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# Genome sequencing and population genomic analyses provide insights into the adaptive landscape of silver birch

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Silver birch (*Betula pendula*) is a pioneer boreal tree that can be induced to flower within 1 year. Its rapid life cycle, small (440-Mb) genome, and advanced germplasm resources make birch an attractive model for forest biotechnology. We assembled and chromosomally anchored the nuclear genome of an inbred *B. pendula* individual. Gene duplicates from the paleohexaploid event were enriched for transcriptional regulation, whereas tandem duplicates were overrepresented by environmental responses. Population resequencing of 80 individuals showed effective population size crashes at major points of climatic upheaval. Selective sweeps were enriched among polyploid duplicates encoding key developmental and physiological triggering functions, suggesting that local adaptation has tuned the timing of and cross-talk between fundamental plant processes. Variation around the tightly-linked light response genes *PHYC* and *FRS10* correlated with latitude and longitude and temperature, and with precipitation for *PHYC*. Similar associations characterized the growth-promoting cytokinin response regulator *ARR1*, and the wood development genes *KAK* and *MED5A*.

Forest ecosystems maintain a large share of Northern Hemisphere biodiversity. Boreal forests comprise roughly 30% of global forest area<sup>1</sup> and contain the highest tree density among climate zones<sup>2</sup>. Moreover, boreal regions are undergoing extensive climate change. Annual temperature increases exceeding 1.5 °C are projected to result in a warming of 4–11 °C by the end of this century, with little concomitant increase in precipitation<sup>1</sup>. At this pace, climate zones will shift northward at a greater speed than trees can migrate<sup>3</sup>. To understand how future populations of forest trees may respond to climate change, it is essential to uncover past and present signatures of molecular adaptation in their genomes. Silver birch, *B. pendula*, is a pioneer species in boreal forests of Eurasia. Flowering of the species

can be artificially accelerated<sup>4</sup>, giving it an advantage over other tree species with published genome sequences (such as poplar<sup>5</sup>, spruce<sup>6</sup>, and pine<sup>7</sup>) for the optimization of fiber and biomass production.

Here we sequenced 150 birch individuals and assembled a *B. pendula* reference genome from a fourth-generation inbred line, resulting in a high-quality assembly of 435 Mb that was linked to chromosomes using a dense genetic map. We analyzed SNPs in the genomes of 80 birch individuals spanning most of the geographic range of *B. pendula*, as well as seven other members of Betulaceae. Population genomic analyses of these data provide insights into the deep-time evolution of the birch family and on recent natural selection acting on silver birch.

A full list of affiliations appears at the end of the paper.

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## RESULTS

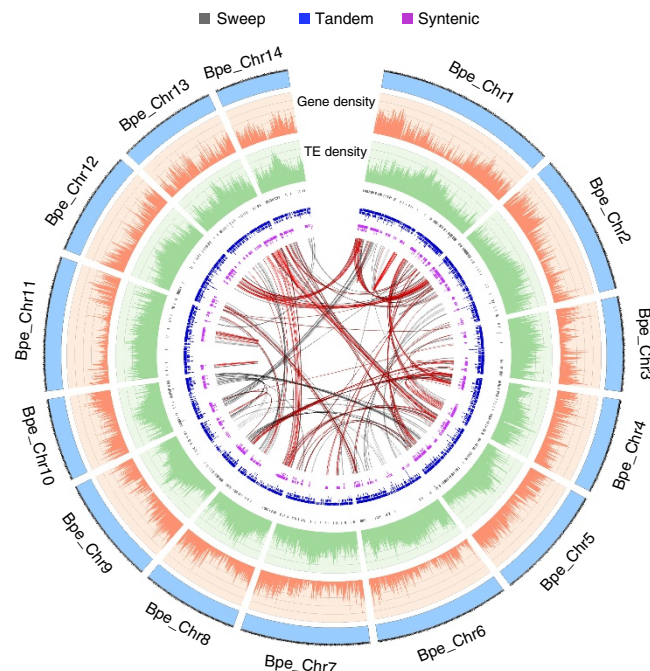
## Genome assembly and gene duplication history

We constructed a hybrid nuclear genome assembly starting from 9× sequence coverage with Roche 454 technology. Assembled contigs were polished with Illumina paired-end data and connected and ordered using mate-pair libraries sequenced on both Illumina and SOLiD platforms, followed by further scaffolding and gap filling with 30× coverage of PacBio reads longer than 6,000 bp. This resulted in a first assembly consisting of 5,642 scaffolds with an N50 of 240 kb (Supplementary Figs. 1 and 2 and Supplementary Tables 1 and 2). Further scaffolding with additional PacBio reads resulted in 3,474 (super)scaffolds with N50 value of 527.7 kb. A total of 391 Mb of scaffolds (89% of the estimated 440-Mb genome) was assembled into 14 chromosomal linkage groups via an ultra-high-density genomic linkage map consisting of 3.4 million markers (Fig. 1 and Supplementary Note). In addition to the nuclear genome, organellar genomes were assembled and annotated (Supplementary Note and Supplementary Figs. 3–6). Evidence for birch gene models was obtained by sequencing EST libraries from 12 different birch tissues or growth conditions, providing 18,951 transcripts with an average length of 1,683 bp, and by carrying out *de novo* assembly of RNA-seq reads, yielding 16,906 transcripts (Supplementary Figs. 7 and 8, Supplementary Tables 3 and 4, and Supplementary Note). We annotated the nuclear genome in two stages. After initial automated gene prediction, 3,192 genes were manually annotated and used to train gene predictors for a second round, yielding 28,153 high-quality gene models (Supplementary Figs. 9–11, Supplementary Tables 5–7, and Supplementary Note), of which 17,746 were supported by nearly full-length transcriptomic evidence.

Transposable elements (TEs) constituted 49.23%, and retrotransposons 30.60%, of the genome (Supplementary Note and Supplementary Table 8). Superfamily Gypsy and Copia retrotransposons were less common (8.5% and 2.3%, respectively), and contained fewer young (<50,000 years) elements than other plant genomes of similar size, whereas the nonautonomous TRIM group of retrotransposons was significantly more abundant<sup>8</sup>, at 6.4% versus 1.26% (at most, in *Pyrus*, pear;  $P < 2.2 \times 10^{-16}$ , Grubbs test for one outlier). This suggests that the parasitic life cycle of TRIMs may attenuate replication of autonomous retrotransposons in *B. pendula*, thus limiting their contribution to genome size.

Syntenic alignment of the *B. pendula* genome with other eudicots, including grapevine (Supplementary Figs. 12 and 13 and Supplementary Table 9), demonstrated that the birch genome has not undergone whole-genome duplications (WGDs) subsequent to divergence from these species (Supplementary Note). As such, the only internally duplicated blocks in the *B. pendula* genome date from the ancient gamma hexaploidy event at the base of core eudicots<sup>9,10</sup>.

Gene duplication and divergence is a major source of functional novelty in eukaryotic genomes, and in plants both polyploidy and tandem duplications have been implicated in the evolution of phenotypic novelty<sup>11,12</sup>. Using self-self syntenic analysis, we separated the duplicated portion of the *B. pendula* genome into two pools: duplicate genes deriving from the ancient hexaploidy event, and those stemming from ongoing tandem (segmental) duplications (Supplementary Note). Gene ontology (GO) functional enrichment analysis (Supplementary Tables 10–12) revealed that transcription factors (TFs) were strongly overrepresented among polyploid duplicates (Supplementary Table 12), which corresponds with theoretical and empirical results indicating biased retention of highly interconnected genes following the duplication of entire functional modules<sup>13,14</sup>. This result appeared to hold for adaptive genome evolution in eudicots in general and

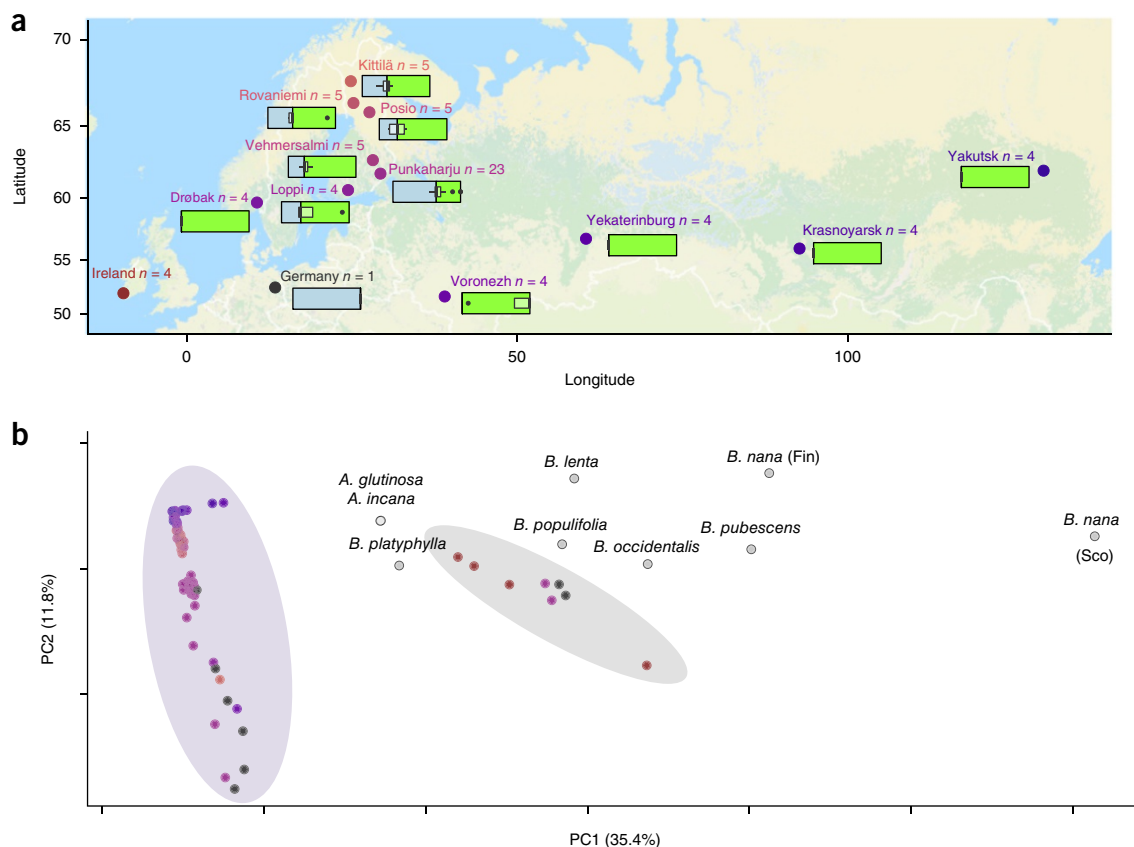


**Figure 1** The pseudomolecule-level assembly of the silver birch genome shows an absence of WGD events since the ancient gamma hexaploidy. Center, syntenic links between gene duplicates and triplicates dating from the hexaploid event. Ancient triplicate links still preserved in the modern genome are shown in red. Bpe\_Chrl, *B. pendula* chromosome.

after independent WGDs, as we obtained highly similar functional enrichments in corresponding analyses of the *Arabidopsis thaliana* and poplar genomes<sup>15</sup> (Supplementary Table 12), which have experienced their own lineage-specific WGDs<sup>5,16</sup> since diverging from a common ancestor with birch. In contrast to biased retention of modular TF function after WGDs, tandemly duplicated genes in these three species<sup>15</sup> were enriched for environmental responses and secondary metabolism, which, although distinctive by species, were also significantly overlapping ( $P < 2.2 \times 10^{-16}$ , Fisher's test; Supplementary Note, Supplementary Fig. 14, and Supplementary Tables 11 and 12). Tandemly expanded gene families shared by all three species were enriched for secondary metabolism, bacterial defense, hormonal response, and hormone and nutrient transport. Adaptations possibly related to the arborescent habit were visible in convergent tandem expansions shared by *B. pendula* and *Populus trichocarpa*, including genes associated with fungal pathogen defense, cell wall biogenesis, and cellulose synthase activity (Supplementary Tables 11 and 12). Through evolutionary information stored within single genomes, these results suggest that whereas polyploid duplicates tend to diversify core processes in developmental and physiological regulation, tandem duplicates enhance the diversity of a plant's environmental response capacity, which is in concurrence with previous studies<sup>15,17,18</sup>.

## Population-level signatures of interspecies admixture

To examine recent *B. pendula* adaptation and to place the species into perspective within its parent clade, we sequenced the genomes of five other diploid birch species (*Betula nana*, *Betula platyphylla*, *Betula populifolia*, *Betula occidentalis*, and *Betula lenta*), the tetraploid birch *Betula pubescens*, two alder species from the related genus *Alnus* (*Alnus incana* and *Alnus glutinosa*), and *B. pendula* individuals originating from 12 populations native to Ireland, Norway, Finland, and



**Figure 2** Population genomics of silver birch and Betulaceae relatives. (a) Dispersion of 80 silver birches sampled from 12 sites across most of the geographic range of *B. pendula*. Populations are plotted with dots color-coded based on dispersion by latitude and longitude. ADMIXTURE analysis of SNP variation (superimposed with bar plots; center line, median; box, interquartile range; whiskers, 1.5× interquartile range; points, outliers) shows that Finland is a mixing zone between European (blue) and Asian (green) source populations. Samples from Ireland were highly admixed with other birch species and/or polyploid and were removed from the analysis. Source: OpenStreetMap contributors. (b) PCA shows clear separation between *B. pendula* populations and other sampled birch species (open circles). Eight *B. pendula* individuals (gray shading) were putative polyploids or interspecies admixed individuals. These included all Irish individuals and two individuals from Punkaharju, Finland. The main *B. pendula* population formed a cline along PC2 (purple shading). Fin: Finland, Sco: Scotland.

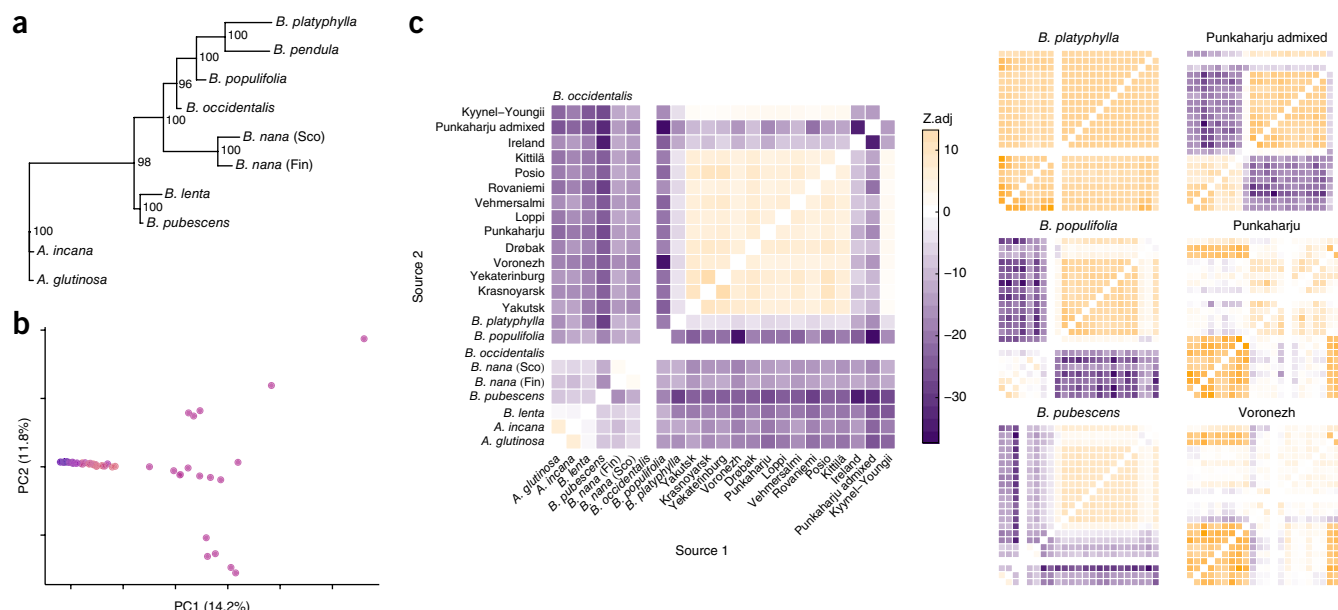
Russia (Fig. 2a and Supplementary Tables 13 and 14). Additionally, eight ornamental varieties of *B. pendula* were included to scan for candidate gene mutations. SNPs were called using GATK<sup>19</sup> for this collection of 89 individuals (Supplementary Table 15). Principal component analysis (PCA) of fourfold degenerate neutrally evolving SNPs demonstrated that whereas most *B. pendula* individuals formed a single linear cline along PCA axis 2, another set consisting of eight *B. pendula* individuals was separated by PCA axis 1, following a trajectory suggestive of admixture from other birch species (Fig. 2b and Supplementary Fig. 15). Flow cytometric analysis of the latter set showed ploidy levels of four for two individuals where cambium material was available, and all atypical individuals showed high levels of heterozygosity (Supplementary Figs. 16 and 17), suggesting the possibility of novel polyploidies or admixture events with parents of polyploid origin.

Hybridization among birch species is well studied<sup>20,21</sup>. Explicit allele-sharing analyses with three-population  $F_3$  tests<sup>22</sup> demonstrated traces of interspecific admixture within the main *B. pendula* population, suggesting that gene flow is ongoing and possibly bidirectional (Fig. 3 and Supplementary Table 16). Some *B. pendula* individuals appeared to be highly admixed, including a few individuals from Punkaharju, Finland, where the main population was not admixed (Fig. 3c). The diploid species *B. occidentalis* and *B. populifolia*

displayed strong signatures of introgression from other species, including *B. pendula* (Fig. 3c and Supplementary Table 16). A phylogeny of diploid birches and alders estimated from SNP data (Fig. 3a) placed *B. lenta* at the first split within the *Betula* genus, as suggested previously<sup>23</sup>, and supported by limited ribosomal DNA (rDNA) internal transcribed spacer (ITS) data<sup>24</sup>. Notably, when SNP called as a diploid, the tetraploid *B. pubescens* was placed in the same clade as *B. lenta*. Further analysis with three-population tests showed high levels of allele sharing between these two species (Supplementary Table 16), suggesting that a species closely related to *B. lenta* may be one of the diploid ancestors of *B. pubescens*. In an ITS-based phylogeny<sup>24</sup>, *B. pubescens* was placed together with *B. pendula*, which suggests that it could be the second ancestor of the apparently allotetraploid *B. pubescens*.

For further analysis within *B. pendula*, we excluded the outlying individuals to reduce the confounding effects of interspecies admixture and polyploidy; this necessitated the removal of all Irish samples. Analysis using ADMIXTURE showed very weak population structure with a split into two ancestral populations, roughly divided into eastern and western groups (Fig. 2a and Supplementary Fig. 15), with some gene flow occurring between them, in Finland (Fig. 3c and Supplementary Table 16). This probably reflects allopatric division during the last Ice Age, followed by subsequent admixture when the





**Figure 3** Phylogenetic splits and admixtures among Betulaceae. **(a)** A phylogeny for *Betula* species, estimated from SNPs in neutrally evolving sites and noncoding sites 2 kb upstream and downstream of genes. Fin, Finland; Sco, Scotland. **(b)** PCA plot of 60 *B. pendula* individuals with low levels of admixture and known location information. Individuals diverging along PC2 are all from Punkaharju, Finland, indicating possible admixture from unknown source populations. **(c)** Heat maps for three-population  $F_3$  test statistics. Introgression is significant if the false discovery rate-adjusted Z-score ( $Z_{adj}$ ; see **Supplementary Note**) is significantly negative (adjusted Z-score < -1.96, purple).

two populations rejoined after ice-sheet retreat, as has been suggested on the basis of chloroplast DNA evidence<sup>25</sup>. The small number of ancestral populations is probably due to the high degree of interbreeding within birch populations; as a wind-pollinated species, birch pollen can spread more than 1,000 km<sup>26</sup>.

With the inclusion of ornamental cultivars, we sought to discover mutations in candidate genes that may account for their horticulturally interesting phenotypes. Among these were *B. pendula* ‘Youngii’, a weeping birch with a pendulous growth habit for which an in-frame stop codon was found in the birch *AtLAZY1* ortholog (**Supplementary Note** and **Supplementary Fig. 18**). *LAZY1* is a member of the IGT protein family<sup>27</sup> and regulates tiller orientation in rice and maize as well as inflorescence branch angle in *Arabidopsis*<sup>28–30</sup>. It is thought to influence gravitropism through regulation of auxin transport and signaling<sup>28–30</sup>. Lateral organs in maize *lazy1* mutants fail to grow vertically, giving rise to a phenotype similar to that observed in ‘Youngii’. The stop codon in the birch *LAZY1* ortholog could thus explain its weeping phenotype.

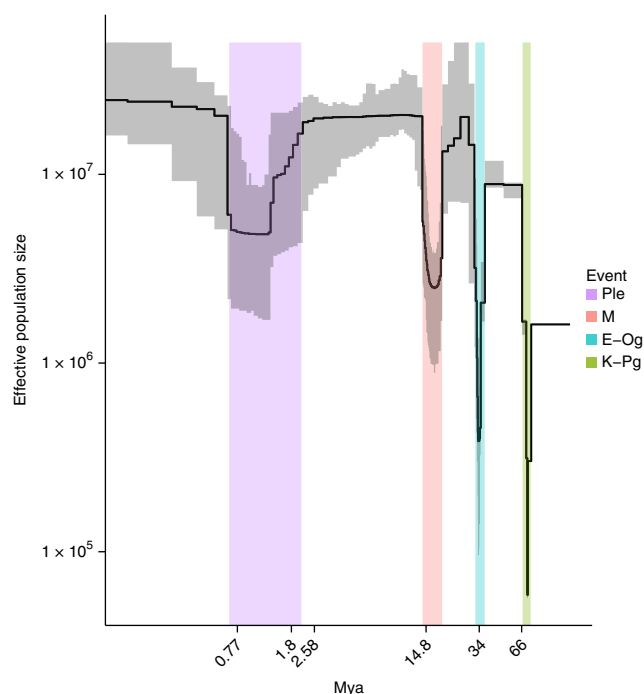
### Population history shows ancient bottlenecks

For analyses of *B. pendula* effective population size ( $N_e$ ) over time, we removed ornamental varieties, which do not have clear origin records, narrowing the analysis to 60 individuals. Within this set, linkage disequilibrium decay was slower (**Supplementary Note** and **Supplementary Fig. 19**), and nucleotide diversity (estimated to be 0.0088, **Supplementary Table 17**) was roughly 30% lower than in *Populus tremula* and *Populus tremuloides*<sup>31</sup>. The ancestral alleles for Betulaceae were reconstructed using the *B. pendula* reference genome and eight diploid *Betula* and *Alnus* species by estimating a phylogenetic tree and resolving ancestral states at nodes (**Supplementary Note**). The reconstruction was used to estimate the site frequency spectrum for the 60 *B. pendula* genomes using ANGSD<sup>32</sup> and to generate a stairway plot<sup>33</sup> elucidating  $N_e$  history over time (**Fig. 4**, **Supplementary Fig. 20**,

and **Supplementary Note**). With a mutation rate estimate of  $1 \times 10^{-9}$  mutations per generation<sup>34</sup> and a generation time of 20 years (**Supplementary Note**), the stairway plot revealed population bottlenecks over deep time that correspond roughly with known events of environmental upheaval (**Fig. 4**). An early  $N_e$  drop coincident with the great extinction event at the Cretaceous–Paleogene (K–Pg) boundary was clearly visible, followed by a rapid population expansion. Later  $N_e$  bottlenecks appeared during Eocene–Oligocene, mid-Miocene, and Pleistocene periods that correspond with other well-known episodes of environmental change<sup>35,36</sup>. The inferred history is largely supported by fossil evidence for Betulaceae<sup>23</sup>. Notably, the  $N_e$  dips we observed could be associated with cladogenetic events during Betulaceae history, as alder–birch speciation occurred soon after the K–Pg event, 60 million years ago (Mya), and the mid-Miocene event ~14 Mya may have included the white-barked birches, for which there is fossil evidence from the late Miocene, 10 Mya<sup>23</sup>. In contrast, we observed no  $N_e$  bottlenecks during Holocene population history, for which a highly negative Tajima’s  $D$  of -1.82 would imply ongoing population expansion. Taken together, the *B. pendula* reference genome, resequenced individuals, and additional resequenced species provide an overview of the population genomic history for the entire Betulaceae clade over the past 65 million years.

### Selective sweeps reveal coordinated local adaptation

We analyzed the same population of 60 resequenced silver birch individuals for selective sweeps (**Supplementary Note**), where natural selection acting on a locus sweeps away variation across a genomic region surrounding the locus. Annotation of genes overlapping the sweep region or 2-kb flanking regions on either side indicated that some positive hits were probably artifacts resulting from recent insertions of chloroplast, mitochondrial, or TE DNA. Owing to their haploid nature at insertion, these horizontal transfers probably simulated homozygosity patterns reflective of selective sweeps. In total, 108



**Figure 4** Historical effective population size for silver birch beginning from 80 million years ago to present. Stairway plot showing that the *B. pendula* population has undergone bottlenecks during four known periods of major climate upheaval: the K–Pg (green) the Eocene–Oligocene (E–Og; blue), the mid-Miocene (M; red), and the Pleistocene (Ple; purple). Data are median estimate from 200 bootstrap replicates (black line) and 95% confidence intervals (shading). Tick marks along the x axis show estimates for the Matuyama–Brunhes (0.77 Mya), Calabrian (1.8 Mya), and Gelasian (2.58 Mya) borders; the mid-Miocene disruption (14.5–14.8 Mya), and the E–Og (34 Mya) and K–Pg (66 Mya) events.

genes near organellar or TE inserts were excluded, resulting in a final collection of 913 genes at or around which selection may have swept variation (**Supplementary Table 18**). This set was enriched for nontandem duplicates and single-copy genes. Tandemly duplicated genes did not show significant enrichment, suggesting that selection among tandemly expanded genes acts through a different process. Regarding the age of the genes under selective sweeps, old syntenic orthologs were enriched for sweeps ( $P = 0.0018$ , Fisher's test) whereas young birch-specific genes were significantly depleted of them ( $P < 2.2 \times 10^{-16}$ ). Additionally, birch-specific nontandem genes were depleted of sweeps ( $P = 3.352 \times 10^{-15}$ ), excluding a possible confounding influence from tandem expansions. These results suggest that recent selective sweeps acted mostly on anciently diverged regulatory components.

Although GO categorization has known pitfalls, it provides one of the best means to objectively characterize gene sets<sup>37</sup>. Exploratory functional enrichment analysis of genes in the sweep regions revealed three significantly enriched GO categories: transmembrane receptor protein tyrosine kinase signaling pathway ( $P = 1.24 \times 10^{-5}$ , Fisher's test, Bonferroni adjusted); peptidyl-histidine phosphorylation  $P = 3.91 \times 10^{-5}$ ; and longitudinal axis specification ( $P = 0.00212$ ). These highlight known functions from model systems related to wood and fiber development, light sensing, embryogenesis, and reproductive isolation. The first GO category includes 23 genes influenced by selective sweeps (**Supplementary Table 19**), most of which are phylogenetically verified homologs of *Arabidopsis* genes encoding functionally characterized CLAVATA1-like receptor-like kinases (RLKs), including

BAM3, PXC2, PXC3, MOL1, MIK1, and MDIS1. In *Arabidopsis*, BAM3 controls leaf shape, size, and symmetry, as well as protophloem development<sup>38</sup>. The PXC genes are known to be involved in secondary cell wall formation in developing wood<sup>39</sup>. *MORE LATERAL GROWTH* (MOL1) is involved in cambium homeostasis, normally repressing secondary growth<sup>40</sup>. *MDIS1-INTERACTING RECEPTOR LIKE KINASE1* (MIK1) is related to the PXC genes and also has a role in stem vascular development. MDIS1 forms a receptor complex with MIK1 and MIK2 that mediates the male perception of female chemoattractant LURE1 during fertilization in *Arabidopsis*<sup>41</sup> and contributes to reproductive isolation between species. Transformation of *AtMDIS1* to *Capsella rubella* partially broke down the interspecific reproductive isolation barrier<sup>41</sup>. Natural selection affecting birch *MDIS1* therefore could suggest possible involvement in determination of reproductive barriers between different birch species.

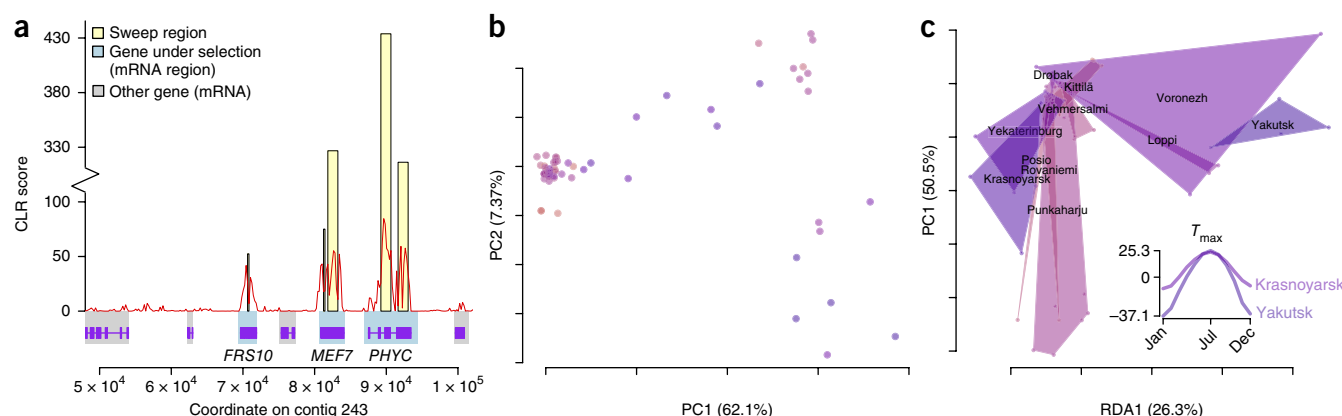
The second GO category highlighted by our exploratory analysis, peptidyl-histidine phosphorylation, includes nine phylogenetically verified orthologs of the phytochrome genes *PHYA*, *PHYB*, and *PHYC*, and genes encoding histidine kinases such as cytokinin receptors AHK2, AHK3, and AHK4 (CRE1), osmosensor AHK1, and ethylene receptors ERS1 and ETR2. The phytochromes are major mediators of red and far-red light responses that have vital roles in plant growth and reproduction<sup>42</sup>. The cytokinin and ethylene receptors control many key aspects of plant physiology and development, including acclimation to abiotic stress, shoot and root vascular development, flowering time, and longevity<sup>43,44</sup>.

The third GO category with putatively important functional enrichment, longitudinal axis specification, includes six phylogenetically verified genes including orthologs of *Arabidopsis* *MONOPTEROS* and *GNL1*, two homologs of *TOADSTOOL 2*, and two homologs of *WRKY2*. *MONOPTEROS* and *GNL1* operate in embryo and vascular development<sup>45,46</sup>, *TOADSTOOL 2* operates in meristem maintenance<sup>47</sup>, and the *Arabidopsis* *WRKY2* protein acts in zygote polarization in embryo development<sup>48</sup>. Additionally, *WRKY2* has a role in pollen development<sup>49</sup> and growth arrest induced by ABA during seed germination<sup>50</sup>.

### Candidate gene adaptation correlates with environment

To assess whether selective sweeps were confounded by population structure, we performed redundancy analysis (RDA)<sup>51</sup> on SNPs in the putative sweep regions by comparing the PCA eigenvectors from their SNP variation to overall population structure from PCA of whole-genome SNPs (**Fig. 5** and **Supplementary Note**). In total, 423 of the 913 genes had a statistically significant (Benjamini–Hochberg adjusted  $P < 0.05$  from permutation test) proportion between—9% and 100%—of their allelic variation explainable by overall *B. pendula* population structure (**Supplementary Table 18**), suggesting that these particular sweeps are at least partially confounded with drift processes. After controlling for population structure, we identified genes showing clinal variation associated with general environmental variables such as temperature and precipitation. These restricting criteria resulted in a small subset of six genes with significant associations and intriguing molecular functions. These genes were verified by phylogenetic analysis as orthologs of the *A. thaliana* genes *SWEETIE*, *KAKTUS* (*KAK*), *ARABIDOPSIS RESPONSE REGULATOR 1* (*ARR1*), *MED5A* (encoding Mediator complex protein MED5A, also known as MED33A), *PHYTOCHROME C* (*PHYC*), and *FAR1-RELATED SEQUENCE 10* (*FRS10*).

*SWEETIE* encodes a protein that may have a central role in sugar homeostasis<sup>52</sup>. In *Arabidopsis*, the *sweetie* mutant shows stunted growth, early senescence, flower sterility, and increased sugar levels. In particular, the mutant has high levels of trehalose, a metabolite



**Figure 5** Selective sweep analysis reveals signatures of recent adaptation in the silver birch genome. **(a)** Putative sweep regions (yellow) around *PHYC* and *FRS10* were obtained by accumulating and filtering composite likelihood ratio (CLR) statistics (red). Gene models are shown for each gene (purple), with the corresponding mRNA region highlighted in gray (or blue for genes under selection). **(b)** PCA plot based on SNPs 2 kb upstream and downstream of the sweep region related to *PHYC*. (Sampling locations are color-coded as in Fig. 2). **(c)** RDA plot of the *PHYC* SNP region. Axis RDA1 is the direction that best correlates with principal components of maximum temperature ( $T_{max}$ ). In this case,  $T_{max}$  explains 26.3% of the variation in the SNP data; the remaining variation is explained by PCA, where PC1 explains 50.5% (y axis). Inset, differences in yearly  $T_{max}$  between Yakutsk and Krasnoyarsk, the locations most separated along RDA1.

associated with signaling in plant interactions with microbes and herbivorous insects, and in responses to cold and salinity. Additionally, *sweetie* shows altered expression for late embryogenesis-abundant (LEA) genes and many DREB2-type TFs. LEAs are anti-aggregation proteins that together with trehalose protect plant cells during the dehydration typical of abiotic stresses such as cold and drought<sup>53</sup>, whereas DREB2A and DREB2B are key transcription factors regulating responses to dehydration and high-salinity stresses<sup>54</sup>. In birch, DREB TFs have been associated with cold acclimation and winter hardiness<sup>55</sup>. These connections may relate to the strong correlation silver birch *SWEETIE* shows with environmental variables.

KAK was identified originally as an endoreduplication repressor in *Arabidopsis* trichomes. However, the *kak1* mutant shows increased C-values in etiolated hypocotyls completely lacking trichomes, suggesting a broader role in the control of endoreduplication<sup>56</sup>. KAK is also expressed in cambium, a secondary meristem that gives rise to both phloem and secondary xylem, where the gene has been suggested to have a role in defining the balance between xylem and phloem formation during vascular development<sup>57</sup>. In leaves, endoreduplication is associated with an increase in cell size and rapid growth, and also higher stress tolerance<sup>58,59</sup>. If KAK is indeed a general regulator of endoreduplication, its correlation with temperature may be of adaptive significance to silver birch.

Together with cellulose and hemicellulose, lignin is an essential component of the secondary cell walls in structural fibers and water-conducting cells, determining their strength and rigidity. Lignin also interferes with the separation and breakdown of cellulose, hindering pulp and paper production and limiting the use of biomass crops for biofuel production. Attempts to reduce lignin production through genetic manipulation have so far resulted in plants with stunted growth and reduced yields<sup>60</sup>. MED5A was recently associated with lignin formation; the *med5a* mutation rescued the stunted growth, collapsed xylem vessels, and lignin deficiency phenotypes in the *Arabidopsis* phenylpropanoid pathway mutant *ref8-1* (ref. 61). The double mutant *med5a ref8-1* showed alleviated phenotypes, but cell wall properties were not restored to wild-type composition. Birch MED5A appears to be under positive selection, as several amino acid positions were significant by Bayes empirical Bayes positive selection analysis (Supplementary Note and Supplementary Table 20), its Tajima's *D* value was in the lower 10% quantile for the *B. pendula*

genome, and its polymorphism correlated with latitude–longitude as well as temperature (Supplementary Table 18). A second component of the same mediator complex that appears among the putatively swept genes is MED12, which is involved in flowering time regulation in *Arabidopsis*<sup>62</sup>.

Cytokinin signaling is of pivotal importance for plant vascular development<sup>63,64</sup>; it is a major positive regulator of cambium activity and controls wood formation in the tree trunk<sup>65</sup>. We identified in our sweep list the birch ortholog of *Arabidopsis* transcription factor ARR1, a key regulator of *Arabidopsis* root meristem size<sup>66</sup> that mediates the balance between cell division and differentiation by integrating auxin and cytokinin responses. ARR1 is involved in cold-induced inhibition of root growth and reduced auxin accumulation<sup>67</sup>, and it also controls *Arabidopsis* drought susceptibility<sup>68</sup>. This may explain the link to the geographic and temperature variables detected here. Variation in cytokinin signaling appears to have a large role in local adaptation in silver birch, as the set of putatively swept genes also includes *AHK2*, *AHK3*, and *AHK4*, and their GO category (peptidyl-histidine phosphorylation) was enriched, as described above.

Finally, the orthologs of *PHYC* and *FRS10* are closely linked in the silver birch genome (Fig. 5), which is also the case in the grapevine and tomato genomes (Supplementary Fig. 21). *PHYC* and *FRS10* act in red and far-red light sensing, shade avoidance, canopy density, temperature-dependent adaptation, and flowering time regulation<sup>69</sup>. *PHYC* was recently connected to temperature-specific regulation of the circadian clock<sup>70</sup>, and in *Arabidopsis* it is strongly linked to altitudinal<sup>71</sup> and latitudinal–longitudinal<sup>70,72,73</sup> clines in flowering time. In addition to *PHYC*, *FRIGIDA* and *FLC* explain a large proportion of flowering time variation in *Arabidopsis*<sup>71</sup>. Birch homologs of their *Arabidopsis* regulators FES, TXR7, and VEL1 were included in the selective sweep gene set. Because bud burst and initiation of flowering depend mainly on (night) temperature<sup>74</sup>, the function of *PHYC* in birch may be related to the photoperiodic control of inflorescence initiation in the autumn, growth cessation, development of cold tolerance, and induction of senescence. *PHYA* and *PHYB*, encoding the other two main phytochromes, were also identified among the putatively swept genes, emphasizing the importance of light sensing for tree adaptation to varying environments. Phenotypic correlates of clinal variation in these and other genes remain obscure but are certainly worthy of more targeted analyses.



## DISCUSSION

Using the *B. pendula* reference genome and resequenced individuals spanning the geographic range of silver birch, we were able to characterize genomic adaptations at several levels. First, we detected enrichments of TF functions that date to the core-eudicot crown radiation. Second, we uncovered a suite of gene duplicates involved in environmental responses that were not polyploidy derived but instead stem from ongoing tandem duplication processes. Such duplicates are generated by the same mechanisms as copy number variants (CNVs), which have come under intense recent study (particularly in animal genomes) as adaptive “tuning knobs” at the inter-population level<sup>75</sup>. In the case of silver birch, the tandem duplicates we observed might be taken as a sort of ‘species average’ that reflects former CNVs fixed by selection and neo- or subfunctionalization.

After several bottlenecks at well-known times of global environmental change, the effective population size of silver birch has increased over the past 1 million years. As expected for a wind-pollinated species with high pollen dispersal, population structure across the species range was particularly weak, which should greatly facilitate future GWAS efforts. Although we found evidence of ongoing gene flow between birch species, they were still clearly separated, and even if hybridization and introgression occurred, it did not blur their genetic distinctiveness. This contrasts with birch cytoplasmic markers, where evidence for allele sharing is common and species limits weak. Similar discordance between nuclear and cytoplasmic markers has been observed in other plant species<sup>76</sup>.

To identify recent selection in silver birch, we analyzed selective sweeps at the intraspecific level. These acted mostly on genes dating from the ancient gamma hexaploidy event. Many of the genes located in the sweep regions were regulators and receptors that hold key positions in triggering developmental or physiological chains of events, suggesting that selection has acted during birch speciation by tuning the timing and cross-talk between different processes. However, tandemly duplicated genes were not enriched for sweeps, and recently duplicated birch-specific genes were significantly depleted of sweeps. Altogether, these findings suggest an ‘exploration–exploitation’ model for tandem duplicates in species evolution. Exploration would occur through generation of novel tandem CNVs within populations; with lineage splitting and lowering of  $N_e$ , polymorphic tandems may become fixed by drift. Another alternative is fixation by selection through a process analogous to soft sweeps. As often reported in mammalian genome analyses<sup>77</sup>, multiple alleles at a locus can be swept to fixation in a ‘soft’ event that evades detection by ordinary criteria. In our example, tandem CNVs among individuals would comprise the alternative (exploratory) ‘allelic’ states, perhaps maintained in populations by balancing selection<sup>75</sup>. Multiple CNV states could then be simultaneously selected for when opportunistic and rapid (i.e., exploitative) responses to environmental change are required. Strong (‘hard’) selective sweep patterns, in contrast, involve selection for a single allelic state and may be more likely among core regulatory components that coordinate developmental timing and physiological cross-talk. Here, particularly when intertwined with population bottlenecks engendered by environmental upheaval, perhaps only genotypes that are unique and have proper timing of responses can be exploited.

We further uncovered candidate genes that show selection associated with environmental responses and that are enriched for functions of practical relevance for forest biotechnology. Notably, several key components of cytokinin signaling, a major positive regulator of vascular cambium activity and radial tree-trunk growth<sup>65,78</sup>, show evidence of recent natural selection. Other examples are *KAK* and

*MED5A*, which can elicit pleiotropic growth and cell wall phenotypes in *Arabidopsis*. If orthologs of these genes function similarly in birch, this information could be used for selective engineering of forest trees for rapid generation of biomass. Similarly, natural variation in photoperiod regulation might be used to understand and alter cambial activity–dormancy cycling and wood production.

## METHODS

Methods, including statements of data availability and any associated accession codes and references, are available in the [online version of the paper](#).

*Note: Any Supplementary Information and Source Data files are available in the online version of the paper.*

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## AUTHOR CONTRIBUTIONS

J.K., P.A., and Y.H. conceived the study, and led the work together with J.S. and V.A.A. J.S. managed and coordinated all bioinformatics activities. L.G.P., A.L., and J.I. performed library construction and sequencing, and R.H., J.I., K.N., J.A.S., C.T.K., C.v.d.S., L.V., M.R., E.O., J.M., S.K.-S., G.E., A.P.M., and B.J.H.M.P. participated in various aspects of biological sample collection, preparation, and quality control. O.-P.S. assembled the genome, with P.A. leading the work. O.-P.S. assembled and P.S. and A.A. annotated the organellar genomes. P.R. carried out linkage mapping and anchored the genome into pseudochromosomes with J.S., O.-P.S., S.R., P.S., J.I.K., O.S. and A.A. J.I., R.R., L.K., A.W., T.P., P.I.H., L.G.P., E.T.P., Y.H., and J.K. produced EST libraries, which were analyzed by O.-P.S., J.S., and S.R. J.I. and L.G.P. produced RNA-seq libraries, and O.-P.S., B.J., and S.R. analyzed the RNA sequencing data. J.S. and S.R. contributed to functional annotation, and P.J.G., H.F., J.V., T.H.T., S.E., A.V., A.K., M.W., S.O.K., A.G., A.S., V.P., M.M.R., M.B., M.P., S.K., O.S., T.B., K.M., K.H., J.P.K., S.S., F.O.A., P.H., F.C., K.V.F., K.-J.L., P.S., V.H.A., L.M., P.E., S.B., D.B., E.X., T.P.S., J.A.S., K.O., O.-P.S., O.B., A.K.K., J.I.K., M.S., A.I.T., J.L., J.d.D.B.-L., P.P., L.P., Ü.N., S.E.E., and K.N. participated in manual annotation, which was coordinated by P.S. and J.S. J.T. and A.H.S. annotated and analyzed the transposable elements. J.S., V.A.A., T.L., and S.R. performed comparative genomics analyses. J.S., V.A.A., O.S., and M.L. analyzed population genomics data. J.S., V.A.A., and T.L. prepared figures. J.S. and V.A.A. wrote the paper, with input from J.K., M.L., K.N., Y.H., P.A., A.H.S., O.-P.S., C.D., and J.I. J.S., O.-P.S., V.A.A., L.G.P., S.R., A.L., R.H., J.I., O.S., K.N., M.R., A.H.S., P.R., P.S., A.A., and G.E. provided text for the supplement. K.N. and S.R. had equivalent overall roles in the project. All authors approved the final version of the manuscript.

## COMPETING FINANCIAL INTERESTS

The authors declare no competing financial interests.

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## ONLINE METHODS

**Sequencing and genome assembly.** For more detailed descriptions of the methods, see the **Supplementary Note**. The individual for the reference genome was a fourth-generation inbred line from the inbred collection of the Natural Resources Institute Finland. It originated from a seed collected in 1967 from an open-pollinated natural stand in Jäppilä, central Finland. Eleven microsatellite markers were used to assess the success of inbreeding. Several 454 libraries were constructed from leaf or cambium tissue for the initial single-end sequencing. Genomic DNA (5 µg) was sonicated on a Covaris S2. End repair, A-tailing, and adaptor ligation were done according to the manufacturer's protocol. Libraries were amplified using KAPA HiFi DNA Polymerase or Phusion HiFi DNA Polymerase. The amplified libraries were purified using AMPure XP beads and size-selected using carboxy-beads on a Magnatrix 1200 robot. Libraries were sequenced on a Genome Sequencer FLX and FLX+ platforms. Libraries for Illumina paired-end sequencing were constructed using 1 µg of starting material sonicated on a Covaris S2. After end repair, A-tailing, and ligation of Y-adaptors according to the manufacturer's instructions, PCR was performed using Phusion HiFi DNA Polymerase. PCR products were purified with AMPure XP beads and size selected. Insert libraries of 350 bp and 500 bp were generated and sequenced on HiScan SQ (100 + 100 bp) and MiSeq (250 + 250 bp) sequencers. For pooled sequencing, barcoded libraries (61 + 2) were prepared and pooled, size selected to inserts of 500 bp, and sequenced with a NextSeq 500 (150 + 150 bp). A SOLiD Mate-Pair kit was used to construct four different sized mate-pair libraries. For the 2-kb and 3-kb mate-pair libraries, Illumina Y-adaptors were ligated to the captured fragments and sequenced on a HiScan SQ Sequencer. For 4-kb and 5-kb mate-pair libraries, SOLiD adaptors were ligated and sequenced on a SOLiD 5500xl Sequencer. Library construction for PacBio sequencing was carried out using the manufacturer's protocols. Genomic DNA was sheared using a Megaruptor, followed by damage repair, end repair, hairpin ligation, and size selection using BluePippin. After primer annealing and polymerase binding, the DNA templates were sequenced on a PacBio RSII sequencer using P4/C2 chemistry and 120-min video time and later using P6/C4 chemistry and 360-min movie time. Contigs were assembled from 454 sequencing data using Newbler with large genome parameter settings. The accuracy of the contigs was improved using the Pilon software with Illumina paired-end sequences. Scaffolding was done using the Illumina and SOLiD mate-pair libraries and SSPACE 2 in a hierarchical manner; first the 2 kb library was mapped to the contigs, then contigs were scaffolded accordingly. Then the procedure was repeated with 3 kb, 4 kb, and 5 kb libraries. Before mapping and scaffolding, adaptors were removed using Cutadapt, and the reads were trimmed using Sickle with a Phred cutoff of 20 for paired-end data and 28 for mate-pair data. The SOLiD mate-pair libraries were converted to FASTQ sequences using XSQ-Tools. The scaffolds were further improved with PBjelly and ~30× of PacBio sequence data with reads longer than 6 kb. The scaffolds were ordered into super-scaffolds using SSPACE-LongRead with a second set of PacBio sequencing data with approximately 20× coverage and read lengths of 10–75 kb (**Supplementary Fig. 1** and **Supplementary Table 1**). Organellar genomes were assembled as part of the genomic assembly. Chloroplast contigs were connected using Gap4 and verified with PacBio data (**Supplementary Figs. 3 and 4**). A separate PacBio assembly was carried out for the mitochondrial genome using a modified Newbler assembler (**Supplementary Note** and **Supplementary Figs. 5 and 6**).

**Linkage mapping.** For linkage mapping, low-coverage whole-genome sequencing data from 63 individuals were mapped to the genome using BWA-mem, and genotype posterior probabilities were calculated by the Lep-MAP3 pipeline using the output from samtools mpileup. The relatedness of individuals was checked using the IBD module of Lep-MAP3, and then the parental genotypes for each potential marker were estimated, providing the final data set for linkage mapping. Markers segregating identically on each contig were joined and collapsed into linkage groups separately for markers informative only paternally, maternally, or both. For each combination of parental markers, 14 linkage groups were found. The number of linkage groups matched the karyotype of the species ( $2n = 28$ ), and thus can be considered pseudochromosomes.

Scaffolds were assigned to chromosomes by requiring a region of 4 kb including 8 markers assigned to the same linkage group; markers not fulfilling

this criterion were removed. Contigs were assigned to intervals that were then mapped to a chromosome. Finally, two maps were constructed and ordered for each chromosome on the basis of paternally informative or maternally informative markers. The orderings were inspected and corrected manually, including orientation of the two maps in the same direction. Finally, the linkage groups and scaffolds assigned to them were arranged to form 14 pseudochromosomes. During the scaffold placement, super-scaffolds showing signs of possible chimeric assembly were split from the regions between markers and placed in their respective linkage groups.

**Assembly validation.** The quality of the assembly was monitored during all stages and using several independent validation methods. Contig assembly was first tested using Scarpa software and the 2-kb Illumina mate-pair library and then validated by mapping a set of high-quality ESTs to the contigs using Exonerate. The completeness of the scaffold-level assembly was tested by mapping the unassembled Illumina reads against the scaffolds and by mapping Trinity *de novo* assembled RNA-sequencing transcripts to the assembly with PASA, as well as through CEGMA. An independent validation of the scaffold-level assembly quality was gap-filling with long PacBio reads. The assembly was additionally evaluated with REAPR using paired-end libraries for error calling and 3 kb mate-pair Illumina libraries for estimating the library breakpoints. To assess the quality of the scaffolding, the assembly was also aligned to long, nonoverlapping PacBio-reads using BLASR by inspecting 12 reads of length 70–75 kb that had at least 20,000 bp 'best hit' mapping (**Supplementary Fig. 2**), with total sequence length of 863 kb. Final tests for assembly quality were carried out using QUAST (**Supplementary Table 2**).

Repeats in the contigs and pseudochromosomes were analyzed using RepeatModeler and RepeatMasker. For LTR retrotransposons, a library of elements was built by combining the output from Repet, LTRharvest/LTRdigest from genomtools, our own custom pipeline, and a library of LTR retrotransposons. For MITEs, MITE-Hunter and RSPB were used. CACTA elements were mined on the basis of structural considerations. RepeatMasker was used to mask TEs using the classified repeat libraries. See **Supplementary Table 8** for a summary of the TEs.

**Sample preparation.** The tissues for EST libraries (**Supplementary Table 3**) were frozen in liquid nitrogen and stored at  $-80^{\circ}\text{C}$  until RNA extraction. Total RNA was isolated, and cDNA libraries were constructed with ZAP-cDNA Synthesis Kit and ZAP-cDNA Gigapack III Gold Cloning Kits according to the manufacturer's instructions. For sequencing, individual phages were excised to pBluescript SK plasmids *in vivo*. The EST clones were grown in colonies in 96-deep-well plates, and aliquots were taken for PCR amplification using universal primers from the excised pBluescript SK plasmids. Following purification of the PCR products, sequencing was performed using T3 primer and BigDye chemistry and analyzed on an ABI 3700 sequencer. Base calling was performed using Phred quality threshold value  $\geq 20$ . EST libraries were also sequenced using the 454 and MiSeq sequencing platforms (**Supplementary Table 4**). PCR products from the libraries were pooled, and a 454 library was constructed and run on a FLX+ platform. From the glycerol stocks, PCR was performed with universal primers in 384-well plates in 20-ml volumes. After amplification, 5 µl from each plate was pooled and purified. A MiSeq library was constructed from the pool and sequenced as paired-end (300 bp + 300 bp). In the assembly, MiSeq paired-end reads were first overlapped and combined using FLASH, and then combined with existing Sanger reads from the sequenced EST library, followed by assembly using Newbler (**Supplementary Fig. 7**).

For the RNA sequencing, 12 genotypes were selected from southern and central parts of Finland. Birches were grown for 1 year and exposed to 150 nL L<sup>-1</sup> of O<sub>3</sub> for 8 h, and leaves were sampled at 0, 2, 8, 24, 48, and 96 h. Total RNA from leaves was isolated and pooled on the basis of clone tolerance to ozone (sensitive, tolerant, and moderate), and mRNA was isolated using the Dynabeads mRNA Purification Kit. RNA-seq libraries were constructed using a TruSeq stranded mRNA kit. mRNA (5 µl) was converted into cDNA using random hexamers, and the second strand was synthesized using DNA polymerase I and dUTP nucleotides. After purification, ends were repaired, and after A-tailing, Y-adaptors containing indexes from the kit were ligated to the fragments. After PCR amplification, the fragments were purified using AMPure XP, pooled in equal concentrations, and paired-end (100 bp + 100 bp)



sequenced on a HiScan SQ sequencer. See **Supplementary Table 5** for a summary of the reads. Preprocessing was done with Cutadapt to remove adapter sequences and to trim reads for low-quality sequence, ambiguous sequences, or poly(A) tails, using Phred threshold 33 and minimum length 60 bp. After preprocessing, paired-end information for each sequence was retrieved and orphaned reads were kept as single reads. A *de novo* transcriptome assembly was carried out using Trinity with 12 different *k*-mers. Sequences from all *k*-mer assemblies were pooled and merged using the Amos pipeline. Contigs shorter than 200 bp were discarded. To minimize redundancy, the final assembly was mapped to the ESTs, and fragments identical to particular ESTs were removed. See **Supplementary Figure 8** for the assembly pipeline. The preprocessed reads were also aligned to the genome using TopHat, and the resulting BAM files were processed using HTSeq to obtain the final gene set.

**Gene annotation.** A modular gene annotation framework was established to estimate gene models by combining evidence from *ab initio* gene predictors based on Hidden Markov Models, spliced transcript evidence from RNA-seq and EST data, protein evidence from orthologous proteins obtained from closely related and model plant species, and manually curated gene sequences. Gene prediction was carried out in two phases. The first phase (**Supplementary Fig. 9**) involved generation of an initial set of automated predictions, followed by manual curation of genes from select gene families to provide verified gene models for the second phase of predictions. RNA-seq transcripts generated using Trinity were splice aligned against the unmasked birch genome using PASA in a strand-specific manner to generate long ORFs. For the first phase of the training, these ORFs were directly used for training the *ab initio* gene predictors and obtaining an initial set of gene models. The EST libraries were splice-aligned against the unmasked birch genome in a similar manner and used as additional evidence. Four Hidden Markov Model-based *ab initio* gene predictors were used: Augustus, SNAP, GlimmerHMM, and GeneMark-ES. Gene prediction parameters were trained for the Augustus, SNAP, and GlimmerHMM using the nonredundant final training set from PASA-aligned RNA-seq transcripts. For Augustus, optimization of the parameters was carried out by eightfold cross-validation. For SNAP and GlimmerHMM, the parameters were optimized by carrying out one round of training with the training set. GeneMark-ES did not require an explicit training data set. Augustus, SNAP and GlimmerHMM were run on the masked birch genome using the optimized parameters, whereas GeneMark-ES was executed on the unmasked genome. Orthologous protein sequences from *A. thaliana*, *Populus trichocarpa*, and *Prunus persica* were splice-aligned against the unmasked birch genome using exonerate to obtain spliced protein evidence. EVIDENCE Modeler was used to generate a single high-confidence gene model set, using *ab initio* gene predictions, spliced protein alignments, and spliced transcript alignments as inputs. Finally, UTR addition was carried out by running PASA with the RNA-seq and EST data.

Manual annotation was carried out using the WebApollo software (see **Supplementary Table 6**) for the list of manually annotated genes. For the second phase of the training, ORF generation using PASA was followed by filtering for complete ORFs and combining with core eukaryotic genes from CEGMA plus manually annotated genes to generate a nonredundant training set for the gene predictors (**Supplementary Fig. 10**). Gene predictors were trained as in the first phase and by passing the new parameters. Finally, UTR addition using PASA was re-executed. The manually curated genes and automated annotations were compared for overlaps using the overlap software. Annotations not having any overlap with the manual annotations were extracted and merged with the manually curated genes, thereby creating a single nonredundant set of merged annotations (**Supplementary Table 6**).

**Protein functional analysis.** Protein functional analysis was carried out using Interproscan. Descriptions for the gene models were generated using AHRD with a custom database of *A. thaliana* and SWISS-PROT proteins. Finally, gene model predictions of 200 conserved single-copy genes were verified manually in CoGe by comparing orthologous birch gene models against *A. thaliana* and *Vitis vinifera* homologs.

**Comparative genomic analyses.** Comparative genomic analyses for *B. pendula* were carried out against *Arabidopsis thaliana*, *Coffea canephora*, *Populus*

*trichocarpa*, *Prunus persica*, *Solanum lycopersicum*, *Theobroma cacao*, and *Vitis vinifera* using OrthoMCL for ortholog clustering (**Supplementary Tables 7 and 10** and **Supplementary Data Set 1**). *Arabidopsis* transcription factor family assignments were used to select the proper settings for the inflation parameter in Markov clustering to balance the precision versus recall rate in orthogroups (*F*-score, see **Supplementary Fig. 11**).

Syntenic path alignment (**Supplementary Fig. 12**) and syntenic depth analysis (**Supplementary Table 9** and **Supplementary Fig. 13**) were carried out in CoGe by aligning the *B. pendula* pseudochromosomes to the chromosome-level *Vitis vinifera* genome using default parameters. Self-to-self SynMaps were generated within CoGe using the Quota Align algorithm with default parameters. DAGchainer provided the list of syntenic duplicates (**Supplementary Data Set 2**). Blast2raw, incorporated into CoGe's SynMap pipeline, was used to calculate tandem duplicates (**Supplementary Data Set 2**). Overlaps between orthogroups and syntenic and tandem gene sets were tested using Fisher's exact test (**Supplementary Fig. 14** and **Supplementary Table 11**). For GO enrichment analyses, syntenic gene pairs and tandem duplicates for *B. pendula*, *P. trichocarpa* and *A. thaliana* were downloaded from CoGe and used as the foreground subset in GO enrichment analysis with GOATOOLS, using all the genes in each respective genome as background and Bonferroni-adjusted  $P < 0.05$  as threshold for significance (**Supplementary Table 12**). Gene models from *B. pendula* and *P. trichocarpa* were annotated with the highest alignment score matches using tblastx versus *Arabidopsis* coding sequences with an *E*-value cutoff of  $1 \times 10^{-5}$ . Whole-genome backgrounds for *B. pendula* and *P. trichocarpa* were custom generated by selecting the sets of genes, for each respectively, that were annotatable against *Arabidopsis* genes, accepting the topmost hit as the match (**Supplementary Data Set 2**).

**Population analyses.** For the population analyses, individuals from six natural Finnish populations were sampled from Punkaharju, Loppi, Vehmersalmi, Posio, Rovaniemi, and Kittilä. Additionally, leaf or DNA samples were obtained from various natural sites in Europe and Russia. Individuals of different birch species were obtained from Helsinki Botanical Garden, *Alnus glutinosa* from Valkeakoski and *Alnus incana* from Huittinen, Finland. Finally, eight trees representing special horticultural forms were selected (**Supplementary Table 13**). For sequencing, DNA from leaf, bud or cambium tissue was isolated using the E.Z.N.A. SP Plant DNA Kit (Omega Bio-tek) whereafter Illumina paired-end libraries were constructed. After PCR amplification, the libraries were pooled and size-selected to an average of 500 bp. Paired-end sequencing was performed on a NextSeq 500 sequencer (150 bp + 150 bp). In addition to the resequenced population, the raw reads of a second *Betula nana* individual from Scotland sequenced earlier was retrieved from the Sequence Read Archive (ERP008033).

After quality control using FastQC, adaptors and low-quality bases were removed from the read ends using Trimmomatic (leading and trailing Phred score <20), and the reads were filtered with a sliding window of size 3, with average Phred score threshold of 15 within the window. Reads < 35 bp were removed, followed by a second round of quality control using FastQC. The trimmed reads were mapped to the unmasked *B. pendula* genome using Bowtie2 with default parameters. The mapped reads were sorted, and duplicated reads were removed using SAMtools (**Supplementary Table 14**). HaplotypeCaller from the Genome Analysis Toolkit GATK was used to estimate the general variant calling file for each individual, and then combined by GenotypeGVCFs to a single variant calling file.

Hard filtering of the SNP calls was carried out with Fisher strand bias (FS > 60.0), mapping quality MQ < 40.0, and thresholding by sequencing coverage based on minimum coverage (DP < 100) and maximum coverage (DP > 1,500). The SNPs were annotated with SnpEff (**Supplementary Table 15**). Linkage disequilibrium analysis was conducted using PLINK (**Supplementary Fig. 19**). For population analyses this set was further filtered to include only fourfold degenerate neutrally evolving sites.

The neutrally evolving SNPs were used to estimate population structure by principal component analysis in EIGENSOFT. Ancestral population structure was estimated from the same SNP set with ADMIXTURE using ancestral population sizes  $K = 1 \dots 10$  and choosing the population with smallest leave-one-out-validation error (**Supplementary Fig. 15**). For further analysis, sample-wise admixture proportions were averaged into site-wise averages.



For analyses of introgression, three-population  $F_3$  tests were run for all population configurations using ADMIXTOOLS (Supplementary Table 16) and adjusted for multiple testing using Benjamini-Hochberg correction (Supplementary Note).

**Flow cytometry.** Flow cytometry (Supplementary Fig. 16) was performed using a BD LSR II flow cytometer as per instrument manufacturer's instructions. Cambium was isolated from frozen tree branches by free-hand section and chopped with woody plant buffer. The samples were filtered through 40- $\mu$ m filter and stained with propidium iodide (PI) 50  $\mu$ g ml<sup>-1</sup> simultaneously with RNase at 50  $\mu$ g ml<sup>-1</sup>. Greenhouse-grown rice leaf samples were used as reference.

**Population data analysis.** A phylogeny for nine representatives of different birch species was estimated using SNPs called with GATK and hard filtering criteria. Only SNPs within 2 kb from coding regions and at neutrally evolving sites within coding regions were used. Phylogeny was estimated using SNPhylo, with SNPs having minor allele frequency >0.1. For ancestral state estimation, Illumina resequencing reads mapped against the birch reference genome were used to call consensus sequences using a combination of SAMtools, bcftools, and vcftools. The consensus FASTA files were read into R, and the ancestral states estimated using the Phangorn package with maximum likelihood estimation and the GTR model of molecular evolution, where edge lengths were estimated separately for each contig. Marginal reconstruction of the ancestral character states was carried out using a maximum *a posteriori* estimate.

ANGSD was run to obtain the site frequency spectrum, heterozygosity (Supplementary Fig. 17) and sitewise estimates of allele frequency statistics following the protocols recommended by the software authors. The analysis was restricted to the set of 60 non-admixed *B. pendula* individuals with accurate information on their sampling origin. A custom R script was developed to calculate population genetic descriptive statistics for each gene (Supplementary Table 17). To filter out loci with low coverage and low numbers of called SNPs, only regions with >50 called nucleotides were used for computing the gene-wise statistics.

The site frequency spectrum from ANGSD was used to estimate the population history with Stairway plots using 200 bootstrap iterations, and a range of 0.7, 1.0, and  $2.0 \times 10^{-9}$  as the mutation rate, and 10, 20 and 40 years as the generation time (Supplementary Note and Supplementary Fig. 20). Selective sweeps were estimated with SweepFinder2. To prepare the input files for Sweepfinder, ANGSD was used to call allele frequencies for each position, given the ancestral states estimated earlier. Sweepfinder2 was run for each contig having mapping data from all individuals, with a grid size of 200 bp. For processing the composite likelihood ratio (CLR) scores into putative sweep regions, a custom R script was developed to carry out several filtering steps. First, the first three CLR scores in contig ends were set to 0. Then, to smoothe the fluctuations of CLR scores in neighboring sites, the scores were filtered with median filtering using a window size of 5. Then, CLR scores were merged into sweep regions if the neighboring scores exceeded the top 1% of CLR scores (CLR > 26). Regions where only one 200-bp site exceeded this threshold were removed from the analysis. The final score for each sweep region was the sum of CLR scores for the sites in the sweep region. Thereafter, sweep regions were excluded that contained too many nonsequenced positions (more than 10% of Ns in the sweep region  $\pm$  2-kb flanking region on each side). Genes overlapping the sweep region or 2-kb flanking regions on either side of the sweep region were selected as genes putatively under selection. This gene set was then filtered for regions that may show artifacts of sweeps, such as sequences originating from organellar DNA or transposable element insertions (Supplementary Tables 18 and 19).

For verifying orthologies of genes identified by sweep analysis, *B. pendula* genes were used as queries for a NCBI local tblastx against the *V. vinifera*, *Arabidopsis* Col-0, *C. canephora*, *S. lycopersicum*, *P. trichocarpa*, and *B. pendula* coding sequence databases downloaded from CoGe, using E-value cutoff

$1 \times 10^{-10}$ . The 10 topmost hits from each database were translated and aligned together using MUSCLE. The alignments were reverse translated, and ambiguous regions were removed using Gblocks, with stringency parameters to allow smaller blocks, gap positions within the final blocks, and less strict flanking positions. Phylogenetic analyses based on the alignments were performed using RAXML-HPC with the GTR substitution model (Supplementary Data Set 3).

To analyze whether a sweep could be explained by drift and population structure, a PCA of each of the sweep regions and 2-kb flanking regions was carried out for SNPs called by ANGSD. The principal components of the sweep regions were computed in R using data from ANGSD.

Weather information for each location was extracted from the National Center for Environmental Prediction (NCEP) website. Daily minimum, maximum, and average temperatures and precipitation information were downloaded based on the GPS coordinates of the closest weather station and processed into monthly averages per location. Redundancy analysis (RDA) was used to decouple the population structure from associations with environmental variables by using the first two principal components from overall population structure as covariates. For the climate data, principal component analysis was carried out in a similar manner using the (months  $\times$  locations) matrix, and the first two principal components were extracted for minimum, maximum, and average temperatures as well as precipitation. The PCs of the climate variables and PCs from the population structure were then used as covariates to model the variation of the dependent variable, in this case the genomic variation around the sweep region and 2-kb flanking windows. The *P* values for the fits were estimated using a permutation test with 30,000 permutations, followed by adjustment for multiple comparisons using Benjamini-Hochberg correction. The coefficient of determination was reported together with the adjusted *P* value. Microsynteny analysis was carried out in CoGe (Supplementary Note and Supplementary Fig. 21).

SAMtools mpileup was used to call the variant sites in the cultivars. Bcftools and vcftools from SAMtools htlib were used to obtain a variant calling file. The variant sites were annotated using snpEff, and loci having a stop codon and a SNP quality score >20 were selected. The mutation in *B. pendula* 'Youngii' (Supplementary Fig. 18) was verified by genotyping; the product was digested with MboI, which at the mutation site cuts the WT gene but not the mutated version.

For molecular evolutionary analyses, repredictions for dubious gene models were conducted using default settings of AUGUSTUS and the genomic sequences of the previously predicted *B. pendula* gene models, plus 500–1,000 bp of upstream and downstream genomic sequence. Tests were conducted on a restricted subset of each gene tree (produced as described above), which in most cases amounted to only orthologous genes from the above species, or in some cases, their paralogs as well. We estimated  $\omega$  (dN/dS) values for each CDS alignment and RAXML (subset) phylogeny using the codeml part of the PAML package. Gaps in the alignment were excluded by PAML. Two types of models were implemented: branch-specific and branch-site. Comparisons of two nested models were performed using a likelihood ratio test to test for the following: asymmetric sequence evolution versus two-ratio model 2, divergent selection (model 3 versus clade model D ( $K = 3$ )), and positive selection (model A null ( $\omega_2 = 1$ ) versus model A ( $0 < \omega_0 < 1$ )). Additionally, Bayes-empirical-Bayes analyses were conducted to estimate individual amino acids that might have evolved under positive selection (Supplementary Table 20).

**Data availability.** OpenStreetMap data are available under the Open Database License (<http://www.openstreetmap.org/copyright>). The genome assembly is available from Ensembl Plants (<http://plants.ensembl.org/>) and the CoGe comparative genomics platform (<https://genomevolution.org/CoGe/GenomeInfo.pl?gid=35079> and <https://genomevolution.org/CoGe/GenomeInfo.pl?gid=35080>). All sequencing data have been deposited in the European Nucleotide Archive (ENA) under accession code PRJEB14544.

## Author Correction: Genome sequencing and population genomic analyses provide insights into the adaptive landscape of silver birch

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Correction to: *Nature Genetics* <https://doi.org/10.1038/ng.3862>, published online 8 May 2017.

In the version of this article initially published, there was a mistake in the calculation of the nucleotide mutation rate:  $1 \times 10^{-9}$  mutations per site per generation was used, whereas the regression formula given by Lynch (M. Lynch, *Trends Genet.* **26**, 345–352; 2010), which was used in the published article, in fact provides a rate of  $9.5 \times 10^{-9}$  mutations per site per generation. This error affects the interpretation of population-size changes over time and their possible correspondence with known geological events, as shown in the original Fig. 4 and supporting discussion in the text, as well as details in the Supplementary Note. Neither the data themselves nor any other results are affected.

Herein, the range of mutation rates and generation times is revised for silver birch to better reflect uncertainties in both parameters. These uncertainties are large in long-lived forest tree species, but the issue at hand is not specific to them. Even in a species as well studied as humans, there is still substantial uncertainty regarding the mutation rate, with a twofold difference between indirect estimates at  $\sim 10^{-9}$  per base pair per year according to divergence, and direct estimates (obtained from pedigrees), in which estimates of approximately  $0.5 \times 10^{-9}$  have been obtained (P. Moorjani, Z. Gao and M. Przeworski, *PLoS Biol.* **14**, e2000744; 2016). The latter rate was used in the article that introduced the stairway plot method (X. Liu and Y.-X. Fu, *Nat. Genet.* **47**, 555–559; 2015). Regarding generation times, in natural stands, birch begins flowering around the age of 10 years. At the other extreme, the maximum age of a birch tree can be up to 120–150 years. An intermediate (average) value between age at maturity and maximum observed life span was used to estimate generation time in the tropical palm *Euterpe globosa* (L. Van Valen, *Biotropica* **7**, 259–269; 1975, and R. J. Petit and A. Hampe, *Annu. Rev. Ecol. Evol. Syst.* **37**, 187–214; 2006). Following a similar strategy would give a maximum generation time of 80 years for birch. The generation-time range was therefore revised in the calculations to the values 10, 40 and 80 years.

Because there are no estimates of the mutation rate available for birch, estimates from other plant species must be used as a proxy. For example, for *Populus trichocarpa*, a mutation rate was calculated on the basis of *Arabidopsis thaliana* mutation rates and fossil evidence (G. A. Tuskan et al., *Science* **313**, 1596–1604; 2006). Studies of mutation-accumulation lines in *A. thaliana* have resulted in estimates of  $7.0 \times 10^{-9}$  to  $7.4 \times 10^{-9}$  mutations per site per generation (S. Ossowski et al., *Science* **327**, 92–94; 2009, and S. Yang et al., *Nature* **523**, 463–467; 2015). However, estimates of *A. thaliana* generation times in nature vary between 1.3 years (M. Falahati-Anbaran, S. Lundemo and H. K. Stenøien, *New Phytol.* **202**, 1043–1054 (2014) and 4 years (S. Lundemo, M. Falahati-Anbaran and H. K. Stenøien, *Mol. Ecol.* **18**, 2798–2811; 2009), depending on the effects of seed banks. Altogether, this would result in a potential range of  $1.75 \times 10^{-9}$  to  $6.7 \times 10^{-9}$  mutations per site per year for *A. thaliana*. The lower bound agrees with a recent estimate from natural populations, wherein rates varied between  $2 \times 10^{-9}$  and  $5 \times 10^{-9}$  mutations per site per year (M. Exposito-Alonso et al., *PLoS Genet.* **14**, e1007155; 2018). Dividing by six, as in Tuskan et al. (G. A. Tuskan et al., *Science* **313**, 1596–1604; 2006) to better agree with fossil evidence, and multiplying by the generation time of birch (by using a range of 10, 40 and 80 years) would result in approximate lower bounds at  $2.9 \times 10^{-9}$ ,  $1.2 \times 10^{-8}$  and  $2.3 \times 10^{-8}$ , and upper bounds at  $1.2 \times 10^{-8}$ ,  $4.5 \times 10^{-8}$  and  $8.9 \times 10^{-8}$  mutations per site per generation, respectively.

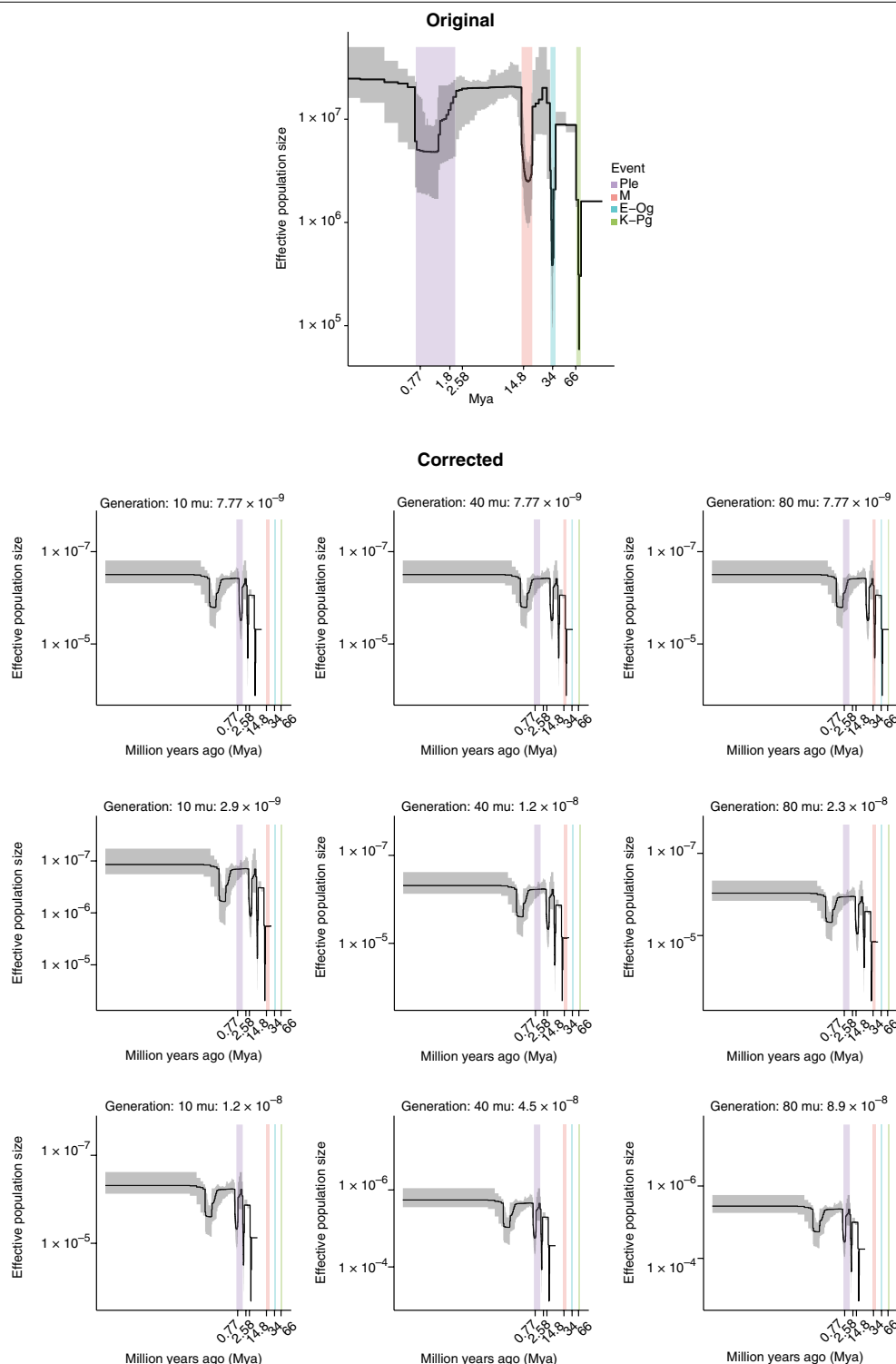
Alternatively, mutation-rate estimates from other woody perennials can be used. In peach (*Prunus persica*), another woody tree species, parent-progeny sequencing has yielded a mutation rate of  $7.77 \times 10^{-9}$  point mutations per site per generation (Z. Xie et al., *Proc. Biol. Sci.* **283**, 20161016; 2016), which is similar to the estimates for *A. thaliana* from mutation-accumulation lines (S. Ossowski et al., *Science* **327**, 92–94; 2009, and S. Yang et al., *Nature* **523**, 463–467; 2015), thus suggesting that mutation rate may not be dependent on generation time, possibly because of purifying background selection (M. Exposito-Alonso et al., *PLoS Genet.* **14**, e1007155; 2018). Thus, one estimate of birch mutation rate can be obtained by simply assuming that the mutation rate per generation is similar across all woody tree species.

To account for the large uncertainty in mutation-rate estimates discussed above, Fig. 4 has been revised to incorporate a range of values. The mutation rate is first plotted by the rate estimated for peach,  $7.77 \times 10^{-9}$ , and then the respective lower and upper bounds for each generation time,  $2.9 \times 10^{-9}$ ,  $1.2 \times 10^{-8}$ ,  $2.3 \times 10^{-8}$ ,  $1.2 \times 10^{-8}$ ,  $4.5 \times 10^{-8}$  and  $8.9 \times 10^{-8}$  mutations per site per generation, respectively. The curve of changes in effective population size ( $N_e$ ) is not altered, but the absolute dates for the  $N_e$  bottlenecks differ according to the assumed mutation rate and generation time. As such, the oldest bottleneck observed may be as young as ~2.6 million years ago (Mya) or as old as 34 Mya, instead of at the Cretaceous–Paleogene boundary, as reported in the original figure. Such a difference of greater than an order of magnitude highlights the problems inherent in all attempts to provide calendar dates for population genomic features, especially in long-lived perennials in which there is vast uncertainty in generation times and mutation rates.

Images of the original and corrected figure panels are shown in the correction notice.



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**Fig. 4 | Historical effective population size for silver birch.** Top, original figure with incorrect mutation rate; bottom, corrected figure. Varying mutation rates and generation times (described herein) were used to estimate historical population dynamics. The stairway plot shows that the *B. pendula* population has undergone bottlenecks that may be associated with some of the four known periods of major climate upheaval: the Cretaceous-Paleogene (K-Pg; green) the Eocene-Oligocene (E-Og; blue), the mid-Miocene (M; red) and the Pleistocene (Ple; purple). Data are median estimates from 200 bootstrap replicates (black lines) and 95% confidence intervals (shading). Tick marks along the x axis show estimates for the Matuyama-Brunhes (0.77 Mya), Calabrian (1.8 Mya) and Gelasian (2.58 Mya) borders, as well as the mid-Miocene disruption (14.5–14.8 Mya) and the E-Og (34 Mya) and K-Pg (66 Mya) events. As can be seen, no set of generation-time and mutation-rate assumptions yields an  $N_e$  dip near the Cretaceous-Paleogene boundary.

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