 bp) and the requirement of a sufficiently 483 long target sequence fragment for correct hybridization. Despite the small size of the on-target 484 fraction, it was sufficiently large to cover most of the target sequences deeply enough for the 485 detection of a very large number of polymorphic sites in specific areas of the oak genome. Second, 486 we observed significant variation in the number of sequences generated per sequencing run. The 487 number of reads generated differed by a factor of up to two, but the small number of samples pooled 488 per run (15) guaranteed that sufficient reads were produced to cover most of the target sequences 489 with a sufficient depth in all samples. However, alternative NGS sequencing platforms, such as the 490 Illumina NextSeq sequencing system (© Illumina), which is able to provide up to 400 million reads 491 per run, should be considered, as such systems would make it possible to multiplex a larger number 492 of samples per run, thereby decreasing the cost per sample analyzed.

493 Despite the variation of coverage and of the number of reads on target, this approach made it possible 494 to recover SNPs with sufficient reliability, not only in genic regions, but also in intergenic regions. 495 Indeed, even in intergenic regions, which are known to be highly redundant in plants, we managed to 496 obtain a sufficiently high sequencing depth to detect a large number of SNP markers. Intergenic 497 markers are particularly important for population genetic studies, in which they may be considered as 498 neutral regions of the genome for the formulation of hypotheses relating to genetic diversity. They 499 may also make a significant contribution to phenotypic variation. Li et al. (2012) showed that 500 intergenic regions in maize play a significant role in quantitative trait variation, particularly for the 5 501 kb window upstream of the gene. These areas are enriched in trait-associated SNPs. There are, 502 therefore, several complementary reasons for which intergenic regions should also be explored in 503 population and quantitative genetic studies. Our study provides a large resource of genic and 504 intergenic SNPs for the exploration of polymorphisms of QTLs previously identified as involved in the response to root waterlogging (Parelle et al., 2007) and bud phenology (Derory et al., 2010). 505 506 Using target sequence enrichment, we targeted not only candidate genes previously identified as 507 involved in drought resistance, response to waterlogging and bud phenology, but also sequences 508 displaying differentiation within species or populations, of displaying significant genotype-phenotype 509 or genotype-environment associations. Such large numbers of polymorphisms in these genomic areas 510 would never have been identified with other methods, such as Radseq. Unlike capture data, Radseq data display a high variability of sequencing depth across loci, thus limiting the detection of 511 polymorphic sites to genomes with a high level of coverage (Harvey et al., 2016). However, even in 512 513 areas with high coverage, Radseq data have been shown to include a much higher proportion of 514 singleton alleles, consistent with a high proportion of spurious allele calls (Harvey et al., 2016).

515

516 We studied the reproducibility of DNA capture by including three replicates in our design. We 517 performed independent DNA extractions from the same tissues, constructed independent libraries and 518 sequenced three replicates in different sequencing runs. Our assay was highly reproducible between 519 replicates, as we obtained very similar results for each metric. We also compared the sets of SNPs 520 independently detected in each replicated individual. The variability of sequencing depth between 521 runs explains the identification of non-identical sets of SNPs (60% to 80%). However, 99.9% of the 522 SNPs identified in each replicated individual were identical. For the retrieval of all polymorphic sites, 523 we would need to increase sequencing effort (i.e. coverage) at the targeted loci. Finally, we also

obtained reproducible results (in 97.67% of the cases) with another genotyping assay (mass
spectrometry), for a set of 250 trees genotyped for 25 SNPs with both technologies (Truffaut et al.,
2017).

527

528 One of the key assets of sequence capture technology is its ability to capture orthologous loci in 529 closely related species. Capture efficiency and coverage decrease with increasing divergence between species. However, it is reasonable to think that a subset of design probes remain useful at the 530 531 genus/family level, particularly for slowly evolving genes. The efficiency of sequence capture 532 between species has been studied with animals. George et al. (2011) used a target capture method 533 designed for humans on four monkey species. Despite sequence divergence of up to 4% between 534 humans and monkeys, they were able to capture 96% of the target sequences. We used targeted 535 probes designed for the Q. robur genome (Plomion et al., 2018) to recover the corresponding target 536 sequence in three other species from the Fagaceae: Q. petraea, Q. suber and F. sylvatica. Even if Q. 537 petraea and Q. robur are considered to be separate species, they belong to the same species complex 538 and can hybridize (Lepais et al., 2009). It has recently been shown that Q. petraea and Q. robur share 539 a mosaic of genes that have crossed species boundaries (Leroy et al., 2017). Logically, as divergence 540 between these two species is very limited, we expected the detection of orthologous sequences in Q. 541 *petraea* to be straightforward. We also considered a more distantly related species from the same 542 genus (Q. suber) and another species (Fagus sylvatica) from a different genus belonging to the same 543 botanical family (Fagaceae). No whole-genome sequence is yet available for Q. suber or F. sylvatica. 544 However, transcriptomic assemblies are available for both species (Lesur et al., 2015; Pereira-Leal et 545 al., 2014). These genetic resources limit the possibility of identifying markers outside the exonic 546 gene space. Without a complete reference genome, it is not possible to detect many of the 547 polymorphisms in intronic and intergenic regions. We were able to capture sequences from both 548 species, despite the use of much smaller numbers of cork oak and beech trees (2) than of Q. robur/Q. 549 petraea trees (293). It would, therefore, be reasonable to expect the detection of a much larger 550 number of markers if a larger number of genotypes was considered. As expected, we clearly showed 551 that, with increasing divergence time, the number of captured target sequences and the fraction of 552 their length captured decrease. The number of SNPs detected in Q. suber and F. sylvatica also 553 decreased with decreasing sequencing depth. Given that the probes were designed based on the 554 Quercus robur reference genome (Plomion et al., 2018), probe hybridization was less efficient for

555 cork oak and beech samples, resulting in partial hybridization, lower levels of coverage and the 556 capture of shorter sequences. Nevertheless, we were still able to detect several thousand SNPs in each 557 species.

558 Our findings demonstrate that this sequence capture assay for the targeted resequencing of oak 559 genomic regions is a cost-effective strategy for generating orthologous markers in related species of 560 the Fagaceae family in the absence of a reference genome.

561

562 **4.2** Estimation of relatedness among individuals in a wild oak population

563

We aimed to develop SNP markers for assessing genetic relatedness in natural populations, with a 564 565 view to estimating genetic parameters and breeding values in evolutionary studies. Genetic relatedness has traditionally been assessed by determining pedigree relationships over multiple 566 567 generations. However, this approach is difficult to implement for long-lived species, such as oaks, due to obvious biological and logistic constraints. We therefore attempted to estimate genomic 568 569 relatedness among trees within a single generation, to find ways of conducting quantitative evolutionary studies in the wild. The genomic relatedness estimated with a large number of 570 571 molecular markers has already been shown to be more efficient than pedigree relationships for the 572 purposes of prediction (Kardos et al., 2015). We show here that genomic capture is a promising 573 molecular technique for this purpose.

574 Our targeted sequence capture approach generated thousands of genotype-specific single-nucleotide 575 variants, making it possible to determine the relatedness between individuals with a high degree of 576 precision. The precision of relatedness estimates increased only slightly when the number of markers 577 increased above a few thousand.

The lack of variation of realized relatedness with the number of markers used (*i.e.* between maf=1% and maf=40%), beyond a few thousand markers, suggests that several thousand markers are sufficient for the estimation of relatedness between individuals and, thus, of the level of inbreeding of each individual (relative to the population). By contrast, we found that the use of a much more limited number of markers (a few tens), as in most traditional population genetic investigations,

583 results in a broad scattering of genomic relatedness values around the expected value, with a 584 tendency towards a large sampling variance. However, our results also show that the genomic 585 relatedness estimated with thousands of markers is systematically slightly lower than that predicted 586 on the basis of pedigree, at least for oaks. A similar trend has been reported for other forest trees (e.g. Bartholomé et al., 2016) and this pattern is predictable, given that allelic frequencies are estimated 587 588 from data for a population of related individuals (Kardos et al., 2015; Wang and Zhang, 2014). 589 Indeed, simulation studies showed that the difference in the proportion of identity by descent (IBD) 590 between the genomes of individuals within populations is systematically overestimated when IBD is 591 estimated on the basis of the level of homozygosity expected at population level under neutral 592 conditions (Kardos et al., 2015). In this case, we expect a systematic downward bias in the estimation 593 of relatedness.

594

4.3 Within-population inbreeding depression is weak, but differs between closely related 596 white oak species

597

598 Inbreeding depression can be defined as a decrease in fitness trait (*i.e.* survival, fertility or growth) 599 values in the most inbred individuals (Charlesworth and Willis, 2009). Studies of inbreeding 600 depression in natural populations were long characterized by the difficulty of accurately estimating inbreeding coefficients. In traditional population genetic studies, using a few tens or hundreds of 601 602 markers, heterozygosity is often used as a proxy for inbreeding and fitness values are regressed 603 against heterozygosity level to estimate inbreeding depression. This method has been shown to be 604 imprecise, partly because heterozygosity at a few loci is not necessarily correlated with inbreeding 605 (Szulkin et al., 2010 and references therein). With the thousands of markers developed in this study, it should be possible to estimate relative inbreeding levels more precisely between individuals. 606

The method we used to estimate the genomic relatedness matrix provides values for inbreeding (and relatedness) relative to population allelic frequencies. Thus, inbreeding values cannot be used to compare homozygosity levels between individuals from different species. However, when estimating inbreeding by considering all individuals (*i.e.* pure *Q. petraea*, pure *Q. robur* and admixed genotypes) to belong to the same population, individuals assigned to the species *Q. robur* tend to be

612 more inbred overall than individuals assigned to the species *Q. petraea*, and, as expected, individuals 613 classified as admixed are less inbred than "pure" individuals (data not shown).

Our results showed that, within species, growth was not affected by inbreeding depression in either species. However, reproductive success within species was characterized by weak inbreeding depression in *Q. petraea*, whereas no such trend was observed in *Q. robur*. A comparison between species showed that both growth and reproductive success were significantly lower in *Q. robur* than in *Q. petraea*. Thus, even though no particular pattern was detected within species, *Q. robur* individuals tended to be more inbred than *Q. petraea*, suggesting that *Q. robur* is more affected by inbreeding depression than *Q. petraea*.

621 As illustrated in a recent paper (Truffaut et al., 2017), the two species occupy contiguous areas in the 622 study plot (i.e. Q. petraea trees are in the north east and Q. robur trees are in the south west). Thus, 623 as individuals compete principally with the surrounding trees, competition on this plot can be 624 considered to occur principally within species. Given that both these species are outcrossers and form 625 large populations (Gerber et al., 2014), strong inbreeding depression would be expected, due to an 626 accumulation of deleterious alleles (genetic load). However, our results in this study do not support 627 this hypothesis. There are two non-exclusive interpretations for these observations. First, inbreeding 628 levels tend to be extremely low in adult trees (the trees were roughly 100 years old), probably 629 because the individuals with the highest levels of inbreeding are eliminated, by natural or human-630 mediated selection, when the stand is young. Second, a context of strong selection and competition may have reduced the inbreeding load (Agrawal, 2010; Hedrick et al., 1999; Whitlock, 2002), by 631 632 purging deleterious mutations. However, the negative correlation between fitness and inbreeding 633 level observed when considering all individuals (from both species) may reflect the spatial 634 distribution of the species in this mixed forest plot, resulting in weaker competition between 635 individuals from different species, attenuating "between species" selection. As both species display 636 differences in fitness-related traits correlated with differences in inbreeding coefficient, Q. petraea seedlings might be expected to outcompete Q. robur in mixed stands. Interestingly, these 637 expectations are supported by recent observations showing that *Q. petraea* seedlings have gained 638 639 ground over Q. robur seedlings in one generation (Truffaut et al., 2017).

Finally, inbreeding may also be limited by an even-age silvicultural regime. In such systems, in which all trees within a given plot belong to the same age class, mating between relatives, and

between parents and offspring in particular, is avoided, which is not the case in other systemsinvolving trees of different ages (Finkeldey and Ziehe, 2004).

644

645 **5** Conclusion

646 We demonstrate here that the combination of targeted sequence capture with next-generation 647 sequencing is an efficient method for studying genetic diversity, at genome scale, in natural oak 648 populations. We show here that this method is highly reproducible and can be extended to related 649 species within the Fagaceae. We also used this technique to assess realized genomic relatedness in 650 natural oak populations. We found that this method could be used to retrieve relationships predicted by pedigree relatedness. Given our results and those of previous SNP transferability studies 651 (Lepoittevin et al., 2015), we conclude that this method can be applied to a large number of white oak 652 species, but also to other more distant oak species from other botanical sections (Hubert et al., 2014) 653 654 or related genera of the Fagaceae family. We used this method to assess genetic relatedness and 655 inbreeding coefficients, but it could also be used for other purposes requiring a very large number of 656 markers, such as phylogenomics, phenotype-genotype-environment associations, and the prediction of breeding values for relevant traits. This work paves the way for evolutionary and genetic studies in 657 658 natura in long-lived tree species, such as oaks, that are difficult to study in controlled or common 659 garden conditions.

660

661 6 Data availability

Sequencing data for the 300 samples considered in this study are available in the NCBI - SRA
database under the Bioproject *PRJNA445867*. The haploid version (scaffolds) of the *Quercus robur*genome (haplome V2.3) has been deposited on the EMBL - ENA database under accession *OLKR01000000*.

666 The set of 191,281 polymorphic sites between *Q. petraea/Q. robur*, *Q. suber* and *F. sylvatica* 667 associated with each trait detailed in Table 1 is available through the EVOLTREE eLab service: 668 *http://www.evoltree.eu/index.php/snp-db*. The list of candidate genes included in the set of target 669 sequences, the 33,931 probe sequences, the description of the probes along with their transferability

670 and the analysis scripts used in our study can be found on the TreePeace website under the 671 *Publications* tab: *http://www.treepeace.fr/?page_id=1401*.

672

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682

683 8 Author contributions

IL contributed to the conception of the work, performed SNP detection, evaluated the success of the sequence capture experiment and contributed to the writing of the draft manuscript. HA analyzed the genomic relatedness between individuals and contributed to the writing of the draft manuscript. CB was responsible for library construction and sequencing. EC was involved in sampling. CP supervised the sequence capture experiment and revised the manuscript. AK contributed to the conception of the work and revised the manuscript.

690

691 9 Conflict of interest

692 The authors declare that the submitted work was not carried out in the presence of any personal,693 professional or financial relationships that could potentially be construed as a conflict of interest.

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695 10 References

- Agrawal, A. F. (2010). Ecological determinants of mutation load and inbreeding depression in
 subdivided populations. Am. Nat. 176, 111–122. doi:10.1086/653672.
 - Albert, T. J., Molla, M. N., Muzny, D. M., Nazareth, L., Wheeler, D., Song, X., et al. (2007). Direct selection of human genomic loci by microarray hybridization. *Nat. Methods* 4, 903–905. doi:10.1038/nmeth1111.
- Alberto, F. J., Derory, J., Boury, C., Frigerio, J.-M., Zimmermann, N. E., and Kremer, A. (2013).
 Imprints of natural selection along environmental gradients in phenology-related genes of
 Quercus petraea. Genetics 195, 495–512. doi:10.1534/genetics.113.153783.
- 701 Bartholomé, J., Bink, M. C., van Heerwaarden, J., Chancerel, E., Boury, C., Lesur, I., et al. (2016). Linkage and Association Mapping for Two Major Traits Used in the Maritime Pine Breeding 702 and Stem Straightness. 703 Program: Height Growth PloS One 11. e0165323. doi:10.1371/journal.pone.0165323. 704
- Benjamini, Y., and Hochberg, Y. (1995). Controlling the false discovery rate: a practical and
 powerful approach to multiple testing. Journal of the Royal Statistical Society Series B. 57,
 289–300.
- Bérénos, C., Ellis, P. A., Pilkington, J. G., and Pemberton, J. M. (2014). Estimating quantitative
 genetic parameters in wild populations: a comparison of pedigree and genomic approaches.
 Mol. Ecol. 23, 3434–3451. doi:10.1111/mec.12827.
- Castellanos, M. C., González-Martínez, S. C., and Pausas, J. G. (2015). Field heritability of a plant
 adaptation to fire in heterogeneous landscapes. Mol. Ecol. 24, 5633–5642.
 doi:10.1111/mec.13421.
- Charlesworth, D., and Willis, J. H. (2009). The genetics of inbreeding depression. Nat. Rev. Genet.
 10, 783–796. doi:10.1038/nrg2664.
- Chessel, D., Dufour A. B., and Thioulouse J. (2004). The ade4 package I: One-table methods. R
 News. 4/1, 5 10.
- Chilamakuri, C. S. R., Lorenz, S., Madoui, M.-A., Vodák, D., Sun, J., Hovig, E., et al. (2014).
 Performance comparison of four exome capture systems for deep sequencing. BMC
 Genomics 15, 449. doi:10.1186/1471-2164-15-449.
- Conner, J. K., Franks, R., and Stewart, C. (2003). Expression of additive genetic variances and
 covariances for wild radish floral traits: comparison between field and greenhouse
 environments. Evol. Int. J. Org. Evol. 57, 487–495.
- Derory, J., Scotti-Saintagne, C., Bertocchi, E., Le Dantec, L., Graignic, N., Jauffres, A., et al. (2010).
 Contrasting relationships between the diversity of candidate genes and variation of bud burst
 in natural and segregating populations of European oaks. Heredity 104, 438–448.
 doi:10.1038/hdy.2009.134.
- Fahrenkrog, A. M., Neves, L. G., Resende, M. F. R., Dervinis, C., Davenport, R., Barbazuk, W. B., et
 al. (2017). Population genomics of the eastern cottonwood (Populus deltoides). Ecol. Evol. 7,
 9426–9440. doi:10.1002/ece3.3466.
- Finkeldey, R., and Ziehe, M. (2004). Genetic implications of silvicultural regimes. For. Ecol. Manag.
 197, 231–244. doi:10.1016/j.foreco.2004.05.036.

- Fu, Y., Springer, N. M., Gerhardt, D. J., Ying, K., Yeh, C.-T., Wu, W., et al. (2010). Repeat
 subtraction-mediated sequence capture from a complex genome. Plant J. 62, 898–909.
 doi:10.1111/j.1365-313X.2010.04196.x.
- George, R. D., McVicker, G., Diederich, R., Ng, S. B., MacKenzie, A. P., Swanson, W. J., et al.
 (2011). Trans genomic capture and sequencing of primate exomes reveals new targets of positive selection. Genome Res. 21, 1686–1694. doi:10.1101/gr.121327.111.
- Gerber, S., Chadœuf, J., Gugerli, F., Lascoux, M., Buiteveld, J., Cottrell, J., et al. (2014). High rates
 of gene flow by pollen and seed in oak populations across Europe. PloS One 9, e85130.
 doi:10.1371/journal.pone.0085130.
- Guichoux, E., Garnier-Géré, P., Lagache, L., Lang, T., Boury, C., and Petit, R. J. (2013). Outlier loci
 highlight the direction of introgression in oaks. Mol. Ecol. 22, 450–462.
 doi:10.1111/mec.12125.
- Harvey, M. G., Smith, B. T., Glenn, T. C., Faircloth, B. C., and Brumfield, R. T. (2016). Sequence
 Capture versus Restriction Site-associated DNA Sequencing for Shallow Systematics. Syst.
 Biol. 65, 910-924. doi:10.1093/sysbio/syw036.
- Hedrick, null, Savolainen, null, and Karkkainen, null (1999). Factors influencing the extent of
 inbreeding depression: an example from Scots pine. Heredity 82 Pt 4, 441–450.
- Holliday, J. A., Zhou, L., Bawa, R., Zhang, M., and Oubida, R. W. (2016). Evidence for extensive
 parallelism but divergent genomic architecture of adaptation along altitudinal and latitudinal
 gradients in Populus trichocarpa. New Phytol. 209, 1240–1251. doi:10.1111/nph.13643.
- Hubert, F., Grimm, G. W., Jousselin, E., Berry, V., Franc, A., and Kremer, A. (2014). Multiple
 nuclear genes stabilize the phylogenetic backbone of the genus Quercus. Syst. Biodivers. 12,
 405–423. doi:10.1080/14772000.2014.941037.
- Kardos, M., Luikart, G., and Allendorf, F. W. (2015). Measuring individual inbreeding in the age of
 genomics: marker-based measures are better than pedigrees. Heredity 115, 63–72.
 doi:10.1038/hdy.2015.17.
- Kent, W. J. (2002). BLAT--the BLAST-like alignment tool. Genome Res. 12, 656–664.
 doi:10.1101/gr.229202. Article published online before March 2002.
- Kruuk, L. E. B. (2004). Estimating genetic parameters in natural populations using the "animal model." Philos. Trans. R. Soc. Lond. B. Biol. Sci. 359, 873–890. doi:10.1098/rstb.2003.1437.
- 763 Kruuk, L. E. B., and Hill, W. G. (2008). Introduction. Evolutionary dynamics of wild populations: long-term 764 the use of pedigree data. Proc. Biol. Sci. 275. 593-596. doi:10.1098/rspb.2007.1689. 765
- Le Provost, G., Lesur, I., Lalanne, C., Da Silva, C., Labadie, K., Aury, J. M., et al. (2016).
 Implication of the suberin pathway in adaptation to waterlogging and hypertrophied lenticels
 formation in pedunculate oak (Quercus robur L.). Tree Physiol. 36, 1330–1342.
 doi:10.1093/treephys/tpw056.
- Lepais, O., Petit, R. J., Guichoux, E., Lavabre, J. E., Alberto, F., Kremer, A., et al. (2009). Species
 relative abundance and direction of introgression in oaks. Mol. Ecol. 18, 2228–2242.
 doi:10.1111/j.1365-294X.2009.04137.x.
- Lepoittevin, C., Bodénès, C., Chancerel, E., Villate, L., Lang, T., Lesur, I., et al. (2015). Single nucleotide polymorphism discovery and validation in high-density SNP array for genetic

- analysis in European white oaks. Mol. Ecol. Resour. 15, 1446–1459. doi:10.1111/17550998.12407.
- Leroy, T., Roux, C., Villate, L., Bodénès, C., Romiguier, J., Paiva, J. A. P., et al. (2017). Extensive
 recent secondary contacts between four European white oak species. New Phytol. 214, 865–
 878. doi:10.1111/nph.14413.
- Lesur, I., Bechade, A., Lalanne, C., Klopp, C., Noirot, C., Leplé, J.-C., et al. (2015). A unigene set
 for European beech (Fagus sylvatica L.) and its use to decipher the molecular mechanisms
 involved in dormancy regulation. Mol. Ecol. Resour. 15, 1192–1204. doi:10.1111/17550998.12373.
- Li, H., Handsaker, B., Wysoker, A., Fennell, T., Ruan, J., Homer, N., et al. (2009). The Sequence
 Alignment/Map format and SAMtools. Bioinforma. Oxf. Engl. 25, 2078–2079.
 doi:10.1093/bioinformatics/btp352.
- Li, X., Zhu, C., Yeh, C.-T., Wu, W., Takacs, E. M., Petsch, K. A., et al. (2012). Genic and nongenic
 contributions to natural variation of quantitative traits in maize. Genome Res. 22, 2436–2444.
 doi:10.1101/gr.140277.112.
- Neves, L. G., Davis, J. M., Barbazuk, W. B., and Kirst, M. (2013). Whole-exome targeted
 sequencing of the uncharacterized pine genome. Plant J. Cell Mol. Biol. 75, 146–156.
 doi:10.1111/tpj.12193.
- Paradis, E. (2010). Pegas: an R package for population genetics with an integrated modular approach.
 Bioinformatics. 26, 419-420. doi:10.1093/bioinformatics/btp696.
- Parelle, J., Zapater, M., Scotti-Saintagne, C., Kremer, A., Jolivet, Y., Dreyer, E., et al. (2007).
 Quantitative trait loci of tolerance to waterlogging in a European oak (Quercus robur L.):
 physiological relevance and temporal effect patterns. Plant Cell Environ. 30, 422–434.
 doi:10.1111/j.1365-3040.2006.01629.x.
- Pereira-Leal, J. B., Abreu, I. A., Alabaça, C. S., Almeida, M. H., Almeida, P., Almeida, T., et al.
 (2014). A comprehensive assessment of the transcriptome of cork oak (Quercus suber)
 through EST sequencing. BMC Genomics 15, 371. doi:10.1186/1471-2164-15-371.
- Plomion, C., Aury, J.-M., Amselem, J., Alaeitabar, T., Barbe, V., Belser, C., et al. (2016). Decoding
 the oak genome: public release of sequence data, assembly, annotation and publication
 strategies. Mol. Ecol. Resour. 16, 254–265. doi:10.1111/1755-0998.12425.
 - Plomion, C., Aury, J.-M., Amselem, J., Leroy, T., Murat, F., Duplessis, S., et al. (2018) (accepted). Oak genome reveals facets of long lifespan. *Nat. Plants*.
- Purcell, S., Neale, B., Todd-Brown, K., Thomas, L., Ferreira, M. A. R., Bender, D., et al. (2007).
 PLINK: a tool set for whole-genome association and population-based linkage analyses. Am.
 J. Hum. Genet. 81, 559–575. doi:10.1086/519795.
- Raj, A., Stephens, M., and Pritchard, J. K. (2014). fastSTRUCTURE: variational inference of
 population structure in large SNP data sets. Genetics 197, 573–589.
 doi:10.1534/genetics.114.164350.
- Ritland, K. (2000). Marker-inferred relatedness as a tool for detecting heritability in nature. Mol.
 Ecol. 9, 1195–1204.

- Suren, H., Hodgins, K. A., Yeaman, S., Nurkowski, K. A., Smets, P., Rieseberg, L. H., et al. (2016).
 Exome capture from the spruce and pine giga-genomes. Mol. Ecol. Resour. 16, 1136–1146.
 doi:10.1111/1755-0998.12570.
- Szulkin, M., Bierne, N., and David, P. (2010). Heterozygosity-fitness correlations: a time for
 reappraisal. Evol. Int. J. Org. Evol. 64, 1202–1217. doi:10.1111/j.1558-5646.2010.00966.x.
 - Tennessen, J. A., Govindarajulu, R., Liston, A., and Ashman, T.-L. (2013). Targeted sequence capture provides insight into genome structure and genetics of male sterility in a gynodioecious diploid strawberry, Fragaria vesca ssp. bracteata (Rosaceae). *G3 Bethesda Md* 3, 1341–1351. doi:10.1534/g3.113.006288.
- 818 Truffaut, L., Chancerel, E., Ducousso, A., Dupouey, J. L., Badeau, V., Ehrenmann, F., et al. (2017).
 819 Fine-scale species distribution changes in a mixed oak stand over two successive generations.
 820 New Phytol. 215, 126–139. doi:10.1111/nph.14561.
- Ueno, S., Klopp, C., Leplé, J. C., Derory, J., Noirot, C., Léger, V., et al. (2013). Transcriptional
 profiling of bud dormancy induction and release in oak by next-generation sequencing. BMC
 Genomics 14, 236. doi:10.1186/1471-2164-14-236.
- Van Raden, P. M. (2008). Efficient methods to compute genomic predictions. J. Dairy Sci. 91, 4414–
 4423. doi:10.3168/jds.2007-0980.
- Wang, B., and Zhang, D. (2014). Association of allelic variation in PtoXET16A with growth and
 wood properties in Populus tomentosa. Int. J. Mol. Sci. 15, 16949–16974.
 doi:10.3390/ijms150916949.
- Whitlock, M. C. (2002). Selection, load and inbreeding depression in a large metapopulation.
 Genetics 160, 1191–1202.
- Wigginton, J. E., Cutler, D. J., and Abecasis, G. R. (2005). A note on exact tests of Hardy-Weinberg
 equilibrium. Am. J. Hum. Genet. 76, 887–893. doi:10.1086/429864.
- Wimmer, V., Albrecht, T., Auinger, H.-J., and Schön, C.-C. (2012). synbreed: a framework for the
 analysis of genomic prediction data using R. Bioinforma. Oxf. Engl. 28, 2086–2087.
 doi:10.1093/bioinformatics/bts335.
 - Zhou, L., and Holliday, J. A. (2012). Targeted enrichment of the black cottonwood (Populus trichocarpa) gene space using sequence capture. *BMC Genomics* 13, 703. doi:10.1186/1471-2164-13-703.
- 836
- 837 11 Tables and Figures
- 838
- 839 11.1 Figure legends

840

Figure 1: Pedigree relationships between *Q. robur* and *Q. petraea* siblings in the sequence
capture experiment. A Pedigree relationships between 8 *Q. petraea* siblings and their parents. B
Pedigree relationships between 7 *Q. robur* siblings and their parents. Rectangles and ellipses

correspond to siblings and parents, respectively. Numbers connecting trees correspond to expected
 genetic relatedness between individuals, based on their known pedigree.

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Figure 2: Distribution of SNPs in target sequences. A Distribution of SNPs in target sequences. B
Distribution of SNPs in target sequences based on sequence types.

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Figure 3: Correlation between sequencing depth and genomic capture efficiency parameters: A
Mean number of reads aligned with target sequences, B mean number of SNPs per sample, C number
of captured target sequences, D target sequence length.

853

Figure 4: Inter-specific transferability of SNPs. Venn diagram showing the distribution of 191,281
polymorphic sites between *Q. petraea/Q. robur*, *Q. suber* and *F. sylvatica*.

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Figure 5: Distribution of Fst between Q. petraea and Q. robur over 21,331 markers. The 21,331
SNPs correspond to a set of markers common to *Q. petraea* and *Q. robur* (also considering those
fixed within populations).

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861 Figure 6: Comparison of expected (pedigree-based) and realized genomic relatedness for different marker sets. Expected relatedness based on pedigree relationship is illustrated in Figure 1, 862 863 and shown on this graph by bold black horizontal lines. Coloured box plots correspond to realized genomic relatedness, as determined with different subsets of SNPs screened according to different 864 thresholds of minimum allele frequency (maf). The pink large-range box plots corresponds to the 865 realized genomic relatedness obtained with the 82 SNPs in the Sequenom assay (see text). The 866 number of pairwise relatedness estimates for each expected relatedness category are as follows 867 : Q.petraea $n_{0.25}=14$, $n_{0.5}=18$, $n_1=1$; Q.robur $n_{0.25}=6$, $n_{0.5}=16$, $n_1=0$. The expected relatedness 868 coefficients are extracted from Figure 1. 869

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Figure 7: Distribution of genomic relatedness between *Q. petraea* and *Q. robur* trees of the Petite Charnie forest.

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Figure 8: Correlation between genomic inbreeding and growth (a) or reproductive success
(b).The solid curves correspond to the regression of growth or reproductive success against
inbreeding coefficient according to the estimated regression coefficients. The doted lines correspond
to the 95% confidence interval of the regressions.



Figure 1



Figure 2



Sequencing depth







Figure 5



Figur₈₄6



Figure₈z



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- 889 **11.2 Tables**
- 890

891 Table 1: List of candidate target sequences selected before probe construction. Eight sets of 892 sequences were selected in total (see Materials and methods): five were reported in earlier studies 893 (Alberto et al., 2013; Guichoux et al., 2013; Le Provost et al., 2016; Ueno et al., 2013), two are 894 unpublished and the last one is provided here.

- 895
- **Table 2: Number of probes and target sequences in intergenic and genic regions.** A total of
- 897 33,931 120 bp probes were designed to capture 15,477 target sequences.
- 898

Table 3: Statistics of the replicated samples. A Comparison of the number of SNPs detected for the three replicated samples. **B** For each genotype (tree #049, tree #402, tree #288), the number of polymorphic and monomorphic sites detected for each replicate (rep #1 and rep #2) was compared with the total number of sites found to be polymorphic in at least one replicate. Htz is the number of polymorphic sites; Hmz is the number of monomorphic sites; NA is the number of sites for which sequencing depth was insufficient for the detection of polymorphism.

905

906 Table 4: Interspecific transferability statistics. For each species, the values for the percentage of 907 reads on target per tree, the percentage of captured sequences per tree, the percentage length in 908 captured sequences per tree, the sequencing depth per tree and the number of SNPs per tree are provided. For Q. petraea and Q. robur, 293 trees (adults and siblings) from the mixed oak stand 909 910 located in the Petite Charnie State Forest were considered. Q. suber data were obtained from two 911 adult trees located at the INRA Research Station at Pierroton. F. sylvatica data were obtained from 912 two adult trees located in St Symphorien. In total, 15,477 target sequences selected in the 3P Q. 913 robur genome were considered for all species.

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Set of sequences	Selection criteria of target sequences	Phenotypic or environmental variation	Number of candidate target sequences	Reference
1	Species divergence	unknown	17	Guichoux et al., 2013
2	Species divergence	unknown	1,560	Leroy et al., 2018
co	Genotype-Phenotype association	Time of leaf unfolding	681	Unpublished
4	Genotype-Phenotype association	Time of leaf unfolding	40	Alberto et al., 2013
2	Genotype-Environment association	Temperature	740	Unpublished
9	Differential expression	Response to waterlogging	4,694	Le Provost et al., 2016
7	Differential expression	Dormancy	6,069	Ueno et al., 2013
8	None (intergenic regions)	Unknown	1,822	This study
Total			15,623	

Table 1

Sequence type	Number of probes	Number of targets
Intergenic region	10,227	4,031
Genic region	23,704	11,446
Exon	11.120	4.960
Intron	6.731	2.991
Intron-exon junction	5,853	3,495
Total	33,931	15,477

Table 2

Run ID	Tree ID	Number of captured target	captured length (%)	depth (X)	SNPs	common SNPs	identical alleles	different alleles
υc	049	15,030 (97.11%)	95.78 06.62	137	13,804	12,422	12,417	5
r —	402	15,053 (97.26%)	95.62	47 47	14,291	010.01	00001	÷
Я	402	15,069 (97.36%)	96.23	124	16,318	10,643	T0,032	Ŧ
J	288	14,884 (96.17%)	94.62	67	12,431	10 000	10000	0
Я	288	15,038 (97.16%)	95.90	158	13,561	TU,3U0	TO'AOO	0
в								
Tree ID	Number	r of ic sites	repeat 1			repeat	2	F
	(* 6	Htz	Hmz	NA	Htz	Hmz	NA	I
049	15,4	13,804	1,658	0	14,080	1,382	0	f
402	17,8	305 12,330	4,765	710	16,318	1,486	1	
288	15,0	084 12,431	1,951	702	13,561	1,520	3	

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Species	reads ON target (%)	captured sequences (%)	captured length (%)	sequencing depth (X)	number of SNPs
Q. petraea Q. robur	25.20	97.19	95.82	98.24	13,219
Q. suber	15.86	92.07	87.18	56.03	9,093
E sylvatica	8.93	70.63	51.60	26.30	3,000