



**ANNALS of
SILVICULTURAL RESEARCH**

Log traceability and supply-chain verification by DNA markers in *Fagus sylvatica* L. in Italy

Journal:	<i>Annals of Silvicultural Research</i>
Manuscript ID	ASR-2022-0009.R1
Manuscript Type:	Research Paper
Date Submitted by the Author:	05-Jul-2022
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Keywords:	traceability, <i>Fagus sylvatica</i> , molecular markers

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Abstract

Verifying the match of a timber log with its population of origin represents a meaningful improvement for the preservation and valorisation of the wood chain.

In this study, we tested the applicability of the molecular approach for tracing the supply chain of timber of European beech (*Fagus sylvatica* L.) to check if the declared source of timber truly complied with the declared origin.

Samples were collected in four sawmills and in the declared forest standing populations. Different assignment tests for the “unknown” timber were used. Ordination pattern and STRUCTURE analyses identified three clusters and one case of mismatch between the analysed wood and the declared source. The Bayesian method (GeneClass2) and the machine learning approach (AssignPop) assigned 42-72% of the cases to the proper source, with higher uncertainty in one site.

This study permitted to detect the lack of congruence of the declared wood source with the actual logging site by DNA markers, screening sampling reference populations (standing trees) and timber from sawmills. Thus, the application of DNA markers confirmed to be able to find failure in the wood supply chain. This application could act as a benchmark to further develop an independent supply-chain verification system that should be implemented as part of the inventory process.

Keywords: Traceability, *Fagus sylvatica*, molecular markers, value-chain.

Introduction

Timber certification consists of a statement (certificate) attesting to the origin of raw wood material and its status and/or qualifications. Certification can serve different purposes. It can be used to demonstrate compliance of timber with standards related to forest management or environmental issues. Furthermore, it can provide information about the forest of origin, which is not generally disclosed by the producer or manufacturer (Lowe and Cross 2011). This second use case requires several control points

covering the supply chain of domestic and export markets and has implications for the process's applicability. Controlling the origin of wood is frequently one of the primary requirements for avoiding trading timber that was illegally logged. A related problem is the certification of timber logged outside a registered concession or within a protected area (Lawson and MacFaul 2010).

Illegal logging and trade in illegal timber and wood products cause economic and ecological problems, as they create market disadvantages for products sourced from legal and sustainable forestry. Verifying the origin of a timber log with its declared population of origin would have positive consequences in terms of preserving and valorising the wood chain.

Isotopic methods can be applied to differentiate wood samples originating from different geographical origins, since wood biomass incorporates elements (mainly carbon, hydrogen, oxygen and nitrogen) with specific signatures based on the environment where it was formed. In practice, this approach can be impacted by the variability of local environmental conditions (Dormontt et al. 2015), such as altitude, distance from the shoreline, rainfall, soil processes, geology, etc. An individual, absolute isotopic “fingerprint” is difficult to detect, and it is possible that samples from different geographical origins could show similar isotope discrimination. There could also be cases that result in a discrepancy between individuals from the same population or even between different tissues from the same individual (Lowe and Cross 2011, Dormontt et al. 2015, Vlam et al. 2018).

Genetic methods based on the analysis of shared ancestry or frequency of alleles can provide information on the geographic region of origin within species through population genetics or phylogeographic analyses. Fingerprinting can even identify individuals or associate seized wood material to its stump (Dormontt et al. 2015). The ability of population genetic approaches and phylogeographic analyses to determine the geographic provenance of trees is due to the existence of a spatial genetic structure within species, which can usually be found at different spatial scales. In this regard, molecular markers, such as chloroplast DNA (cpDNA) and nuclear microsatellite (nSSR) markers and RAPD (random amplified polymorphic DNA) can identify the geographical structure of populations (Deguilloux et al. 2002, Degen et al. 2013). Additionally, species-specific nSSRs can be used to generate DNA profiling databases for individual/species

identification (Jolivet and Degen 2012, Nithaniyal et al. 2014). Most of the studies have been focussed on the timber of tropical South America, Africa and Malaysia (Degen et al. 2001, Tnah et al. 2009, Tnah et al. 2010, Degen et al. 2017, Chaves et al. 2018, Finch et al. 2020, Honorio Coronado et al. 2020). Furthermore, tracing single individual and species identification is a frequent issue that has been solved by DNA fingerprinting and DNA barcoding (Degen et al. 2013, Nithaniyal et al. 2014). At the European level, significant efforts have been made for broadleaf species, such as the oaks *Quercus robur* L. and *Q. petraea* (Matt.) Liebl (Blanc-Jolivet and Liesebach 2015) - but also other oak species in temperate areas of the globe (Schroeder et al. 2016) - and ash (*Fraxinus excelsior* L.); (Tereba et al. 2017).

In this study, we aimed to test the applicability of the molecular approach for tracing the supply chain of timber of European beech (*Fagus sylvatica* L.) to check if the claimed source of timber complied with the real origin. Our main goal was to develop a three-step method that could be fast, reliable, and easy to be applied:

- (i) DNA extraction from samples (logs or boards) with a declared origin, and from their reference samples (cores or leaves) from standing tree populations, tested with different extraction protocols;
- (ii) nSSR amplification and fragment analysis;
- (iii) genotype assignment.

To our knowledge, this is the first case study in Italy aiming at identifying and tracing logs of commercial interest by a molecular marker technique on a species of wide economic importance.

Material and methods

Collection of raw material

Samples were collected in four sawmills and in the corresponding forest standing populations declared as their source, used as reference.

The standing reference populations were in four geographical regions in Italy (Fig. 1; Tab. S1).

[Here the Fig.1]

Two populations were in the Alps in northern Italy (Val di Susa, VDS, number of samples n=23, and Sella Nevea, SNV, n=21). One population was in the central Apennines (Simbruini, SIM, n=31), and another was in the southern Apennines (Serra San Bruno, SSB, n=34). Sampled material consisted of young leaves or wood microcores collected from adult trees at least 100 m apart. The samples were immediately placed in individual bags with silica gel pearls and stored until DNA extraction.

Samples collected in sawmills consisted in wood cores extracted from boards (or logs). According to the declared source, we named the samples collected in the sawmills as VDSt (n = 28), SNVt (n = 25), SIMt (n = 14), SSBt (n = 26). The differences in sample size were due to the availability of wood material and to the effective samples that were amplified.

DNA extraction

Extraction from leaves was performed on dry material using a commercial kit and the protocol recommended by the provider (DNeasy Plant mini kit ®, Qiagen).

Extraction from wood was performed testing two commercial kits:

- (i) DNeasy Plant mini kit ® (Qiagen) with the addition of Polyvinylpyrrolidone (PVP) 2.5% w/v in the AP1 buffer and an extended incubation time of 1 hour;
- (ii) DNeasy PowerPlant ® Pro kit (Qiagen).

Extraction yield was assessed spectrophotometrically (Eppendorf Biophotometer®) and quality was assessed by the ratio of absorbances at 260 nm and 280 nm.

Marker selection and laboratory analyses

Microsatellite markers were selected among the most informative ones available in the literature (Weising and Gardner 1999, Pastorelli et al. 2003, Sebastiani et al. 2004), taking into consideration the length of the amplicon (Tab. 1).

[Here the Tab.1]

Genomic DNA was amplified using a master mix designed for multiplexed microsatellites (Type-it[®] Microsatellite PCR Kit, Qiagen) with the following protocol: 2X Type-it Multiplex PCR Master mix, 10x primer mix (2 mM of each primer), DNA 15 ng, with water added up to a total volume of 25 ml. Amplification was performed in a thermal cycler (Eppendorf Mastercycler pro) under the following conditions: 95 °C for 5 minutes; 3-step cycling (28 for leaves, 32 for wood) at 95 °C for 30 seconds, 60 °C for 1.5 minutes, and 72 °C for 30 seconds; final elongation 60 °C for 30 minutes. Two fluorescently labelled primer (Life Technologies) mixes were prepared for nSSR: Mix 1 (FS1-15 (VIC[®]), FS3-04 (PET), FCM5 (FAM)) and Mix 2 (FS1-03 (NED), FS1-11 (VIC[®]), FS4-46 (FAM)).

PCR products were diluted to a concentration of 1:60, and 1 ml of each sample was added to a mixture of 15 ml Hi-Di[®] Formamide (Applied Biosystems) and 0.2 ml size standard 500 LIZ[®] (Applied Biosystems). After denaturation at 95 °C for 5 minutes, samples were run in a capillary sequencer (ABI PRISM 310, Life Technologies). Results were analysed by GeneMapper[®] Software (Applied Biosystems), and allelic profiles were scored by automatic binning and visual checking.

Statistical analyses

DNA extraction protocols comparison

Differences in the yields and quality of DNA extracted from wood between the two commercial kits with the modified protocols, were tested with a paired z-test.

Genetic diversity and markers' evaluation

The numbers of observed (NA) and effective (NE) alleles as well as expected heterozygosity (HE) (Nei 1973) were calculated using GeneAIEx v6.5 (Peakall and Smouse 2012, Peakall and Smouse 2006). The fixation index (F) at each locus was obtained by computing a hierarchical AMOVA using Arlequin v3.5.2.2 (Excoffier and Lischer 2010), and its statistical significance was determined by a non-parametric approach using 1,000 permutations Tab. 1). Null allele frequency (f_{na} , Tab. 1) for each locus was estimated via FreeNA (Chapuis and Estoup 2007).

We assessed the statistical power of our markers to detect genetic differentiation among populations given the observed genetic diversity and sample sizes using the program

POWSIM v4.1 (Ryman and Palm 2006). POWSIM simulates genetic drift in populations of effective size N_e over a specified number of generations t to produce a level of genetic differentiation described by:

$$F_{st} = 1 - (1 - 1/2N_e)^t \quad (\text{eq. 1})$$

The given F_{ST} (resulting from the AMOVA, Tab. S2) may be obtained through many combinations of N_e and t , with care to maintain overall allele frequencies reasonably similar to those of the base population, and to prevent excessive loss of low frequency alleles. We set the program with $N_e = 800$, $t = 50$. This combination of N_e and t was also selected to simulate an infinitely large base population drifting over a prolonged period. Each set of populations is then sampled, a test for genetic homogeneity is applied, and the proportion of significant outcomes (χ^2 and Fisher test, $p < 0.05$) represents the estimate of power (ranging from low power 0 to the maximum of 1) the probability of detecting the minimum amount of genetic differentiation) under the actual conditions. Simulation scenarios were completed with 999 replications.

Ordination pattern and genotype assignment

To delineate the major ordination pattern of the reference material collected from the standing populations and the unknown material using a multivariate approach, a Discriminant Analysis of Principal Components (DAPC) was performed using the package *adegenet* (v2.1.3) in R (v4.0, R Development Core Team 2020). The method transforms the genetic data into principal components and then uses k-means clustering to define groups of individuals to minimise within-group variation while maximising among-group variation. We kept axes for principal component analysis that explained about 90% of the variation, and k was selected based on the Bayesian Information Criterion (BIC).

Genetic assignment methods aim to assign or exclude reference populations as possible origins of individuals. Thus, they can infer whether wood samples from sawmills did or did not originate from the reputed forest concession.

Individual assignment tests were performed using the Bayesian approach in *STRUCTURE* (Pritchard et al. 2000) and *GeneClass2* (Piry et al. 2004). A machine-learning function implemented in the R package *assignPOP* (v1.2; Chen et al. 2016) was also applied.

STRUCTURE was used to infer the populations' genetic structure and assign each individual to a population using the origin of the reference samples to assist clustering. Twenty independent runs were performed by setting the number of groups between 1 and 10, each run consisting of a burn-in period of 10,000 iterations followed by 100,000 Monte Carlo Markov Chain (MCMC) iterations, assuming an admixture model and correlated allele frequencies. The most likely number of groups (K) was chosen based on the ΔK statistic, analysed by the Evanno method (Evanno et al. 2005), using the Structure Harvester (Earl and vonHoldt 2012). The runs were averaged with CLUMPP v1.1.2 (Jakobsson and Rosenberg 2007) and graphically represented using DISTRUCT (Rosenberg 2003)

GENECLASS2 uses jack-knifing (leave-one-out) as its default method to assign individuals to the source population. The simulation algorithm generates population samples of the same size as the reference population sample. The assignment criterion is then computed for each individual from the newly simulated population minus itself (leave-one-out procedure). The program iterates until the total number of simulated assignment criterion values is reached. We set an assignment threshold score of 0.05. Computation was executed according to the Bayesian method (Rannala and Mountain 1997, Cornuet et al. 1999), enabling probability computation through a resampling algorithm (1,000 simulated individuals).

In AssignPOP, a set of individuals from source populations were used as the training dataset, with the remaining individuals allocated to the test dataset. Genetic training data are reduced in dimensionality using PCA, which is used to build a user-chosen machine-learning classification function. The classification function is used to assign “test” individuals to source populations. This entire process was automatically repeated 1,000 times by a resampling cross-validation procedure (Monte Carlo) to validate the baseline reference data. When the predictive model built from the baseline was satisfactory, it was applied to perform the assignment test on unknown individuals.

Results

DNA extraction and extraction protocols comparison

Comparison of the two commercial kits used for DNA extraction from wood cores showed an average concentration of 22.2 ng/ μ l for samples extracted with the Dneasy

PowerPlant Pro kit and 3.9 ng/ μ l for samples extracted with the Dneasy Plant Mini kit (z value 7.85 versus a critical z value of ± 1.95). No significant difference was detected in the quality of the DNA extracted – as determined by the ratio of absorbances 260/280 – with an average value around 1.2 for Dneasy extracted samples, and around 1 for Power plant extracted samples. A z value of -2.01 was obtained versus a critical z value of ± 1.95 .

Genetic diversity

Based on the analysis of the standing tree populations (reference material), a total of 106 alleles were observed across the 6 loci (Tab. 1). The number of alleles per locus varied from 4 to 24. The frequency of the null allele was on average 10% over all loci. High values of expected heterozygosity were observed for each locus, while the fixation index was very low.

The POWSIM power analysis (Tab. 2) showed that simulated datasets with the same number of markers (six) and comparable genetic diversity (106 loci) would be able to detect significant F_{ST} values as low as the true $F_{ST} = 0.03$ (Tab. S2), as resulted from power = 1, $p < 0.05$, in both χ^2 and Fisher tests (Tab. 2). When the drift steps were omitted ($t = 0$), the power when assuming divergence among populations was 0.036 (χ^2 test) and 0.051 (Fisher test) (Tab. 2).

[Here Tab. 2]

Discriminant analysis of principal components (DAPC)

The cluster analysis of principal components (DAPC; Fig. 2) of the samples derived from both the source sites (standing trees) and the sawmills revealed four main clusters.

[Here Fig.2]

One cluster constituted the VDS site and sawn material VDS_t, together with the reference samples from SNV; another two clusters were formed by the material from SIM along with SIM_t (site and sawmill), and SSB together with SSB_t (site and sawmill), respectively; the sawn material from SNV_t formed a single group.

Genotype assignment

Individual assignment by Bayesian methods

The STRUCTURE analysis showed $k = 3$ as the most likely clustering (Fig. 3). One cluster consisted of the two northern sites VDS and SNV, with their respective sawn material, VDS_t and SNV_t. The second cluster accounted for the central Italy site SIM and its relative sawn material SIM_t. The third cluster constituted the southern SSB site and its sawn material SSB_t. In SNV_t in particular, a high degree of mixture of all gene pools could be observed.

[Here Fig.3]

GENECLASS2 assignment (Tab. 3(A)) allocated 72% of the samples from SIM_t to the SIM reference material. The correct assignment rate was medium for VDS_t and SSB_t timber: 42% and 49%, respectively assigned to VDS and SSB. Only 8% of the samples of SNV_t were assigned to SNV.

Individual assignment by machine learning

The assignment in assignPOP (Tab. 3(B)) resulted in a high to medium percentage of association of the “unknown” individuals (sawn material) from SIM_t, VDS_t and SSB_t with the proper provenance SIM (54%), VDS (48%) and SSB (46%), respectively. For SNV_t sawn material there was a low rate of correct assignment (28%), with an almost equal probability of assignment to the other origins.

[Here Tab. 3]

Discussion

This work aimed to develop a reliable method to genetically verify the congruence of the declared wood source with the actual logging site (reference sample). We used DNA markers to trace timber along the wood supply chain, sampling reference populations (standing trees) and timber from sawmills.

Molecular markers are useful in timber tracking, as the high levels of polymorphism can be used to sufficiently differentiate individuals and provide enough information for

identifying population origin (Lowe et al. 2010, Ogden and Linacre 2015). Generally, microsatellites with amplified fragments usually not larger than 500 bps, as those used in the current study, can be successfully used in this analysis (Rachmayanti et al. 2006, Tani et al. 2003). Despite the capability of molecular markers in tracing timber, DNA extraction from wood material, both processed and unprocessed, is challenging for different reasons: a small amount of genetic material is obtained; the DNA extracted from wood can be excessively fragmented; and the DNA is highly contaminated by polymerase inhibitors (Finkeldey et al. 2010). Protocols for the extraction from vegetable samples have been tested to extract good quality and adequate amount of DNA from wood. The use of a commercial kit with a standard extraction protocol can decrease the risk of human error and can help in certifying the methodology. Nevertheless, the usual addition of PVP to the standard procedure generally brings satisfactory improvements in terms of DNA yields per gram of dried wood sample (Rachmayanti et al. 2009, 2006, Shabrina et al. 2019), as PVP can effectively reduce the amount of PCR inhibitory substances (e.g., cellulose and lignin). Thus far, however, no single method can be applied to all species, likely for reasons related to the variable biochemical composition of plant tissues of different species. The fact that the PVP addition did not enhance the DNA yield and quality in the species under investigation in this study can be ascribed to this case. Conversely, when using a kit explicitly aimed at processing difficult samples, even without additional PVP, an elevated quantity of fragments in the range expected for the selected markers was produced, resulting in an adequate substrate for further amplification.

Factors influencing the isolation success have been found to include the length of the DNA fragments, how much the wood has been processed, which part of the wood has been sampled, and the wood's age. Molecular markers used to identify timber origin must be short and simple sequences, as a negative correlation has been observed between amplicon size and amplification success (Lowe et al. 2010, Rachmayanti et al. 2009, 2006). Additionally, unprocessed wood results in higher success rates since the thermal treatments and chemicals applied in wood processing are deleterious for DNA integrity (Rachmayanti et al. 2009). Amplification success is higher if DNA is isolated from outer sapwood than older wood (transition zone between sapwood and heartwood and the inner heartwood). However, when sampling from outer parts, it is crucial to avoid surface

tissues likely contaminated by other organisms' nucleic acids (Jiao et al. 2015, Rachmayanti et al. 2006). In general, the more the material is aged, the more fragmented DNA can be extracted (Deguilloux et al. 2002, Dumolin-Lapegue et al. 1999, Jiao et al. 2015, Speirs et al. 2009, Tani et al. 2003). In this study, the unknown material consisted of recently sized logs (SIMt) or recently sized tables sampled in sawmills. Only material from SSBt was aged and sampled from splints. However, the extraction, even from this more-aged material, did not show evident issues, as usually reported in studies (Deguilloux et al. 2002, Jiao et al. 2015, 2014) dealing with ancient wood.

Typically, in natural forests, a genetic structure with a clear differentiation can be observed at local and regional spatial scales. In this study, the analysis of the sampled standing tree populations matched the expected Italian beech populations' spatial variability. In fact, we could detect an Alpine cluster and two different Apennine groups, in agreement with the picture widely reported in previous papers (Emiliani et al. 2004, Magri et al. 2006, Vettori et al. 2004), mainly caused by past glacial periods and the repeated retreat-colonisation events of tree populations (Hewitt 2004, Petit et al. 2003). The presence of this well-defined genetic structure of beech populations in Italy made it possible to properly detect the provenance of the unknown material with good to high probability in three out of four cases, with the timber from Sella Nevea (SNVt) proving the exception. We observed correct assignment probabilities ranging from 42% to 72%. These values are comparable to assignment percentages evaluated by nSSR for a wide set of species (Degen et al. 2013) ranging from 50% to 70% when working at the population level. Higher rates have been reported in studies attempting to determine the origin at the country level. It is unlikely that the observed low assignment of SNVt to the proper source is caused by the low number of markers used. In fact, for the given level of samples, differentiation and alleles, our nuclear markers had a significantly low probability of error in rejecting the null hypothesis of the lack of differentiation. Nor did the use of chloroplast markers (results not shown), which confirmed the regional differentiation, result in a more detailed assignment.

The assignment can also depend on the applied method (Degen et al. 2017). In fact, in this study, we could observe that the PCoA and STRUCTURE analyses revealed a possible overlapping of VDS_t with both VDS and SNV. Thus, it was impossible to assign

SNVt to a specific geographic location in the Alps because there is no differentiation between individuals from different reference locations in this genetic cluster, indicating gene flow. Differently, the Bayesian method (GeneClass2) and the machine learning approach (AssignPop) showed to be complementary and could allocate better unknown samples to the site of origin. Both the Bayesian method and the machine learning approach detected a higher percentage for the putative origin site for VDS_t, SIM_t and SSB_t. In the case of SNV_t, GeneClass2 allocated almost the same percentage to VDS and SSB, while AssignPop did not detect differences in the assignment distribution. According to some studies, the low success of assignment can be related to high levels of intrapopulation diversity, as the more alleles the tested population has, the higher the chances of a population having alleles in common with other populations (Chaves et al. 2018). However, this was not the case for SNV, which recorded the lowest allelic richness and genetic diversity. The low probability observed in the case of Sella Nevea might be due to an error in the indication of the provenance site, but we cannot totally exclude the lack of information caused by missing data that led to the omission of some samples in the reference population. Since for this kind of investigation, based on the probability that a genotype could belong to a specific provenance, the availability of an exhaustive and geographically widespread reference database is crucial (Cavers et al. 2005, Ogden and Linacre 2015), this uncertainty could have been overcome by extending the reference data sampled on site and to bordering areas and possibly in neighbouring countries to exclude the provenance of material from abroad.

Conclusions

This study permitted to detect the lack of congruence of the declared wood source with the actual logging site by the application of DNA markers sampling reference populations (standing trees) and timber from sawmills. Thus, the application of DNA markers confirms to be able to find failure in the wood supply chain.

Regrettably, we could not extend the analysis to other sites. For this reason, we reckon that genetic inventories with extensive and systematic samples across the whole species distribution area are the basis for identifying the country of timber origin. Higher spatial resolution is needed to control the source of timber at the level of logging concessions.

A solid traceability system should rely on different types of information. Implementing regular genetic monitoring in forest inventories would be desirable since it would integrate with collected data (e.g., dendrometric data, stand structure, etc.) and extend the reference database. Furthermore, sharing existing data on an accessible database of reference materials to enable effective tests would be a primary instrument for stakeholders involved in the certification process and help ensure a reliable and applicable certification system.

Acknowledgements

This research was funded by “PRIN bando 2015 – La filiera corta nel settore biomasse-legno: approvvigionamento, tracciabilità, certificazione e sequestro di Carbonio. Innovazioni per la bioedilizia e l'efficienza energetica”, financed by the Ministry of University and Research (MIUR) – D.M. 1826 del 20/9/2016.

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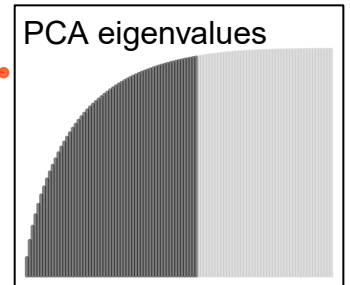
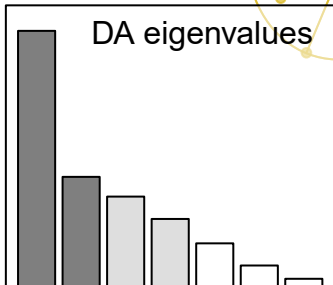
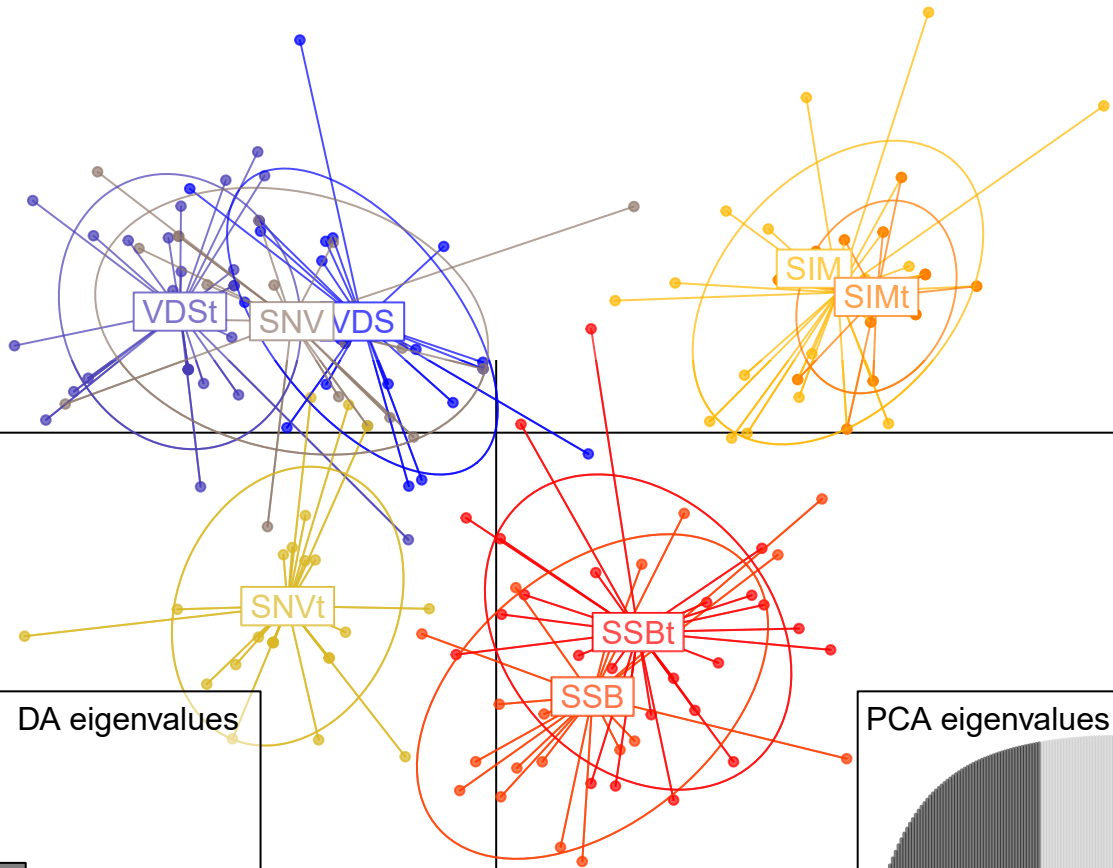
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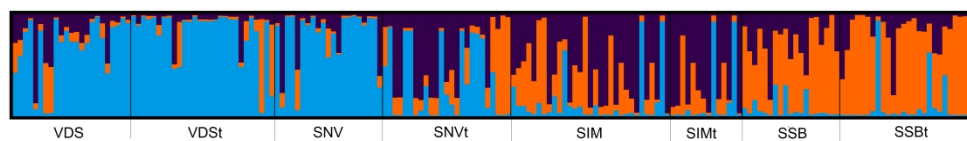


Figure 3 - Assignment per individual based on Bayesian clustering of microsatellite data using STRUCTURE. Individuals are represented by vertical bars. Colours per individual represent percentage assignment to each of the three assumed genetic clusters.

Additional material

Materials and Methods

Genetic diversity and genetic differentiation

To assess genetic diversity at the population level (reference tree populations), within- and among-population genetic parameters were evaluated by different approaches. Expected heterozygosity (H_E) (Nei 1973) were calculated using GeneAIEx v6.5 (Peakall & Smouse 2006, Peakall & Smouse 2012; Tab. S1). As an assessment of allelic richness through measuring allele frequencies must consider variation in population sizes, allelic richness (AR) and the number of private alleles (PA) were computed using the rarefaction method with HP-RARE software (Kalinowski 2005; Tab. S1). The inbreeding coefficient (F_{IS}) for each population was obtained by computing a hierarchical AMOVA using Arlequin v3.5.2.2 (Excoffier & Lischer 2010), and its statistical significance was determined by a non-parametric approach using 1000 permutations (Tab. S1; Tab. S2). Population differentiation (F_{ST}) was calculated in GeneAIEx v6.5 (Peakall & Smouse 2006, Peakall & Smouse 2012; Tab. S3).

Tab. S1 - Genetic indices computed on reference samples

label	N	AR	PA	H_E	F_{IS}
VDS	23	10.3	1.1	0.76	0.20
SNV	21	8.3	1.0	0.73	0.14
SIM	31	10.1	1.3	0.80	0.22
SSB	34	10.8	2.4	0.77	0.20

N = Total number of sampled individuals; AR = allelic richness; PA = number of private alleles with rarefaction; H_E = expected unbiased heterozygosity; F = fixation index for each sampled reference population

Tab. S2 – Analysis of molecular variance (AMOVA) of the four reference populations

Source of variation	Df	Variance components	% variation	F-statistic	
Among populations	3	0.035	2.3	F_{ST}	0.03
Among individuals within populations	105	0.292	19.1	F_{IS}	0.19
Within individuals	109	1.206	78.6	F_{IT}	0.21
Total	217	2.431	100		

Tab. S3 – Population pairwise F_{ST} . *** indicate significant differences ($p < 0.05$)

	VDS	SNV	SIM	SSB
VDS	-			
SNV	0.003	-		
SIM	0.032 ***	0.046 ***	-	
SSB	0.048 ***	0.068 ***	0.037 ***	-

Tables and captions

Table 1 - Nuclear microsatellite markers used in this study.

Locus	bp	NA	NE	f_{na}	HE	F
FS1-15	69–131	22	7.3	0.068	0.89	0.02
FS3-04	192–207	4	1.9	0.002	0.47	0.02
FCM5	260–346	23	9.1	0.147	0.93	0.03
FS1-03	68–120	21	5.0	0.147	0.82	0.01
FS1-11	92–126	12	4.4	0.166	0.81	0.07
FS4-46	170–311	24	6.8	0.109	0.90	0.05

Base pairs (bp), numbers of alleles (NA), effective numbers of alleles (NE), frequency of null allele (f_{na}), expected heterozygosity (He) and fixation index (F)

Table 2 - POWSIM power simulation results for the 6 microsatellite loci indicating proportion of statistical significance ($p < 0.05$) to detect population structure among *Fagus sylvatica* locations. t = generations to drift before sampling.

t	FST	Power probability
0	0	0.051
50	0.03	1

Table 3 – Values of A) membership probabilities calculated in GeneClass2 and B) membership probabilities (Standard Deviation) obtained by AssignPop via Monte Carlo assignment, trained using all loci and a set of individuals from source populations.

A)	VDS	SNV	SIM	SSB
VDS _t	42	16	11	31
SNV _t	34	8	27	31
SIM _t	11	2	72	15
SSB _t	16	5	30	49

B)

VDS _t	0.48 (0.11)	0.22 (0.11)	0.15 (0.12)	0.15 (0.11)
SNV _t	0.24 (0.13)	0.28 (0.13)	0.26 (0.14)	0.22 (0.12)
SIM _t	0.18 (0.10)	0.09 (0.10)	0.54 (0.12)	0.19 (0.13)
SSB _t	0.18 (0.15)	0.09 (0.09)	0.27 (0.18)	0.46 (0.19)

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