

1 **Lifetime genealogical divergence within plants leads to epigenetic mosaicism**
2 **in the long-lived shrub *Lavandula latifolia* (Lamiaceae)**

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21 Summary

- 22 • Epigenetic mosaicism is a possible source of within-plant phenotypic heterogeneity, yet
23 its frequency and developmental origin remain unexplored. This study examines whether
24 the extant epigenetic heterogeneity within long-lived *Lavandula latifolia* (Lamiaceae)
25 shrubs reflects recent epigenetic modifications experienced independently by different
26 plant parts or, alternatively, it is the cumulative outcome of a steady lifetime process.
- 27 • Leaf samples from different architectural modules were collected from three *L. latifolia*
28 plants and characterized epigenetically by global DNA cytosine methylation and
29 methylation state of methylation-sensitive amplified fragment length polymorphism
30 markers (MS-AFLP). Epigenetic characteristics of modules were then assembled with
31 information on the branching history of plants. Methods borrowed from phylogenetic
32 research were used to assess genealogical signal of extant epigenetic variation and
33 reconstruct within-plant genealogical trajectory of epigenetic traits.
- 34 • Plants were epigenetically heterogeneous, as shown by differences among modules in
35 global DNA methylation and variation in the methylation states of 6-8% of MS-AFLP
36 markers. All epigenetic features exhibited significant genealogical signal within plants.
37 Events of epigenetic divergence occurred throughout the lifespan of individuals and were
38 subsequently propagated by branch divisions.
- 39 • Internal epigenetic diversification of *L. latifolia* individuals took place steadily during
40 their development, a process which eventually led to persistent epigenetic mosaicism.

41 **Key words:** epigenetic mosaicism, DNA methylation, genealogical signal, high performance
42 liquid chromatography, *Lavandula latifolia* (wild lavender), methylation-sensitive amplified
43 fragment length polymorphism (AFLP), subindividual variation

44

45 Introduction

46 Within-plant variance in phenotypic traits of reiterated, homologous organs that perform the
47 same function (leaves, flowers, fruits, seeds) is often very high, sometimes contributing more
48 to total population-wide variance than differences among individuals (Herrera, 2009, 2017;
49 Palacio *et al.*, 2019). Depending on its magnitude and spatio-temporal patterning, this
50 subindividual phenotypic variance can have multiple ecological effects. These include
51 optimizing the exploitation of limiting resources such as light, water or nitrogen (Osada *et*
52 *al.*, 2014; Ponce-Bautista *et al.*, 2017; Mediavilla *et al.*, 2019), altering the outcome of

53 interactions with animals (Sobral *et al.*, 2013, 2014; Shimada *et al.*, 2015; Wetzzel *et al.*,
54 2016), driving selection on reproductive traits (Austen *et al.*, 2015; Dai *et al.*, 2016; Arceo-
55 Gómez *et al.*, 2017; Kulbaba *et al.*, 2017), and enhancing tolerance of environmental
56 unpredictability (Tíscar Oliver & Lucas Borja, 2010; Hidalgo *et al.*, 2016). Because of these
57 ecological effects, subindividual variability can eventually influence the fitness of
58 individuals and become itself a target for natural selection, since plants not only have
59 characteristic trait means but also characteristic trait variances and spatio-temporal patterns
60 of subindividual heterogeneity (Herrera, 2009, 2017; Kulbaba *et al.*, 2017; Harder *et al.*,
61 2019).

62 The evolutionary significance of within-plant heterogeneity in phenotypic traits of
63 reiterated structures will depend on its causal mechanisms (Herrera, 2017). Position in
64 relation to external environmental gradients (light, air temperature) or internal
65 developmental axes (nodal position on branches) are factors ordinarily contributing to
66 within-plant variation in phenotypic traits (Herrera, 2009). Subindividual polymorphisms in
67 chromosome number (aneusomaty; D'Amato, 1997) or the DNA sequence of nuclear (Wang
68 *et al.*, 2019; Orr *et al.*, 2020) and plastid genomes (García *et al.*, 2004; Sun *et al.*, 2019) can
69 also account for subindividual phenotypic heterogeneity (Whitham *et al.*, 1984). The
70 phenotypic effects of these polymorphisms, however, have been investigated on few
71 occasions. Genetic heterogeneity caused by somatic mutations is unlikely to be a pervasive
72 driver of within-plant heterogeneity in wild plants, given the paucity of well-documented
73 genetic mosaics and the extremely low somatic mutation rates reported whenever such
74 mosaics have been found (Padovan *et al.*, 2013; Ranade *et al.*, 2015; Schmid-Siegert *et al.*,
75 2017; Gerber, 2018; Wang *et al.*, 2019; Orr *et al.*, 2020). Somatic mutations altering nuclear
76 or plastid DNA sequences are not, however, the only molecular mechanism with a capacity
77 to produce genomic heterogeneity and induce phenotypic variation within individual plants.
78 Cytosine methylation is a major epigenetic mechanism in plants with roles in gene
79 expression, transposon activity, and plant growth and development (Finnegan *et al.*, 2000;
80 Cokus *et al.*, 2008; Lister *et al.*, 2008), hence subindividual heterogeneity in pattern and
81 level of DNA methylation could also partly account for within-plant variation in organ traits
82 (Herrera & Bazaga, 2013; Alonso *et al.*, 2018; Herrera *et al.*, 2019).

83 Epigenetic mosaics in which homologous organs in different parts of the same genetic
84 individual differ in extent and/or patterns of DNA methylation have been documented for
85 clonal and non-clonal plants (Bitonti *et al.*, 1996; Gao *et al.*, 2010; Bian *et al.*, 2013; Spens

86 & Douhovnikoff, 2016), and associations between phenotypic heterogeneity and
87 subindividual epigenetic variation, either natural or experimentally induced, have been also
88 found (Herrera & Bazaga, 2013; Alonso *et al.*, 2018; Herrera *et al.*, 2019). In adult
89 *Lavandula latifolia* (Lamiaceae) plants there is substantial subindividual heterogeneity in
90 global DNA cytosine methylation. Such variation is correlated with within-plant variation in
91 the number and size of seeds produced (Alonso *et al.*, 2018), which supports a causal link
92 between epigenetic mosaicism and subindividual phenotypic heterogeneity. In the perennial
93 herb *Helleborus foetidus* (Ranunculaceae), artificial augmentation of within-plant
94 heterogeneity in global DNA cytosine methylation enhanced within-plant variance in
95 phenotypic traits, as predicted by the hypothesis that epigenetic mosaicism can contribute to
96 within-plant variation (Herrera *et al.*, 2019).

97 Two different mechanisms could lead to within-plant epigenetic mosaics and
98 associated phenotypic heterogeneity of the sort documented by Alonso *et al.* (2018) for *L.*
99 *latifolia*. These mosaics could mostly reflect ephemeral epigenetic modifications
100 experienced recently by different plant parts independently of each other or, alternatively,
101 represent relatively stable somatic conditions reflecting past epigenetic changes which took
102 place at different moments in the plant's ontogeny. Under this latter scenario, subindividual
103 epigenetic heterogeneity at a given moment in a plant's lifetime should be genealogically
104 structured, representing the signature of past localized changes within the plant that were
105 propagated and maintained through successive divisions of terminal meristems. This
106 mechanism for internal epigenetic divergence by propagation of stable epimutations is
107 essentially identical to that proposed previously to account for stable subindividual genetic
108 mosaicism via propagation of somatic mutations (Whitham & Slobodchikoff, 1981; Buss,
109 1983a, b; Whitham *et al.*, 1984; Gill *et al.*, 1995; Schmid-Siegert *et al.*, 2017; Orr *et al.*,
110 2020).

111 This study evaluates the hypothesis that extant epigenetic heterogeneity within old
112 plants of the evergreen shrub *L. latifolia* is genealogically structured and can thus be
113 considered the outcome of a lifetime process of cumulative epigenetic diversification taking
114 place within plants. Leaf samples from many distinct modules from the same individual
115 plant were characterized epigenetically by measuring global DNA cytosine methylation and
116 assessing the methylation state of a large number of methylation-sensitive anonymous DNA
117 markers. Analyses were accomplished for three wild plants. By combining these data with
118 detailed information on the branching history of each plant, and then applying quantitative

119 analytical methods borrowed from phylogenetic research to assess genealogical signal and
120 reconstruct changes in epigenetic traits, we aim to assess whether extant epigenetic
121 heterogeneity among the modules of the same individual attests their past genealogical
122 trajectories within the plant.

123 **Materials and methods**

124 Study species and field methods

125 *Lavandula latifolia* Med. is a dome-shaped, long-lived evergreen shrub inhabiting the
126 undergrowth of mid-elevation woodlands in the eastern Iberian Peninsula. Branching is
127 dichasial and generally conforms to Leeuwenberg's development model (Hallé *et al.*, 1978;
128 Hallé, 1986). This branching pattern leads to crowns of adult plants being made up of
129 morphological units consisting of distinct leaf clusters borne by short stems, many of which
130 produce one terminal inflorescence in early summer (Alonso *et al.*, 2018: Fig. 1). Each of
131 these clusters of even-aged leaves borne by short terminal branchlets will be hereafter
132 termed a 'module' following Hallé's (1986) definition ('the leafy axis in which the entire
133 sequence of aerial differentiation is carried out, from the initiation of the meristem that
134 builds up the axis to the sexual differentiation of its apex'). Mean seed mass and total seed
135 production vary widely among modules of the same plant (Herrera, 1991, 2000), and this
136 variation was shown by Alonso *et al.* (2018) to be related to subindividual mosaicism in
137 global DNA cytosine methylation. Further details on the natural history, reproductive
138 biology and demography of *L. latifolia* can be found in Herrera (1991), Herrera & Jovani
139 (2010), Herrera & Bazaga (2016), and references therein.

140 Field sampling for this study was conducted on September 2017 at a large *L. latifolia*
141 population growing near Arroyo Aguaderillos in the Sierra de Cazorla (Jaén province,
142 southeastern Spain). Three of the shrubs whose leaves and seeds had been previously
143 sampled by Alonso *et al.* (2018; plants TSE03, TSE04 and TSE05) were harvested by
144 digging up their roots and brought to the laboratory. Each plant was an individual arising
145 from a single taproot. Fresh leaf samples were collected from as many modules as possible
146 of each plant within a few hours of harvest, subject to the constraint that all leaves in the
147 sampled module should be healthy and free of any visible damage. Samples were placed in
148 paper envelopes, quickly dried at ambient temperature in containers with abundant silica gel,
149 and stored dry at ambient temperature for subsequent DNA extraction. A total of 20, 38 and
150 22 leaf samples were obtained from as many modules of plants TSE03, TSE04 and TSE05,
151 respectively.

152 Laboratory methods

153 Harvested plants were 23 (TSE03), 29 (TSE04) and 27 (TSE05) years old at time of
154 collection as determined by ring counting (Herrera, 1991). A detailed map of the branching
155 architecture of each plant was drawn by hand, and tips of branchlets from which leaf samples
156 had been collected, and all internal branching nodes leading to them, were tagged and
157 numbered. The linear distances between successive branching nodes, and between the last
158 branching nodes and tips, were measured along branches. The age of every branching node
159 and terminal branchlet (module) was determined by ring counting.

160 Dried leaf material was homogenized to a fine powder using a Retsch MM 200 mill
161 and total genomic DNA was extracted from approximately 35 mg of ground leaf material
162 using the Bioline ISOLATE II Plant DNA Kit and the manufacturer protocol. Aliquots from
163 the DNA extract of each module ($N = 80$) were used for estimating genome-wide
164 methylation of DNA cytosines and obtaining epigenetic fingerprints as detailed below.

165 Genome-wide percent cytosine methylation was determined for the leaves of each
166 module using the chromatographic technique described by Alonso *et al.* (2016; see also
167 Alonso *et al.*, 2018). Genomic DNA was digested with DNA Degradase Plus™ (Zymo
168 Research, Irvine, CA), a nuclease mix that degrades DNA to its individual nucleoside
169 components. Digested samples were stored at -20°C until analyzed. Two independent
170 technical replicates of DNA hydrolyzate were prepared for each module, and the 160
171 samples (80 modules x 2 replicates) were processed in randomized order. DNA cytosine
172 methylation was determined for each sample by reversed phase HPLC with
173 spectrofluorimetric detection. Global cytosine methylation was estimated as $100 \times$
174 $5\text{mdC}/(5\text{mdC} + \text{dC})$, where 5mdC and dC are the integrated areas under the peaks for 5-
175 methyl-2'-deoxycytidine and 2'-deoxycytidine, respectively. The position of each nucleoside
176 was determined using commercially available standards (Sigma Aldrich).

177 Variation among modules of the same plant in epigenetic fingerprint was investigated
178 using a variant of the amplified fragment-length polymorphisms technique (AFLP) which
179 allowed to identify instances of within-plant polymorphism in the methylation state of
180 methylation-susceptible anonymous 5'-CCGG sequences. As we were interested in detecting
181 heterogeneity in genomic DNA methylation profiles among modules from the same
182 genotype, our AFLP method used exclusively primer combinations based on the
183 methylation-sensitive HpaII enzyme. HpaII cleaves 5'-CCGG sequences but is inactive
184 when either or both cytosines are fully methylated, and cleaving may be impaired or blocked

185 when one or both of the cytosines are hemi-methylated (McClelland *et al.*, 1994; Roberts *et*
186 *al.*, 2007). In absence of DNA sequence variation among samples, as expected for leaves
187 from the same plant, any within-plant polymorphism in these methylation-sensitive AFLP
188 markers (MS-AFLP hereafter) will reflect subindividual heterogeneity in the methylation
189 state of the associated 5'-CCGG site (see Verhoeven *et al.*, 2010; Herrera *et al.*, 2012,
190 Herrera & Bazaga, 2013; for applications of this simplified AFLP method in epigenetic
191 studies of plants and fungi). MS-AFLP analyses and fragment scoring were performed
192 following the protocols described in Herrera & Bazaga (2013, 2016). Leaf samples were
193 fingerprinted using eight primer combinations, each with two (HpaII) or three (MseI)
194 selective nucleotides, which were chosen on the basis of repeatability and ease of scoring
195 (Supporting Information Table S1). Scoring error rates were determined for each individual
196 marker by running replicated analyses for 25 leaf samples (31.2% of total), and estimated as
197 the ratio of the number of discordant scores in the two analyses to the total number of
198 replicated samples. To minimize the possibility of spurious within-plant polymorphisms
199 arising from scoring errors, only the $N = 467$ markers with scoring error rates equal to zero
200 were retained for the analyses (Supporting Information Table S1).

201 Data analysis

202 **Extant epigenetic heterogeneity** All statistical analyses reported in this paper were carried
203 out using the R environment (R Core Team, 2020). Heterogeneity in global cytosine
204 methylation among sampled plants, and among modules of the same plant, was tested by
205 fitting a linear model to the data, treating plants and modules nested within plants as fixed-
206 effect predictors. The contribution of differences between plants, and between modules
207 within plants, to total sample variance in global cytosine methylation were estimated by
208 fitting an intercept-only random effect model to the data using the lmer function of the lme4
209 package (Bates *et al.*, 2015), with plants and modules as hierarchically nested random
210 effects. Confidence intervals of variance estimates were computed using the function
211 confint.merMod in lme4.

212 For each plant, a module x MS-AFLP marker binary matrix was obtained whose
213 elements were the methylation state of each marker in the given module (1 = unmethylated;
214 0 = methylated). In each matrix, only those markers occurring unmethylated in at least one
215 module of the plant were retained for analysis ('informative markers' hereafter), because our
216 MS-AFLP procedure did not allow to discriminate between homogeneous methylation and
217 fragment absence for those markers which did not occur in any module of a plant. Another

218 binary matrix was obtained for all plants and modules combined using only those markers
219 that were simultaneously informative in all plants ($N = 400$). Multivariate analyses of within-
220 and among-plant variation in epigenetic fingerprints were then conducted on these data,
221 which included nonmetric multidimensional scaling (function metaMDS in package vegan;
222 Oksanen *et al.*, 2019) and analysis of molecular variance (adonis function in vegan) on the
223 pairwise matrix of Jaccard dissimilarity between modules.

224 **Within-plant genealogy of epigenetic heterogeneity** Two Newick-formatted genealogical
225 trees were constructed for each plant by collating the topological information from drawings
226 of branching architecture and the quantitative data on linear or temporal distances between
227 branching nodes. In these genealogical trees the modules sampled were at the tips and
228 branch lengths were either the linear distance between branching nodes or their age
229 difference in years (Supporting Information Fig. S1). Although growth- and age-based
230 branch lengths were correlated in the three plants ($R^2 = 0.57$ - 0.61), we used trees based on
231 these two different metrics to explore whether epigenetic changes taking place over a plant's
232 lifetime were best explained in terms of the amount of growth or the time elapsed between
233 successive branching nodes.

234 The dichotomous branching trees used here to depict genealogical relationships among
235 extant modules of *L. latifolia* plants are conceptually equivalent to the phylogenetic trees
236 commonly used to depict the evolution of contemporary taxa by descent with modification.
237 There is the advantageous difference that our genealogical trees are errorless pedigrees
238 representing true relationships rather than uncertain inferential hypotheses as it usually
239 happens with phylogenetic trees (Felsenstein, 2004). Methods borrowed from phylogenetic
240 research were used to investigate the genealogical component of extant within-plant
241 epigenetic mosaicism in *L. latifolia* shrubs (see Orr *et al.*, 2020, for a comparable approach).
242 'Genealogical tree', 'genealogical character estimation' and 'genealogical signal' will be
243 used hereafter as the within-plant equivalents to 'phylogenetic tree', 'ancestral character
244 estimation' and 'phylogenetic signal', respectively (Paradis, 2012; Münkemüller *et al.*,
245 2012). Assessment of genealogical signal and genealogical character estimation will be
246 consistently used throughout this paper to evaluate the genealogical basis of extant
247 epigenetic mosaicism within *L. latifolia* plants. The first approach provides a quantitative
248 assessment at the whole plant level of the association between trait similarity and proximity
249 in the genealogy, while the second will inform on the spatio-temporal patterns of epigenetic
250 changes taking place within individual plants over their lifetimes.

251 For continuous traits (global DNA cytosine methylation and coordinates of modules on
252 axes from nonmetric multidimensional scaling), genealogical character estimations were
253 carried out with function `contMap` from the `phytools` package (Revell, 2012). Genealogical
254 signal was tested with the `philoSignal` function in the `philosignal` package (Keck et al., 2016)
255 and Moran's I method, which relies on an autocorrelation approach, makes no assumptions
256 on model of change and incorporates information on branch length (Münkemüller *et al.*,
257 2012). Within-plant genealogical signal in the methylation state of individual MS-AFLP
258 markers (0 = methylated, 1 = unmethylated) was tested using Fritz & Purvis' (2010) D
259 statistic for binary traits, a measurement of character dispersion on the genealogy. In these
260 analyses, only the most informative polymorphic markers (frequency of commonest
261 methylation state < 0.85) were considered for each plant ($N = 3, 6$ and 6 markers for plants
262 TSE03, TSE04 and TSE05, respectively). Computations were conducted with function
263 `phylo.d` from the `caper` package (Orme *et al.*, 2018). Randomization tests were used to assess
264 statistical significance of differences between observed D values and expectations from
265 random ($D = 1$) and Brownian motion ($D = 0$) distributions of methylation state of individual
266 markers across tips of genealogical trees (Fritz & Purvis, 2010). Transition rates between the
267 methylated and unmethylated states of individual markers on plant genealogies were
268 explored by fitting 'Equal rates' (ER) and 'All rates different' (ARD) discrete evolution
269 models to genealogical trees using the `fitDiscrete` function in the `geiger` package (Pennell *et*
270 *al.*, 2014). Akaike information criterion (AIC) for fitted ER and ARD models were
271 compared with the `aic.w` function of the `phytools` package. The stochastic mapping
272 procedure of Bollback (2006), as implemented in function `make.simmap` of the `phytools`
273 package, was used for genealogical character estimation of the methylation state of markers
274 with genealogical signal within plants.

275 **Results**

276 Extant epigenetic heterogeneity

277 There was substantial subindividual heterogeneity in global DNA cytosine methylation
278 among the even-aged leaf samples from different modules. In addition to differences among
279 plants ($F_{2,92} = 12.18$, $P = 0.00002$), cytosine methylation also differed among modules of the
280 same plant ($F_{77,92} = 1.59$, $P = 0.017$). Estimated within-plant variance in global cytosine
281 methylation (0.119; 95% confidence interval = 0.015–0.252) was roughly comparable to
282 among-plant variance (0.100; 95% confidence interval = 0.0056–0.647), further stressing the
283 quantitative importance of within-plant variation in the extent of genomic methylation.

284 Modules from the same plant were broadly scattered on the plane defined by axes from
285 nonmetric multidimensional scaling of the pairwise dissimilarity matrix (Fig. 1), thus
286 revealing substantial within-plant epigenetic heterogeneity in the multivariate space defined
287 by those informative MS-AFLP markers which were shared by all plants ($N = 400$). Analysis
288 of molecular variance on the matrix of pairwise dissimilarities between modules revealed
289 that 29.2% of total epigenetic variance in the sample was accounted for by differences
290 among modules of the same plant.

291 Subindividual epigenetic heterogeneity was also apparent when informative MS-AFLP
292 markers were considered individually. The three plants studied were closely similar with
293 regard to the number of informative markers (range = 427-431 markers; Table 1). In every
294 plant a small, but non-negligible fraction of these markers (range = 5.8-7.7%; Table 1) were
295 polymorphic in methylation state among modules of the same plant, i.e., occurred in the
296 methylated and unmethylated states in different parts of the same shrub. About one third of
297 polymorphic informative markers occurred predominantly in the methylated state, and two
298 thirds occurred predominantly in the unmethylated state (Table 1). About 20% of
299 subindividually polymorphic loci ($N = 75$, all plants combined) were polymorphic in more
300 than one plant (Supporting Information Fig. S2).

301 Within-plant genealogy of epigenetic heterogeneity

302 Within-plant heterogeneity in global DNA cytosine methylation had genealogical signatures
303 in two plants (TSE04 and TSE05), as denoted by statistically significant or marginally
304 significant Moran's I autocorrelations, irrespective of the branch length metric used to
305 construct genealogical trees (Table 2). Genealogical character estimations revealed some
306 nodes early in the plants' lives whose descendant branches were consistently characterized
307 until the time of collection by divergent values of global cytosine methylation (marked by
308 arrows in Fig. 2).

309 Heterogeneity among modules of the same plant in multilocus epigenetic fingerprints,
310 as assessed by coordinates from nonmetric multidimensional scaling of pairwise distance
311 matrices (MDS1 and MDS2), had statistically significant genealogical signals in plants
312 TSE03 (axis MDS1) and TSE04 (axes MDS1 and MDS2), irrespective of branch length
313 metric used to construct genealogical trees (Table 3). Genealogical character estimations for
314 MDS1 and MDS2 revealed one or more branching nodes early in the lives of the plants
315 studied whose descendant branches were subsequently characterized by divergent multilocus
316 epigenetic fingerprints until the time of collection (nodes marked by arrows, Fig. 3).

317 Variation across modules of the same plant in the methylation state of the most
318 polymorphic MS-AFLP markers was significantly related to module genealogy (Fritz-
319 Purvis' $D < 1$) in about two thirds of instances tested (15 markers x 2 branch length metric
320 combinations) (Fig. 4; Supporting Information Table S2). One, five and three markers
321 exhibited genealogical signal in plants TSR03, TSE04 and TSE05, respectively (Fig. 4). The
322 D statistic did not depart significantly from Brownian motion expectations ($D = 0$) in any of
323 these cases (Fig. 4). Genealogical character estimations of within-plant variation in
324 methylation state for markers with genealogical signal are shown in Fig. 5 on trees whose
325 branch lengths are age differences (results were closely similar for trees based on linear
326 distances between nodes; Supporting Information Fig. S3). Genealogical clumping of marker
327 methylation state at tree tips (modules) was evident in all cases, although the size of clumps
328 and their time of divergence along the plant's lifetime ranged widely among markers. In
329 some cases the initial divergence in methylation state occurred when plants were ≤ 5 yr old,
330 and its subsequent propagation over many years of branching without methylation change
331 eventually formed large genealogical clumps (e.g., TG_CTA_297, TC_CCT_367). In other
332 cases, in contrast, the initial methylation divergence at the base of a clump occurred when
333 plants were already ≥ 15 yr old, and the resulting genealogical clumps were smaller and
334 involved fewer modules (TC_CGC_347, TC_CCT_200; Fig. 5). Some instances of recent
335 reversals in methylation state were apparent within the genealogically oldest clumps (e.g.,
336 TG_CTA_297, TC_CCT_367; Fig. 5).

337 Paired 'Equal rates' (ER) and 'All rates different' (ARD) discrete evolution models
338 fitted to within-plant methylation state data for individual markers generally provided better
339 support for the ER model (63.3% of instances; Supporting Information Table S2). When
340 ARD models provided a better fit (36.7% of instances), mean (\pm SE) transition rate from the
341 unmethylated to the methylated state (0.360 ± 0.029) was only slightly higher than the
342 transition rate from the methylated to the unmethylated state (0.295 ± 0.022) (Supporting
343 Information Table S2).

344 Discussion

345 Extant subindividual epigenetic heterogeneity and its origin

346 Plants of *L. latifolia* were epigenetically heterogeneous at time of collection. Leaves from
347 different modules in the same plant differed in global DNA methylation and MS-AFLP
348 profiles at multilocus and single-marker levels. These findings extend those of Alonso *et al.*
349 (2018) for this species showing within-plant heterogeneity in global methylation for a

350 superset of the plants considered here. Global methylation may vary within species or
351 individuals in relation to plant age or tissue of origin (Mankessi *et al.*, 2011; Vining *et al.*,
352 2012; Alonso *et al.*, 2017; Gao *et al.*, 2019), but none of these factors can account for
353 heterogeneous genomic methylation within plants of *L. latifolia* found here and by Alonso *et*
354 *al.* (2018), since all DNA samples from the same plant were from even-aged leaf cohorts.
355 The same applies to within-plant heterogeneity in epigenetic fingerprint and methylation
356 state of MS-AFLP markers.

357 A demographic study on the study population revealed that mean longevity of *L.*
358 *latifolia* individuals that flowered at least once during their lifetimes was 22 years, and only
359 ~7% of these lived for >30 years (Herrera & Bazaga, 2016). Since plants included in this
360 study were 23-29 yr old at time of collection, our results refer to the older age class in the
361 population. Insofar as within-plant patterning of epigenetic features is a cumulative process
362 taking place over a plant's lifetime, as suggested by this study, such patterning would
363 possibly have been weaker or harder to detect in younger individuals. This is supported by
364 variation among plants studied in frequency of significant genealogical signals of epigenetic
365 features, which tended to increase from the youngest (TSE03) to the oldest (TSE04)
366 individual. It should also be kept in mind that insufficient statistical power probably
367 hindered detection of genealogical signal in the younger plants. With very small trees ($N \sim$
368 20 tips, as in TSE03 and TSE05), all methods for detecting genealogical signal have high
369 Type II errors (Blomberg *et al.*, 2003; Münkemüller *et al.*, 2012). Larger genealogical trees
370 from species with longer longevities (e.g., trees) should be most favorable for the detection
371 of genealogical signal.

372 Trees used for assessing genealogical signal in extant epigenetic heterogeneity
373 represent the developmental pedigree of modules at the tips, and describe the ontogenetic
374 unfolding of each individual over its lifetime. All modules in a plant derive from the same
375 ancestor, namely the initial seedling arising from a seed, and genealogical trees depict the
376 topology of descendant lineages arising from branching events. Pairs of modules physically
377 closer at tips of a tree are also historically and developmentally closer to their most recent
378 common ancestor module than pairs of modules located farther away in the tree. These
379 relationships, along with the regularly dichasial branching pattern that characterizes *L.*
380 *latifolia* shrubs, justify our application of methods from phylogenetic research to assess
381 genealogical signal and perform genealogical reconstructions of within-plant epigenetic
382 changes (see also Orr *et al.*, 2020). These methods could be used for the same purpose on

383 other woody perennials that follow Leeuwenberg's model of architecture (see, e.g.,
384 Hamilton, 1985; Hallé, 1986; Navarro *et al.*, 2009; for tropical and non-tropical examples).

385 Results of this study agree with expectations from the hypothesis advanced by Alonso
386 *et al.* (2018) that subindividual variation in epigenetic features of *L. latifolia* plants was the
387 consequence of the concerted action of plant sectoriality (plant body's compartmentalization
388 into physiologically semi-autonomous subunits; Watson, 1986) and the differential action on
389 plant parts of some factor(s) inducing persistent changes in extent and/or patterns of DNA
390 cytosine methylation (e.g., pathogens, herbivores, insolation, UV light, water shortage,
391 nitrogen deficiency; reviewed by Alonso *et al.*, 2016). Sectoriality will constrain the
392 horizontal circulation of phloem-mobile molecules that regulate DNA methylation (McGarry
393 & Kragler, 2013; Lewsey *et al.*, 2016), thus contributing to maintain within-plant
394 heterogeneity in epigenetic features arising from random epimutations or localized responses
395 to environmental agents, as previously suggested in relation to other subindividually variable
396 traits (Orians & Jones, 2001; Herrera, 2009). We found here that extant within-plant
397 heterogeneity in epigenetic features (global DNA cytosine methylation, MS-AFLP
398 multivariate fingerprint, methylation state of specific MS-AFLP markers) exhibited
399 statistically significant genealogical signals. Results were considerably robust irrespective of
400 whether branch lengths of genealogical trees were linear or temporal distances between
401 nodes. This points to an equivalence of ageing and growing as ultimate agents of within-
402 plant epigenetic diversification over a plant's lifetime.

403 Genealogical character reconstructions revealed that early events of internal epigenetic
404 divergence took place when plants were still very young (< 5 yr), before reaching the age of
405 first reproduction (Herrera & Bazaga, 2016). In general, the timing of epigenetic
406 modifications spanned the entire lifespan of individuals, thus revealing that epigenetic
407 features experienced steady changes throughout individual plants' lives. This was
408 particularly evident in the case of changes in methylation state of subindividually
409 polymorphic MS-AFLP markers, where changes conforming to a Brownian motion model
410 took place over the life of individuals with about similar estimated probabilities in both
411 directions, and even the reversion to the 'ancestral' methylation state could be documented.
412 A corollary of this finding is that *L. latifolia* individuals can produce slightly different
413 epigenetic fingerprints over its lifetime if sampled repeatedly over a sufficiently broad
414 timespan. This expectation is upheld by preliminary results for plants from our study
415 population which were sampled on two occasions nine years apart (C. M. Herrera,

416 *unpublished data*).

417 Toward an epigenetic mosaicism hypothesis

418 Modular construction by continual organogenesis and reiterated production of homologous
419 structures is a quintessential plant feature which has motivated the consideration of plant
420 individuals as non-unitary metapopulations of semi-autonomous modules, a notion departing
421 from the common zoocentric definition of organismic individuality (Gerber, 2018). This
422 view led to the incorporation of selection at the subindividual level as a possible
423 evolutionary mechanism (Buss, 1983a, b; Pineda-Krch & Poore, 2004), and provided the
424 foundations for the ‘genetic mosaicism hypothesis’ (GMH) (Whitham & Slobodchikoff,
425 1981; Whitham *et al.*, 1984; Gill *et al.*, 1995). The following premises synthesize the GMH
426 (slightly modified from Gill *et al.*, 1995): (i) spontaneous mutations occur among the
427 proliferating meristems; (ii) the meristematic and modular basis of plant development
428 assures that many of these mutations are preserved and expanded hierarchically among
429 modules as the plant grows; (iii) the differential growth and survival of ramets, branches and
430 shoots should alter the genotypic configuration of the plant as it grows; and (iv) the within-
431 plant phenotypic heterogeneity arising from genotypic heterogeneity will affect individual
432 fitness through effects on the progeny, plant responses to the environment, or responses of
433 animal consumers (Whitham & Slobodchikoff, 1981; Whitham *et al.*, 1984; Gill *et al.*, 1995;
434 Herrera, 2009).

435 Studies focusing on subindividual genetic heterogeneity in wild plants have produced
436 few good examples of genetic mosaicism in non-clonal woody plants, and generally
437 documented very low somatic mutation rates (Cloutier *et al.*, 2003; Padovan *et al.*, 2013;
438 Ranade *et al.*, 2015; Schmid-Siebert *et al.*, 2017; Wang *et al.*, 2019; Orr *et al.*, 2020; see also
439 Herrera, 2009, for review). This tends to deny the evolutionary importance of genetic
440 mosaicism advocated by the GMH (Pannell & Eppley, 2004; Gerber, 2018). In contrast, the
441 few investigations that have so far addressed the possibility of subindividual variation in
442 epigenetic features among homologous organs have found relatively high frequencies of
443 somatic epigenetic variants, discernible within-plant epigenetic mosaicism, and/or
444 relationships between subindividual epigenetic heterogeneity and within-plant phenotypic
445 variation (Bitonti *et al.*, 1996; Herrera & Bazaga, 2013; Alonso *et al.*, 2018; Herrera *et al.*,
446 2019; Hofmeister *et al.*, 2020). The present study has extended these previous findings by
447 showing that steady epigenetic diversification over plants’ lifetimes can lie behind extant
448 subindividual epigenetic mosaics.

449 Taken together, results obtained so far bearing on subindividual epigenetic variation
450 motivate our proposal of an ‘epigenetic mosaicism hypothesis’ (EMH) consisting of exactly
451 the same elements i-iv above as the original GMH but where the terms ‘mutation’, ‘genetic’
452 and ‘genotype’ are replaced with ‘epimutation’, ‘epigenetic’ and ‘epigenotype’, respectively.
453 Two additional components of GMH, namely inheritance of somatic mutations and impact
454 of mosaicism on individual fitness, will often apply to EMH as well. Transgenerational
455 epigenetic inheritance has been documented for model and non-model plants (Jablonka &
456 Raz, 2009; Hauser *et al.*, 2011; Quadrana & Colot, 2016). In *L. latifolia* there is extensive
457 transgenerational transmission of genome-wide global cytosine methylation levels and
458 methylation state of anonymous epigenetic markers (Herrera *et al.*, 2018). Although the
459 ecological impact has been rarely investigated, there is also evidence that within-plant
460 epigenetic mosaicism can influence plant fitness (Alonso *et al.*, 2018; Herrera *et al.*, 2019).

461 By incorporating the within-plant realm to the already well-accepted consensus that
462 epigenetic variation is an important source of phenotypic variance among individuals and
463 populations (Bossdorf *et al.*, 2008, 2010; Lira-Medeiros *et al.*, 2010; Medrano *et al.*, 2014;
464 Kooke *et al.*, 2015; Groot *et al.*, 2018), the EMH offers a particularly favorable arena for
465 formulating and testing novel hypotheses on the ecological and evolutionary roles of
466 epigenetic variation while holding constant the influence of genetic factors. For example,
467 lifetime internal epigenetic diversification within individuals, as documented here for *L.*
468 *latifolia*, may represent a mechanism of ‘exploration’ of the epigenetic landscape endowing
469 each plant with a broader phenotypic space to cope with challenges of the abiotic and biotic
470 environment. The breadth of such epigenetic sampling (i.e., within-individual
471 epigenetic/phenotypic variance) should vary depending on life expectancy and species-
472 specific patterns of meristem divisions related to the architectural model. Simple predictions
473 from hypotheses framed around the EMH are amenable to experimentation by manipulating
474 within-plant epigenetic heterogeneity while keeping genetic background constant, e.g., by
475 localized application of chemical agents which alters methylation and monitoring effects on
476 phenotypic heterogeneity and ecological consequences (Herrera *et al.*, 2019). These
477 investigations are bound to contribute new insights on the mechanistic basis and ecological
478 and evolutionary implications of the within-plant component of phenotypic variance in plant
479 populations.

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488 **Author contributions**

489 CMH designed the research, did field work, conducted data analyses, and led the writing;
490 CA contributed data analyses and interpretations, and supervised laboratory work; RP and
491 PB performed HPLC and MS-AFLP analyses, respectively; all authors reviewed and edited
492 the manuscript.

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737 **Supporting Information**

738 Additional Supporting Information may be found online in the Supporting Information
739 section at the end of the article.

740 **Fig. S1** Tree representations of genealogical relationships among the modules sampled in
741 each of the three *Lavandula latifolia* individuals studied.

742 **Fig S2** Venn diagrams showing the distribution among plants of subindividually
743 polymorphic methylation-sensitive AFLP markers.

744 **Fig. S3** Within-plant genealogical character estimation of the methylation state of
745 polymorphic MS-AFLP markers on trees whose branch lengths are linear distances between
746 nodes.

747 **Table S1** Primer combinations and number of fragments that were used in the MS-AFLP
748 analyses of leaf DNA samples.

749 **Table S2** Summary of tests of within-plant genealogical signal, and fits of discrete character
750 models, for highly polymorphic individual MS-AFLP markers.

751 **Table 1** Variation among modules of the same plant in methylation state of MS-AFLP
752 markers. See supporting Information Fig. S2 for distribution among plants of shared and
753 unique polymorphic markers.

Plant	Modules sampled	Informative markers*	Subindividually polymorphic markers		
			Total	Predominant state:	
				Methylated	Unmethylated
TSE03	20	431	33	11	22
TSE04	38	427	33	9	24
TSE05	22	427	25	8	17

754 * Out of a total of the $N = 467$ markers considered whose estimated scoring error rates were
755 equal to zero. Individual markers were considered informative for a given plant if they were
756 recorded in at least one of its modules (see Materials and Methods: Data analysis).

757 **Table 2** Tests of genealogical signal for among-module variation in global DNA cytosine
758 methylation within the three *Lavandula latifolia* plants studied.

Plant	Branch length metric of genealogical tree			
	Linear distance		Age difference	
	Moran's <i>I</i>	<i>P</i> -value *	Moran's <i>I</i>	<i>P</i> -value *
TSE03	-0.0498	0.45	-0.0518	0.46
TSE04	0.0210	0.046	0.0248	0.048
TSE05	0.0199	0.067	0.0836	0.015

759 * Determined by permutation tests with 10^5 repetitions.

760 **Table 3** Tests of genealogical signal for among-module variation in multilocus epigenetic
761 fingerprints, assessed by coordinates from nonmetric multidimensional scaling (Fig. 1).

Plant	Coordinate	Branch length metric of genealogical tree			
		Linear distance		Age difference	
		Moran's <i>I</i>	<i>P</i> -value *	Moran's <i>I</i>	<i>P</i> -value *
TSE03	MDS1	-0.0529	0.46	-0.0610	0.53
	MDS2	0.0437	0.0074	0.0440	0.010
TSE04	MDS1	0.0513	0.0055	0.0673	0.0039
	MDS2	0.1452	<10 ⁻⁵	0.2070	<10 ⁻⁵
TSE05	MDS1	-0.0576	0.56	-0.0449	0.43
	MDS2	-0.0894	0.86	-0.0871	0.79

762 * Determined by permutation tests with 10⁵ repetitions.

763 **Legends to figures**

764 **Fig. 1** Scatterplot of the $N = 80$ modules (dots) sampled from three *Lavandula latifolia*
765 plants (TSE03, TSE04, TSE05) on the plane defined by nonmetric multidimensional scaling
766 of the matrix of pairwise epigenetic distances (MDS1 and MDS2; coordinates scaled to
767 standard deviation unit and centered to the mean). Epigenetic distances between modules
768 were obtained from the binary matrix of methylation states for the $N = 400$ informative MS-
769 AFLP loci shared by all plants. Ellipses denote the 95% bivariate confidence intervals
770 around individual plant means. A small amount of random variation was added to the
771 location of each point to reveal modules with identical coordinates.

772 **Fig. 2** Genealogical character estimation of within-plant changes in global DNA cytosine
773 methylation for the three *Lavandula latifolia* plants studied (TSE03, TSE04, TSE05). Two
774 genealogical trees were used for each plant, whose branch lengths were either linear
775 distances (left column) or age differences (right column) between nodes (Supporting
776 Information Fig. S1). Estimated changes in trait value along branches are color-mapped on
777 each tree according to the scales shown. Limits of color scales differ slightly among plants
778 because they were adjusted in each case to the corresponding minimum and maximum
779 values. The arrows mark branching nodes referred to in the text.

780 **Fig. 3** Genealogical character estimation within individual *Lavandula latifolia* plants of
781 changes in multilocus epigenetic fingerprints of individual modules, as described by their
782 coordinates on the axes obtained from nonmetric multidimensional scaling of pairwise
783 distance matrices (MDS1 and MDS2). Two trees were used for each plant, whose branch
784 lengths were either linear distances (two left columns) or age differences (two right columns)
785 between nodes. Estimated changes in trait value along branches are color-mapped on each
786 tree. In each tree, the color scale was defined by the minimum (white) and maximum (red)
787 values for the corresponding plant and axis (see Fig. 1). Scales have been omitted to reduce
788 cluttering. The arrows mark branching nodes referred to in the text.

789 **Fig. 4** Summary of analyses of within-plant genealogical signal (Fritz-Purvis' D statistic) in
790 the methylation state of highly polymorphic MS-AFLP markers (frequency of commonest
791 methylation state < 0.85 ; $N = 3, 6$ and 6 markers for plants TSE03, TSE04 and TSE05,
792 respectively). Blue and red dashed lines mark expected values from random and Brownian
793 motion distributions of methylation state across tips of genealogical trees. Filled dots denote
794 markers whose methylation state simultaneously exhibited significant genealogical clumping

795 within plants (D significantly < 1) and nonsignificant departure from Brownian motion
796 expectation ($D = 0$). See Supporting Information Table S2 for detailed numerical results.
797 **Fig. 5** Within-plant genealogical character estimation of the methylation state of nine highly
798 polymorphic MS-AFLP markers with significant genealogical signal (Fig. 4, and Supporting
799 Information Table S2). For each marker, methylation state in the sampled modules (tree tips)
800 and estimated posterior probabilities at nodes are coded as grey (methylated) or orange
801 (unmethylated). Tree branch lengths represent differences in age between nodes and units of
802 horizontal axes are years. Markers are identified by primer combination and fragment size in
803 base pairs (Supporting Information Table S1), and correspond to plants TSE03
804 (TG_CTA_297), TSE04 (TT_CAC_413, TA_CTA_278, TC_CGC_347, TA_CAC_251,
805 TC_CCT_367) and TSE05 (TC_CGC_241, TC_CCT_200, TC_CCT_390).

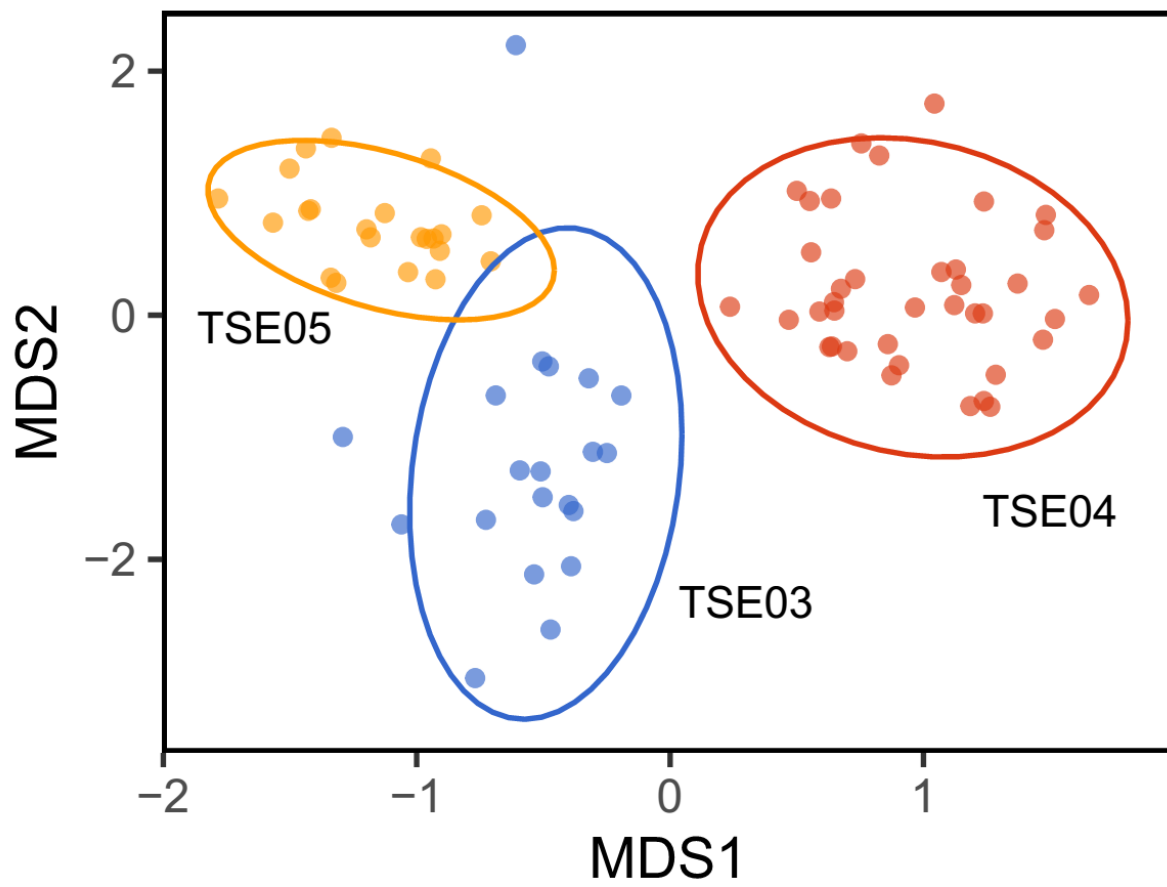


Fig. 1

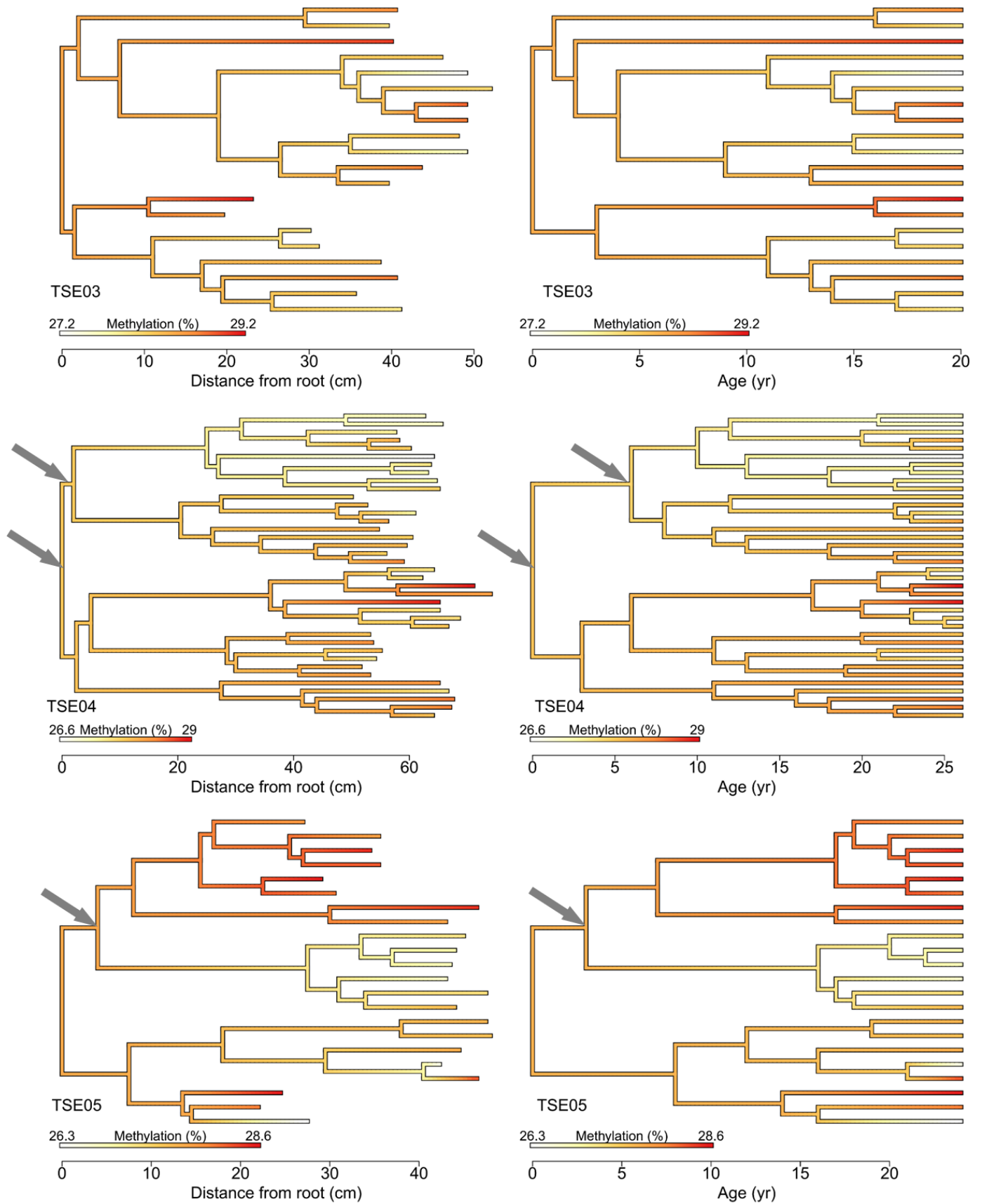


Fig. 2

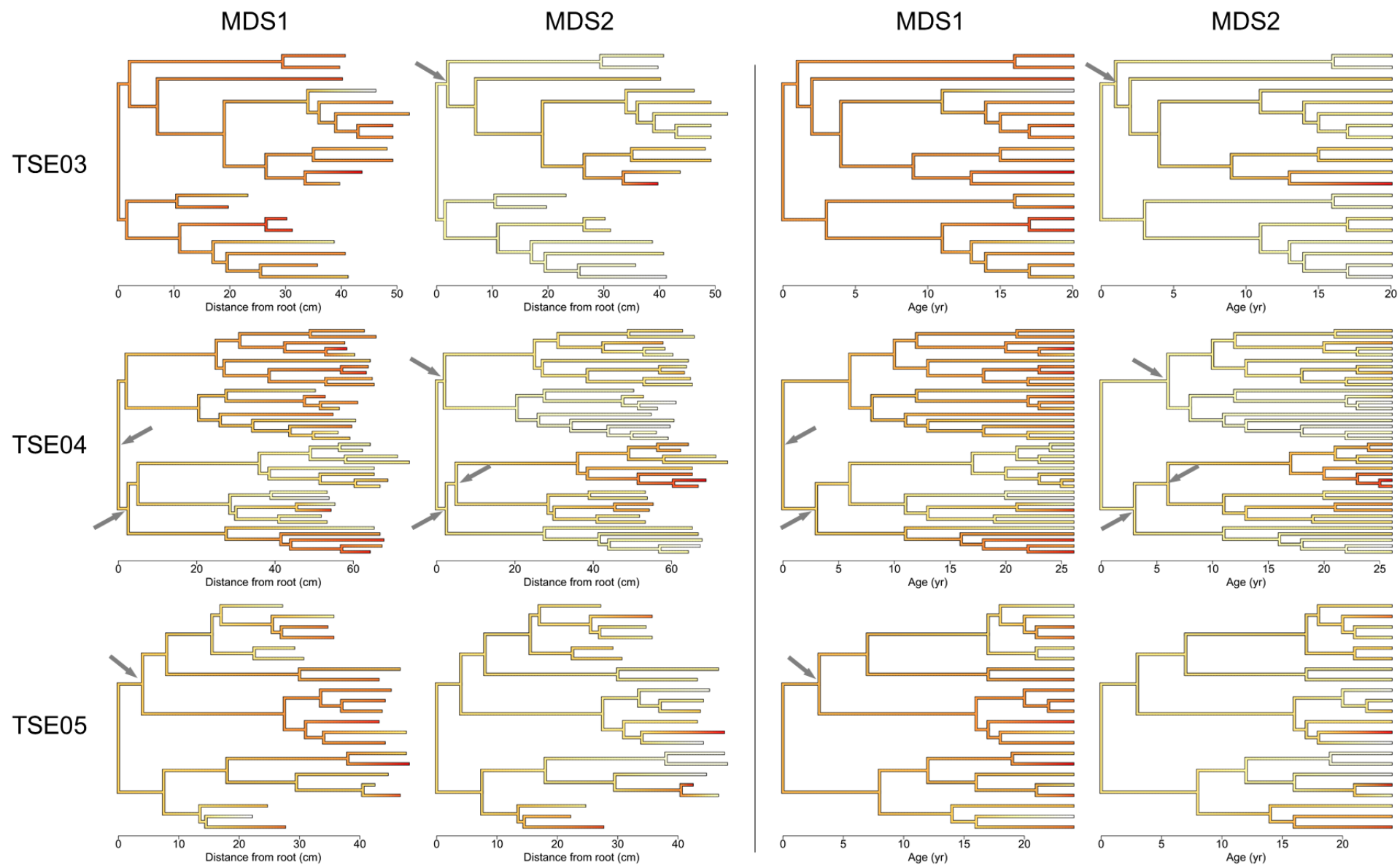


Fig. 3

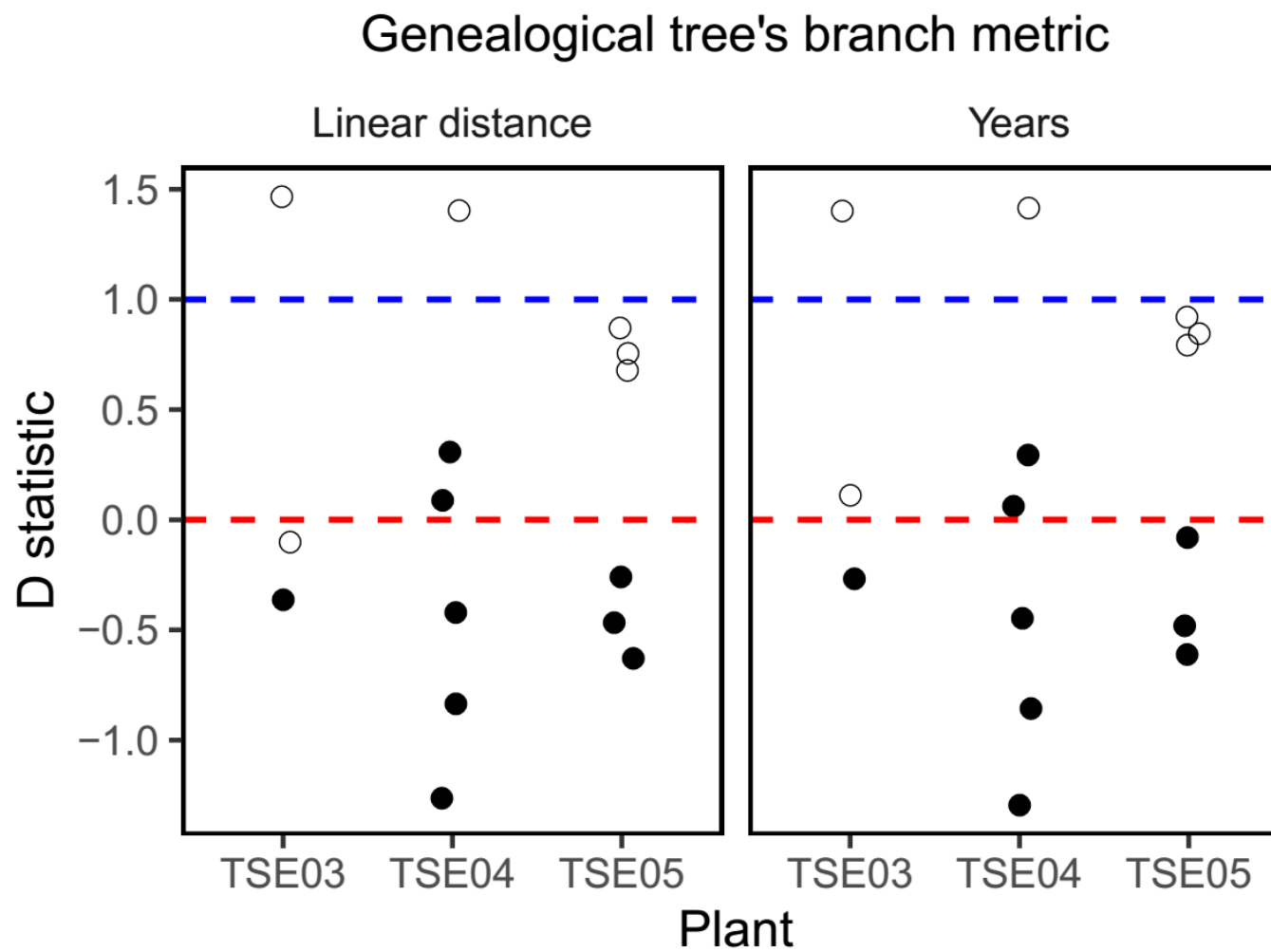


Fig. 4

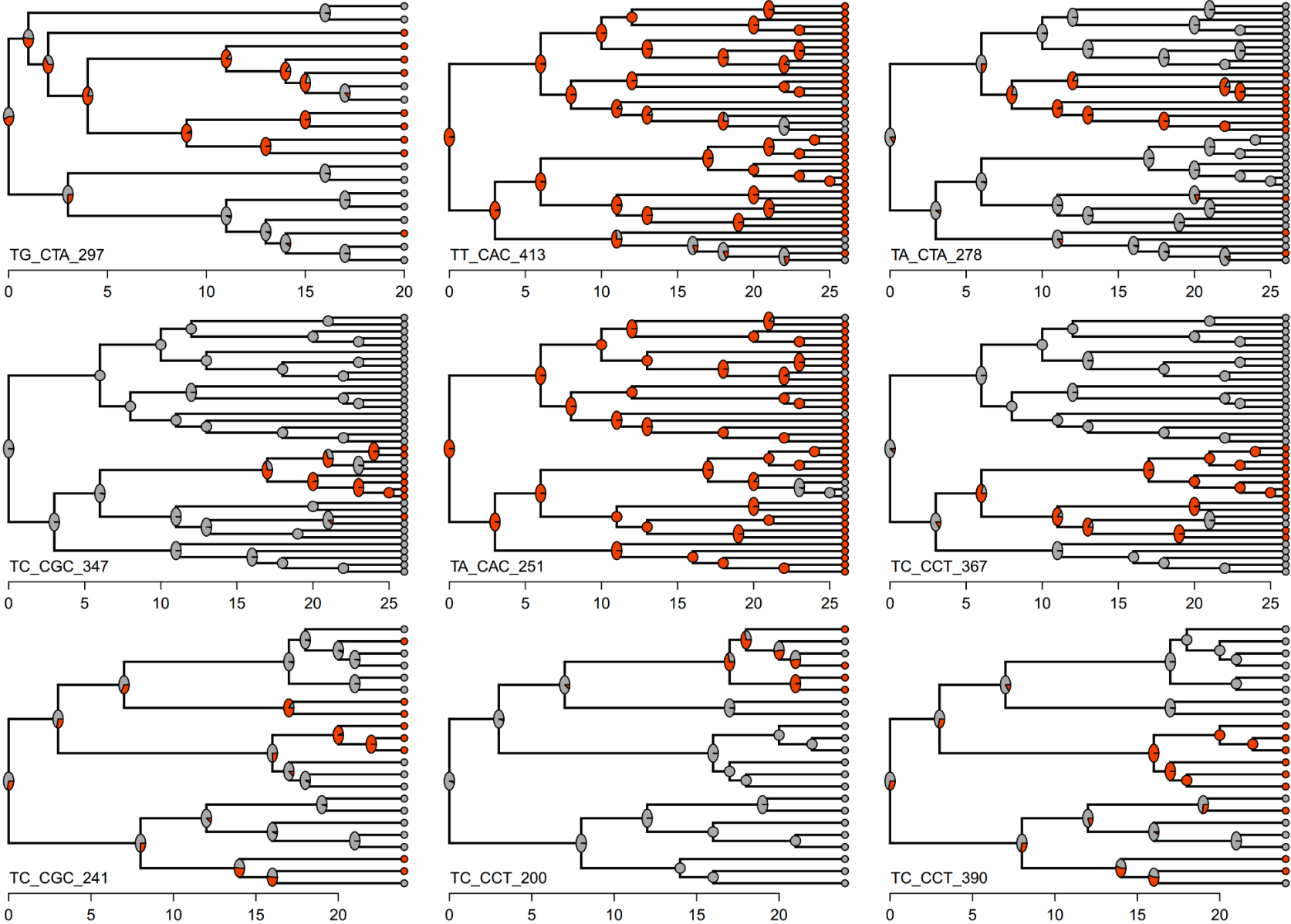


Fig. 5