The mitogenome of Norway spruce and a reappraisal of mitochondrial recombination in plants

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Abstract

Plant mitogenomes can be difficult to assemble because they are structurally dynamic and prone to intergenomic DNA transfers, leading to the unusual situation where an organelle genome is far outnumbered by its nuclear counterparts. As a result, comparative mitogenome studies are in their infancy and some key aspects of genome evolution are still known mainly from pre-genome, qualitative methods. To help address these limitations, we combined machine learning and *in silico* enrichment of mitochondrial-like long reads to assemble the bacterial-sized mitogenome of Norway spruce (Pinaceae: *Picea abies*). We conducted comparative analyses of repeat abundance, intergenomic transfers, substitution and rearrangement rates, and estimated repeat-by-repeat homologous recombination rates. Prompted by our discovery of highly recombinogenic small repeats in *P. abies*, we assessed the genomic support for the prevailing hypothesis that intramolecular recombination is predominantly driven by repeat length, with larger repeats facilitating DNA exchange more readily. Overall, we found mixed support for this view: recombination dynamics were heterogeneous across vascular plants and highly active small repeats (*ca*. 200 bp) were present in about a third of studied mitogenomes. As in previous studies, we did not observe any robust relationships among commonly-studied genome attributes, but we identify variation in recombination rates as a underinvestigated source of plant mitogenome diversity.

Key words: mitogenome, repeats, recombination, rearrangement rates, structural variation

Introduction

Mitochondria share an α-proteobacterium ancestor, a conserved core proteome, and an

almost universal function as the site of cellular energy production (Gray 2014). Despite the broad

similarity of mitochondria across eukaryotes, the vestigial mitogenome is remarkably diverse in

size, content, and architecture (Burger et al. 2003). Nowhere is this heterogeneity showcased

more clearly than in plants: closely-related mitogenomes can vary 100-fold or more in size

(Sloan et al. 2012), substitution rates (Cho et al. 2004), and rearrangement rates (Cole et al.

2018). Gene repertoires are fluid due to recurrent horizontal (Rice et al. 2013; Sanchez-Puerta

2014) and intergenomic transfers (Adams & Palmer 2003). Multichromosomal architectures

(Alverson, Rice, et al. 2011; Sloan et al. 2012) have also been reported from the relatively few

sequenced plant mitogenomes. Underlying the dynamism of plant mitogenomes appears to be the

evolution of pervasive homologous recombination (Palmer & Herbon 1988; Gray et al. 1999).

Recombination is the predominant mode of double-strand break repair in plant

mitogenomes (Maréchal & Brisson 2010; Gualberto & Newton 2017). In contrast, animal

mitogenomes tend to use non-homologous end joining pathways or may simply degrade

damaged molecules (Alexeyev et al. 2013). Recombination can preserve sequence identity, but

reliance on this pathway may lead to unsuppressed intramolecular recombination at dispersed

repeats (Maréchal & Brisson 2010). While these differences in repair pathways likely contribute

to the broad mitogenome differences among eukaryotic kingdoms, the importance of

recombination and other mechanisms in generating the diversity within plants is unclear. Our

limited understanding stems, in part, from the few available assembled mitogenomes.

While sequencing plant genomes is routine, only one mitogenome is published for every

four nuclear genomes (NCBI Genome Resource; accessed March 11, 2019). Often, mitogenome

assembly requires isolating DNA from intact mitochondria because frequent intergenomic

transfers (e.g. Alverson, Rice, et al. 2011), the rapid decay of intergenic sequence homology

(Guo et al. 2016), and the unclear physical organization of the mitogenome (Sloan 2013) can

preclude identifying mitochondrial sequences from whole genome read data (e.g. Nystedt et al.

2013). Strategies to identify mitogenomic scaffolds using GC-content and genome copy number

can be effective (e.g. Naito et al. 2013), but this simple strategy becomes prone to false positives

and low recovery rates with increasingly large, complex genomes (Eldfjell 2018).

We used machine learning to assemble the bacterial-sized mitogenome of a coniferous

forest tree, Norway spruce (Pinaceae: *Picea abies*), from whole-genome shotgun sequencing

reads. The *P. abies* mitogenome helps to fill a phylogenetic gap in comparative analyses, and to

that end we analyzed gene repertoires; sources of genome size heterogeneity; intraspecific

variation; and mutation, recombination, and rearrangement rates in gymnosperms. Prompted by

the detection of highly recombinogenic small repeats in *P. abies*, we reevaluated published

recombination rates in seed plants. Despite early recognition of recombination as a factor in

generating the diversity of eukaryotic mitogenomes (Palmer & Herbon 1988; Gray et al. 1999),

surprisingly little attention has been given to recombinational dynamics as a source of

mitogenomic diversity within plants. As in previous studies, we found no clear relationship

between mitogenome traits and potential mechanisms in – for example, genome size and the

proportion of intergenomically transferred DNA – but the role of recombination as a driver of

mitogenomic diversity within plants merits further scrutiny.

Results and Discussion

Mitogenome assembly

We combined machine learning, *in silico* enrichment of long sequencing reads, and assembly

reconciliation to produce a highly contiguous *P. abies* draft mitogenome (Fig. 1). First, we

developed a support vector machine (SVM) to identify mitochondrial-like scaffolds from the *P.*

abies v. 1.0 genome assembly (Nystedt et al. 2013). Training the SVM classifier requires

positive and negative training data and a set of variables – that is, genomic features – expected to

discriminate between mitochondrial scaffolds and those from other genomes (Fig. 1). In brief, we

used the whole genomes of *Arabidopsis thaliana* and *Populus trichocarpa* and 18 organelle

genomes as training data and calculated depth of coverage, guanine and cytosine (%GC) content,

and a kmer score based on the frequency of nonredundant 'mitochondrial-like' oligonucleotide

sequences in each scaffold for use as genomic features (see Methods).

After training the SVM, we applied it to a reduced set of the *P. abies* v. 1.0 assembly

(Nystedt et al. 2013) comprising 49,500 scaffolds (see Methods). Of these, the SVM identified

301 scaffolds totaling 4.69 Mb as potentially mitochondrial in origin. While the SVM achieved

higher resolution than simple coverage *vs.* GC-content plots (Eldfjell 2018), the recall and false

discovery rate (FDR) indicated this classification alone would be insufficient to identify the

mitogenome from the *P. abies* v. 1.0 assembly. Recall ranged from 0.53 to 0.92 and the FDR

from 0.13 to 0.22 with varying test-to-

training ratios (Table S1), although SVMs

achieved more accurate classifications in

our tests with other draft gymnosperm

assemblies (Eldfjell 2018).

We then used the SVM classified

scaffolds as 'bait' to enrich genomic

Pacific Biosciences (PacBio) Sequel®

reads for mitochondrial-like sequences *in*

silico through kmer matching (Fig. 1). We

screened *ca.* 35.5 million subreads totaling

244 Gbp for 27-mer matches to any of the

SVM-classified scaffolds, which yielded

3.23 reads totaling 4.1 Gb, with an average

read length of 12,041 bp. We reasoned that

the long read length could enable us to

recover a greater proportion of the

mitogenome, as only a single 27-mer

match is required, while the improved

contiguity of the assembly could allow

removal of dubious contigs post-assembly.

We assembled the enriched PacBio

reads using three assemblers selected

based on their prior performance

(Jayakumar & Sakakibara 2017; Giordano

et al. 2017; Fig. 1). The most contiguous

results per assembler are summarized in

Table 1. Upgrading the initial assemblies

with unused overlap information in the

corrected PacBio reads with FinisherSC

(Lam et al. 2015) only appreciably

Figure 1. Strategy used to assemble the mitogenome of *Picea abies*. First, a support vector machine (SVM) was trained to identify mitochondrial-like scaffolds from the *P. abies* genome assembly. We used the classified scaffolds to identify PacBio Sequel® subreads containing mitogenome-like 27-mers. These enriched reads were then assembled using three different pipelines. Scaffolds from each assembler with at least one mitochondrial protein coding gene were retained for assembly reconciliation and base-pair correction, thus yielding the final mitogenome draft.

improved the N50 of the canu (Koren et al. 2017) assembly (Table 1). Despite initial differences

in size and contiguity, the assemblies were similar when considering only contigs containing at

least one of the 41 mitochondrial protein-coding genes conserved in *Gingko* and *Cycas*, which

we term high-confidence assemblies (Table 1). Finally, reconciliation using the Contig Integrator

for Sequence Assembly (CISA) pipeline (Lin & Liao 2013) produced a final draft assembly

measuring 4.9 Mb over four contigs with a mean depth of coverage of 284x.

Table 1. Summary statistics for assemblies of Pacific Biosciences Sequel® subreads enriched *in silico* for
mitogenome-like kmers. 'Raw' refers to the full contig output produced by each assembler. Upgraded
assemblies have been processed with FinisherSC. High-confidence assemblies contain only contigs with at
least one protein-coding mitochondrial gene. N50 and L50 are calculated from contig lengths.

Assembler	#Contigs	N50 (Mb)	L50	Longest contig (Mb)	Assembly size (Mb)
canu					
raw	383	0.06	18	1.10	10.25
upgraded	276	0.32	4	2.36	9.76
high conf.	4	1.28	2	2.56	5.29
MECAT					
raw	104	0.43	5	2.29	9.24
upgraded	95	0.43	5	2.29	9.22
high conf.	6	0.70	2	2.29	5.10
SMARTdenovo					
raw	59	0.76	2	3.60	7.31
upgraded	55	0.76	2	3.60	7.32
high-conf.	4	3.60	1	3.60	5.13
final draft	**4**	**3.42**	**1**	**3.42**	**4.90**

Genome annotation and gene repertoires in gymnosperms

General features of the *P. abies* mitogenome are summarized in Table 2. All 41 protein-

coding genes inferred to be present in the last common ancestor of angiosperms were also found

in *P. abies* and, after re-annotation, also in *P. glauca* and *Pinus taeda.* This pattern of

conservation in Pinaceae is consistent with the early-diverging gymnosperms *Gingko* and *Cycas*,

which may suggest a less dynamic gene repertoire than in angiosperms, although *Welwitschia*

has undergone extensive gene loss (Guo et al. 2016). In addition, the *P. abies* mitogenome has

also acquired complete copies of four plastid genes, *psaB*, *psbH*, *psbN*, and *psbT*, although the

functionality of these genes, if any, cannot be inferred from the data here. We also identified 20

transcribed open-reading frames: 14 had uniquely

mapping transcripts of ≥ 99% coverage and

identity (high confidence), two were of medium

confidence, where either identity or coverage was

< 99% but > 97%, and four low confidence genes

had less than 97% support (Table S2). Intron and

tRNA content were similar in *P. abies* and *P.*

glauca (Jackman et al. 2015), but both have

apparently undergone losses compared to the

other gymnosperms, including the Pinaceae

conifer *Pinus taeda*.

The repetitive fraction of the *P. abies*

mitogenome is larger than the entire mitogenome

of most plants (Table 2; Table S3). Relative to

genome size, however, repeat content in *P. abies*

Table 2. Characteristics of the *Picea abies* mitogenome.

Genome	
Size (Mb)	*ca.* 4.90
GC content	44.7%
Annotation	
Repeat content	15.15%
Direct and inverted	14.25%
Tandem	1.12%
Nuclear-mitochondrial DNA	28.89%
Transposable elements	7.72%
Plastid-derived DNA	0.34%
Genes	1.0%
Protein coding genes	41
Hypothetical proteins high/medium/low confidence	14/2/4
tRNAs	17
rRNAs	3

is unremarkable. Dispersed repeats in an analysis of 82 angiosperms comprised 14% of the

mitogenome on average (Dong et al. 2018), identical to the proportion in *P. abies* (Table 2).

However, dispersed repeats in *P. abies* tended to be about half the size of a typical tracheophyte,

at 312 bp *vs.* the 641 bp average (Wynn & Christensen 2019), indicating a relative enrichment of

smaller repeats (Fig. S1). At the same time, *P. abies* also had eight pairs of repeats ≥ 10 Kb,

more than all but two of the 79 land plants analyzed by Wynn & Christensen (2019).

Repeats and DNA transfers do not explain Picea *mitogenome expansion*

Extraordinary mitogenome size heterogeneity among plants has long been recognized

(Ward et al. 1981), but the sources appear to vary among species. We analyzed intergenomic

transfers and the proliferation of repetitive sequences as potential causes of genome size

variation in gymnosperms. Transfer of plastid DNA made a negligible contribution to the

genome size of the three Pinaceae conifers (Table 3), in contrast to some angiosperms, such as

cucurbits (Alverson, Rice, et al. 2011; Alverson et al. 2010). Similarly, no relationship between

genome size and repeat proliferation was evident: the 410 kb *Cycas* mitogenome was

proportionally the most repetitive (Chaw et al. 2008), and relative repeat content was similar in

Picea and *Ginkgo* despite their > 10x size differences (Table 3). An ambiguous correlation

between genome size and repeat proliferation was also observed in *Silene* species (Sloan et al.

2012), whereas larger cucurbit genomes tend to contain proportionally more repeats (Alverson et

al. 2010; Alverson, Rice, et al. 2011; Rodríguez-Moreno et al. 2011).

Table 3. Potential sources of mitogenome size variation among gymnosperms. Dispersed repeats include
those in the inverted and direct orientation ≥ 50 bp and with ≥ 80% identity. Nuclear-mitochondrial DNA
are shared sequences with no direction of transfer inferred. Transposable elements comprises long-
terminal repeats (LTR) and non-LTR retrotransposons. Numbers in parenthesis indicate percent coverage
of the mitogenome.

	Cycas	*Ginkgo*	*Picea abies*	*Picea glauca*	*Pinus taeda*	*Welwitschia*
Genome size (Mb)	0.41	0.35	4.90	5.20†	1.19	0.98
Plastid-derived DNA (kb)	18 (4)	0 (0)	17 (0)	18 (0)	3 (0)	9 (9)
Dispersed repeats (kb)	109 (26)	51 (15)	699 (14)	885 (17)	83 (7)	42 (4)
Nuclear-mtDNA (Mb)	-	0.35 (100)	1.40 (29)	2.29 (44)	0.60 (50)	-
Transposable elements (kb)	6 (1)	3 (1)	482 (10)	386 (7)	129 (10)	15 (2)

†Gene-containing scaffolds extracted from the putative 5.9 Mb assembly.

Interpreting shared nuclear-mitochondrial content is difficult because 1) the nuclear

genomes of *Cycas* and *Welwitschia* are not sequenced; 2) the direction of transfer can rarely be

determined with confidence under the simplest circumstances (Alverson et al. 2010); and 3) the

Cycas and *Ginkgo* mitogenomes are highly conserved, which implies that any import from the

nucleus predate their divergence *ca*. 354 My ago (Lu et al. 2014) and may no longer be

detectable in either nuclear genome (Guo et al. 2016). Therefore, it is unsurprising that the four

species with nuclear reference genomes show an equivocal relationship between mitogenome

size and shared nuclear-mitochondrial sequence (Table 3). Remnants of nuclear transposable

elements (TEs) can potentially act as markers of sequence import because they have an

unambiguous origin and generally do not proliferate after transfer (Knoop et al. 1996;

Goremykin et al. 2012). All three Pinaceae species showed an increase of TE remnants, but

relative content was similar among the two *Picea* species and the 4-fold smaller *P. taeda*

mitogenome (Table 3). Similarly, *Welwitschia* was relatively depauperate in TE elements,

despite its large mitogenome size (Table 3). Increased taxon sampling may help to clarify this

ambiguous relationship between TE import and genome size, and but a similarly unclear

relationship has also been reported among angiosperms (Alverson et al. 2010; Alverson, Rice, et

al. 2011; Rodríguez-Moreno et al. 2011, Goremykin et al. 2012). Overall, gymnosperms tend to

reinforce observations in angiosperms: repeat proliferation and DNA imports do not broadly

explain mitogenome size heterogeneity (Alverson, Rice, et al. 2011; Sloan et al. 2012; Dong et

al. 2018; Alverson et al. 2010).

Abundant repeat-mediated recombination at small repeats

Homologous intermolecular recombination is used to repair double-stranded breaks and

recover stalled replication forks (Maréchal & Brisson 2010). In many plants, intramolecular

recombination also occurs at dispersed

repeats and results in an individual harboring

alternative genome configurations (AGCs)

that differ in structure but are identical in

sequence. Intramolecular recombination

dynamics range from completely inert

(Alverson, Rice, et al. 2011; Dong et al.

2018) to astonishingly friable (Skippington

et al. 2015), yet recombination rates are not

widely estimated (see next section).

However, recombination resulting in DNA

exchange produces predictable AGCs that

are directly observable in a pool of single

molecule sequencing reads (e.g. Dong et al.

2018; Wang et al. 2018; Fig. S2). This

allowed us to identify and estimate the AGC

frequency associated with each pair of

inverted repeats in the *P. abies* mitogenome,

which provides a quantitative estimate of

recombination rates.

Genome rearrangements consistent

with the products of intramolecular

recombination (i.e., AGCs) comprised *ca*. 1%

of the pool of mapped reads. Half of the 598

repeats showed no evidence of recombination,

but AGC frequency averaged 2% at

Figure 2. Recombination frequency at inverted repeats ≥ 50 bp with ≥ 80% pairwise identity as inferred from long reads mapping to expected recombination products (alternative genome configurations; AGCs). A) Most repeat pairs have little or no evidence of recombination, but a minority are highly active. B) Repeat length explains very little of the variation in recombination frequency (r^2 = 0.03).

recombinogenic repeats (Fig. 2A; Table S4). Recombination was asymmetric in most cases, a

result consistent with the repair of stalled replication forks through break-induced replication

(Table S4; Maréchal & Brisson 2010). AGC abundance reached a maximum of 32% at a 186 bp

repeat, with the two possible isomers comprising 12% and 20% of the reads, respectively.

Surprisingly, the most recombinogenic repeats (AGCs ≥ 10%) ranged in size from 50 bp, the

minimum size evaluated, to 948 bp (Fig. 2A; Table S4). Overall, we found negligible correlation

between repeat length and AGC frequency (Fig. 2B; $r^2 = 0.03$, $p < 0.001$), in contrast to some

well-characterized examples (e.g. Mower, Floro, et al. 2012; Skippington et al. 2015; Guo et al.

2016) and contrary to the general expectation that, in the absence of other factors, recombination

should scale positively with repeat length (Arrieta-Montiel & Mackenzie 2011).

A reappraisal of intramolecular mitogenome recombination

258   Plant mitogenomes are widely cited to undergo frequent, reciprocal recombination at

large repeats and only rarely elsewhere based on the results of seminal pre-genomic studies

(Lonsdale et al. 1984; Palmer & Shields 1984; Arrieta-Montiel & Mackenzie 2011). As our

results suggested *P. abies* deviates from this pattern, we tested if this canonical model of

mitogenome evolution is still supported by modern sequencing data. We searched the *ca*. 200

published tracheophyte mitogenomes and identified those that 1) analyzed and reported repeat-

by-repeat recombination dynamics and 2) achieved at least a 60x average read depth of coverage.

This coverage allows identification of AGCs comprising ≥ 1.6% of the read pool and establishes

a baseline for comparing genomes sequenced to varying depths (e.g. 60 to over 1,000x), although

confirmed AGCs have been documented at much lower frequencies (Woloszynska 2009;

Arrieta-Montiel & Mackenzie 2011). Only 18 mitogenomes, representing ferns, gymnosperms,

and four major angiosperm lineages, met both criteria.

Table 4. Recombination patterns summarized from 18 published vascular plant mitogenomes. 'Proportion active' refers to the fraction of repeats producing alternative genome configurations (AGCs) inferred to be the product of recombination in frequencies ≥ 1.6% of the parent molecule. 'Max AGC' denotes the maximum frequency obtained by any AGC in the given repeat size class. Missing data because repeats of a size class do not exist in a given genome are listed as 'na', whereas '-' indicates missing data due to study limitations.

| Species | repeats ≥ 1,000 bp | | repeats < 1,000 bp | | Study |
	Proportion Active	Max AGC %	Proportion Active	Max AGC %	
Chrysanthemum nankingense	na	na	0.17	4	Wang et al. 2018
Cucumis sativus	1	50	0.08	5	Alverson, Rice, et al. 2011
Daucus carota	0.25	50	0.00	0	Iorizzo et al. 2012
Eucalyptus grandis	0.40	31	0.04	23	Pinard et al. 2019
Ginkgo biloba	1.00	50	0.00	0	Guo et al. 2016
Mimulus guttatus	1.00	50	0.38*	50	Mower, Floro, et al. 2012
Monsonia ciliata	-	-	0.00	0	Cole et al. 2018
Monsonia herrei	na	na	0.00	0	Cole et al. 2018
Nymphaea colorata	0.00	8	0.00	0	Dong et al. 2018
Ophioglossum californicum	0.50	25	0.00	0	Guo et al. 2017
Picea abies†	0.50	6	0.13	31	this study
Psilotum nudum	0.00	0	0.03	2	Guo et al. 2017
Silene conica	0.48	13	0.06	15	Sloan et al. 2012
Silene noctiflora	0.78	10	0.00	0	Sloan, Alverson, et al. 2012
Silene vulgaris	1.00	50	0.14	10	Sloan, Alverson, et al. 2012
Vigna angularis	1.00	33	0.26	24	Naito et al. 2013
Viscum scurruloideum	-	-	0.66	50	Skippington et al. 2015
Welwitschia mirabilis	na	na	0.00	0	Guo et al. 2016

† only inverted repeats analyzed
* minimum detection threshold is *ca.* 4%, thus this proportion is underestimated

Recombination patterns across tracheophytes provided mixed support for the canonical model of mitogenome evolution (Table 4, Table S5). Results are summarized as proportions in Table 4 to account for differences in repeat counts across species but are presented in absolute terms in Table S5. Consistent with the canonical model, large repeats tended to be more active than their smaller counterparts (Table 4). However, AGC abundances were not at equilibrium with their reciprocal configurations in most species (Table 4), indicating a lower recombination rate than anticipated from Southern blots (Maréchal & Brisson 2010). Although small repeats were overall less recombinogenic, active small repeats produced AGCs at similar frequencies as their larger counterparts in 33% of species, including *P. abies* (Table 4). All species with active

small repeats except for *Silene conica* showed high AGC frequencies at repeats measuring just

ca. 200 bp in length (Sloan et al. 2012; Mower, Floro, et al. 2012; Naito et al. 2013; Skippington

et al. 2015; Pinard et al. 2019). While references to recombination rates are often necessarily

vague, the 10-50% AGC abundances in these species exceed any reasonable interpretation of

phrases such as 'highly substoichiometric' that are frequently used to describe the activity of

small repeats (Woloszynska 2009; Arrieta-Montiel & Mackenzie 2011). Together, these studies

point to the importance of small repeats as drivers of mitogenome evolution, a long anticipated

(André et al. 1992) but rarely quantified phenomenon.

Differences in nuclear-encoded repair pathways may be the mechanism underlying

heterogeneity in recombination rates (Maréchal & Brisson 2010; Gualberto & Newton 2017).

For example, *Arabidopsis* mutants for mitochondrial repair and surveillance genes recombined

more readily at small repeats than their wild type counterparts (Zaegel et al. 2006; Shedge et al.

2007; Miller-Messmer et al. 2012; Wallet et al. 2015). While mutations associated with these or

undiscovered genes could contribute to the diversity of recombination dynamics, the results in

Table 4 do not show any discernable phylogenetic signal. If nuclear-encoded repair genes

directly explain most of the observed differences in recombination dynamics, then this lack of

signal implies repeated, independent evolution. Identifying factors influencing recombination

rates, and in turn how – or if – their variation explains facets of genome evolution such as

genome size, repeat content, and mutation rate would be aided by more consistent reporting of

recombination rates at a minimum.

Rampant mitogenome rearrangements in Picea

Rearrangements between the *P. abies* and *P. glauca* mitogenomes occurred an average of

every 1,540 bp and blocks of synteny rarely extended beyond gene boundaries (Fig. 3). After

accounting for the draft state of the genomes (Muñoz & Sankoff 2010), a parsimonious

rearrangement scenario (Tesler 2002) required 1,292 events to explain the size and distribution

of synteny blocks between the *Picea* species (1,310 events, unadjusted for assembly scaffold

number). Assuming a divergence time of approximately 15 My ago (95% CI: 10-18 My; Feng et

al. 2018) results in an absolute rearrangement rate of around 36 - 65 rearrangements My^{-1}. This

rate is similar to those observed in *Silene vulgaris* and some closely-related *Monsonia* species

(Cole et al. 2018), which appear exceptionally rearranged relative to other eukaryotes.

Rearrangement inference methods used here and by Cole et al. (2018) do not correct for events

that have been lost due to the erosion of shared sequence over time, which should underestimate

rearrangements given increasing divergence. When comparing similarly diverged mitogenomes

(*ca.* 50% shared sequence) to mitigate this bias, the high rearrangement rate in *Picea* is even

clearer: 43 *vs.* an average of 1 rearrangement My^{-1} (SD: 0.20) for four other species pairs (Table

S6).

A) Mitogenome alignment

B) *rps*19-*rps*3-*rpl*16-*nad*3-*rps*12 gene cluster

Figure 3. The mitogenomes of *Picea abies* and *P. glauca* are extensively rearranged and collinear regions
are limited to genic regions. An absolute rearrangement rate of 36-65 My^{-1} is needed to explain this level
of structural divergence. A) Simplified diagram of the *P. abies-P. glauca* mitogenome alignment, where
colored blocks represent corresponding homologous regions free of internal re-arrangements (locally
collinear blocks; LCBs) and heights are proportional to pairwise sequence identity. LCBs below the center
line in *P. glauca* are inverted with respect to *P. abies*. White space indicates regions with no homology.
Only LCBs longer than 2,000 bp are shown. B) Two gene clusters widely conserved in plant mitogenomes

are also found in *Picea* within a 9 kb block, which also serves to illustrate the typical extent of synteny
beyond genic regions. Gene structures are indicated by yellow boxes and introns by black lines.

Mitogenome shuffling in *Picea* contrasts with the remarkably static Pinaceae nuclear

genome, where a high degree of synteny has persisted among genera after ≥ 140 My of

divergence (Krutovsky et al. 2004; Pelgas et al. 2006; Pavy et al. 2012). Even the nuclear

genomes of Cupressaceae and Pinaceae, which last shared a common ancestor in the

Carboniferous (Leslie et al. 2018), are more collinear than the *P. abies* and *P. glauca*

mitogenomes (Ehrenmann et al. 2015). Despite their extensive rearrangements, two of the three

gene clusters widely preserved in tracheophytes from their α-proteobacterial endosymbiont

(Dong et al. 2018) were also maintained in *Picea* within the same 9 kb block: *rpl*2-*rps*19-*rps*3-

*rpl*16 and *nad*3-*rps*12 (Fig. 3B). The third widely-conserved gene cluster – *rrn*5-*rrn*18 – has not

been preserved in *Picea*, *Pinus taeda*, or *Welwitschia* but persisted in *Cycas* and *Ginkgo* (Guo et

al. 2016), suggesting it was lost in the common ancestor of conifers and gnetophytes.

Intramolecular recombination creates AGCs that can potentially be transmitted as

rearrangements (Gualberto & Newton 2017; Cole et al. 2018). However, the path from isomer to

rearrangement is not straightforward because mitogenomes are subject to genetic drift within an

individual and within a population (Gualberto & Newton 2017). After an AGC is produced, it

must survive multiple rounds of cell division, be recruited into the germline, and persist through

potentially multiple generations before the rearrangement becomes firmly established within an

individual (Davila et al. 2011). Proliferation throughout the population could then occur as other

polymorphisms, probably predominately through stochastic demographic forces but possibly also

through natural selection (Shedge et al. 2010). Several aspects of this process are unknown,

including how mitogenomes are replicated (Gualberto & Newton 2017), the timing of germline

segregation (Lanfear 2018), and when, where, and under what circumstances AGCs arise. For

these reasons, the relationship between recombination within individuals summarized in Table 4

and re-arrangement rates among species is unclear. For example, *Monsonia* and *Silene vulgaris*

have markedly different levels of intramolecular recombination (Table 4), yet have similar

rearrangement rates (Cole et al. 2018).

Sequence divergence in Picea *and among gymnosperms*

A strong dichotomy between rates of sequence and structural evolution is a well-known feature of plant mitogenomes (Palmer & Herbon 1988) and is also the dynamic found in *Picea* and other Pinaceae conifers (Fig. 4). Substitution rates across all protein-coding genes averaged 0.0023 for synonymous and 0.0012 for nonsynonymous sites. These rates are about 1/2 of those in the plastid genome (Sullivan et al. 2017) and 1/4 of the nuclear genome (Buschiazzo et al. 2012;

Figure 4. Mean substitution rates at synonymous (dS) and nonsynonymous (dN) sites vary 100-fold among gymnosperms. Rates within Pinaceae, in gray, are more consistent but are about 6-fold higher in *Picea* than in the 75% smaller *Pinus* mitogenomes.

Chen et al. 2012). Assuming a divergence time of 10 - 17 My between *P. abies* and *P. glauca* (Feng et al. 2018), mitochondrial substitution rates in *Picea* fall within the lower end of absolute rates observed in angiosperms ($dS = 7.65 \times 10^{-11}$; $dN = 4.00 \times 10^{-11}$ assuming a 15 My divergence, cf. Mower et al. 2007) and are slightly higher than in *Cycas* and *Ginkgo* (Guo et al. 2016). Absolute substitution rates in *Pinus*, however, include the lowest rates reported so far in vascular plants (Richardson et al. 2013). Using divergence times from Saladin et al. (2017), rates ranged from $dS = 0.60 \times 10^{-11}$ and $dN = 0.30 \times 10^{-11}$ site/year in *Pinus taeda* to $dS = 3.15 \times 10^{-11}$ and $dN = 0.17 \times 10^{-11}$ in *Pinus strobus*. Previous work with more taxa, but fewer loci, similarly inferred exceptionally low substitution rates in some *Pinus* species (Wang & Wang 2014). At the other extreme, absolute dS measured *ca*. 30.0×10^{-11} site/year in *Welwitschia* and *Araucaria heterophylla* when assuming a divergence from Pinaceae around 342 My ago (Lu et al. 2014). Relative substitution rates were consistent with previous studies analyzing fewer genes or species (e.g. Mower et al. 2007; Guo et al. 2016) and are reported in Table S7.

Intraspecific variation

We used the PacBio Sequel® system to sequence partial mitogenomes from two *P. abies* on the alpine tundra in central Sweden, about 400 km southwest from the reference tree. *Picea* megafossil remains in this region date to *ca*. 11,000 ^{14}C yr^{-1}, which suggests the presence of high-latitude glacial refugia in Scandinavia (Kullman 2008). Both formed small clonal groups above the modern tree line and are spatially associated with megafossils dated to 5,120 and 4,820 ^{14}C yr^{-1}, respectively, raising the possibility that these clonal groups are extremely ancient (Kullman 2001). Thus, the mitogenomes of these two trees is relevant to post-glacial recolonization history and to the study of mitochondrial repair and recombination dynamics in long-lived organisms. In total, we recovered 499,180 and 553,660 bp, respectively, of which 82,443 was shared between the two trees. Nine variant sites were shared by the two trees relative to the reference tree. In contrast, 58 structural variations were found between these putatively ancient trees and the reference individual: 29 insertions, 17 deletions, 13 translocations, nine duplications, and seven inversions. More data is needed to address ecological and molecular hypotheses, but these sequences support a high rate of structural evolution in *Picea*.

Conclusion

Plant mitogenomes are highly variable in size, repeat and gene content, and mutation and rearrangement rates. The underlying processes generating this heterogeneity are largely unclear: for each mitogenome supporting a given mechanistic hypothesis, another often suggests the opposite, such as in the relationship between mutation rate and genome size (cf. Sloan et al. 2012; Skippington et al. 2015, Christensen 2018). The *P. abies* mitogenome and our comparative analyses may lend support to the view of plant mitogenomes as highly idiosyncratic and driven mainly by rapid evolution and/or considerable genetic drift. As in previous studies, we found no clear relationship between genome size, repeat content, intergenomic transfer, or substitution rate. However, we identified recombination as an underinvestigated mechanism of plant mitogenome evolution, despite being recognized as a likely source of the differences among eukaryotes (Palmer & Herbon 1988; Gray et al. 1999). Recombination rates vary extensively among plants and small repeats (< 1,000 bp) are highly active in a third of the reported species. Recombination affects the accumulation of mutations (Maréchal & Brisson 2010; Christensen 2018), influences genome size (Christensen 2018), and induces structural rearrangements

(Palmer & Herbon 1988), making this variation a potential but understudied contributor to the

diversity of plant mitogenomes.

Methods

Machine learning classification of genomic scaffolds

The support vector machine involved four steps: 1) identification of high-quality training

data, 2) training the classifier using this data and pre-determined genomic features, 3) evaluation

of the model performance, and 4) applying the classifier to the *P. abies* v. 1.0 genome assembly.

To curate a reliable set of positive and negative training data from *P. abies*, we first discarded

scaffolds < 500 bp and masked repetitive elements using RepeatMasker v. open-4.0.1 (Smit et al.

2013). We aligned the remaining scaffolds to a set of relatively well-annotated genomes,

including *Arabidopsis thaliana* and *Populus trichocarpa* and 18 additional organelle genomes

(Table S8), using blastn and tblastx in BLAST version 2.2.29+ with default parameters except

for an e-value cutoff of 1e-20. Known numts (nuclear mitochondrial DNA sequences) were

masked from the *Arabidopsis* assemblies. *P. abies* sequences were considered robustly assigned

if they matched an annotated gene region on the subject genome and 1) the top bitscore was >

100; 2) the quotient between the top bitscore and next-best hit was > 1.2; 3) alignments spanned

at least 150 bp for blastn and 100 aa for tblastx; 4) blastn and tblastx annotations were identical if

both existed; and 5) at least two different reference species contained the BLAST hit. Scaffolds

meeting all these criteria were then classified as mitochondrial or non-mitochondrial and as

training data for the SVM classifier. This yielded 2.0 Mb and 51.1 Mb of positive and negative

data, respectively, for training the SVM.

After identifying robust positive and negative training datasets, we trained the SVM

classifier to distinguish between them using pre-determined features. Following previous studies

(Nystedt et al. 2013; Jackman et al. 2015), we used the standardized values of the natural

logarithm of mean depth of coverage, the proportion of ambiguous nucleotides (%N), and the

proportion of guanine and cytosine (%GC) for each scaffold, and a kmer-based score based on

the frequency of nonredundant 'mitochondrial-like' or 'non-mitochondrial-like' oligonucleotide

sequences in a scaffold. We calculated probability tables for occurring kmers based on positive

and negative training sequences and calculated a score for each accordingly:

448
$$s_c = \prod_{k=1}^{n} p_{c,kmer_k}$$

449   where c is the class, n is the number of kmers in the scaffold, and $p_{c,kmer_k}$ is the probability of

450   kmer k occurring in a sequence of class c. The final kmer classifier score for each scaffold was

451   calculated as:

452
$$s = \frac{\log s_{pos} - \log s_{neg}}{L}$$

where L is the scaffold length and s_{pos} and s_{neg} are the positive and negative scores as defined

above. Thus, smaller scores are consistent with plastid or nuclear sequence whereas larger scores

indicate a scaffold comprising more mitochondrial-like kmers. Optimal kmer size for the

classifier ($k = 7$, here) was assessed using test and training data from *A. thaliana* and visual

inspection of the receiver operating characteristic (ROC) curves. After defining these features,

we used SVMlight (Joachims 1999) with a Gaussian kernel to conduct the training step and to

apply the learned model to the set of *P. abies* scaffolds with length > 500 bp and coverage >

100x.

We assessed the SVM model performance using cross-validation. The initial training data

were divided randomly into nine trials varying in training subsets from 10-90%, each consisting

of 100 cross-validation tests. Kmer scores, SVM training, and classification were carried out as

described above. For each trial, we calculated the mean false discovery rate (FDR) and true

positive recall.

Mitogenome assembly

We screened subreads from 77 PacBio Sequel® SMRT cells generated from genomic

DNA (Street et al., unpublished data) for 27-mer matches to any of the SVM-classified scaffolds

using BBDuk v. 35.14 (http://jgi.doe.gov/data-and-tools/bb-tools). Enriched reads were

assembled using canu v. 1.7 (Koren et al. 2017), MECAT v. 1.3 (Xiao et al. 2017), and

SMARTdenovo (https://github.com/ruanjue/smartdenovo). Canu was run using default

parameters, except `corMaxEvidenceErate` was set to 0.15 as recommended in the manual

for repetitive and GC-skewed genomes. MECAT was also run using default parameters using a

target of 45x, 55x, and 65x coverage of the longest corrected reads. For MECAT and canu, we

specified a genome size of approximately 5.0 Mb, which is used to estimate the coverage of the

input reads. Because SMARTdenovo does not include a preassembly read correction step, we

used the 45x, 55x, and 65x corrected reads from MECAT as input and the default parameters

were then used for assembly.

We selected the most contiguous assembly from each assembler for further refinement.

We passed these assemblies and their constituent reads to FinisherSC to reconstruct the overlap

graph and identify contigs that can be robustly merged (Lam et al. 2015). Next, we used tblastx

to identify the contig in each upgraded assembly containing the best match to each of the 41

protein-coding genes in the *Cycas* mitogenome (Chaw et al. 2008), which we retained as 'high-

confidence' assemblies. Then, we used pairwise alignments from NUCmer (Kurtz et al. 2004) to

break the three assemblies at major disagreements. The resulting contigs were put through the

assembly reconciliation pipeline implemented in CISA v. 1.3 (Lin & Liao 2013) to re-assemble

them into a single draft. After checking for circular contigs using dot plots, we evaluated the

quality of the assembly by aligning the corrected reads using minimap2 (Li 2018) to the draft and

visually inspected their congruency in IGV v. 2.4.14. (Thorvaldsdóttir et al. 2012). Finally, we

used the partial order alignment graph approach implemented in Racon to re-call the consensus

sequence from the minimap2 alignment (Vaser et al. 2017).

Genome annotation

We reused repeat libraries curated for the *P. abies* v. 1.0 assembly (Nystedt et al. 2013)

as input for RepeatMasker v. 4.0.7. (Smit et al. 2013) to identify TEs including long

retrotransposons (LTRs), transposable elements, and other interspersed elements (LINEs and

SINES). We additionally used RepeatModeler v. 1.0.8. to identify potential *de novo* repeats with

significant similarity to the RepBase and Dfam databases (Smit et al. 2013). Direct and inverted

repeats were identified with self *vs.* self blastn searches following Guo et al. (2016).

Mitochondrial DNA of plastid origin (MIPTs) were identified following the methods of Guo et

al. (2016) and shared mitochondrial-nuclear DNA following the methods of (Alverson, Rice, et

al. 2011).

We used MAKER v. 3.01.2 to annotate protein-coding genes (Cantarel et al. 2008).

Complex repeats identified by RepeatMasker and RepeatModeler were hard-masked prior to

annotation, whereas simple repeats were re-annotated and soft masked by MAKER. As empirical

gene evidence, we used the transcriptomes of *P. abies*, *P. glauca*, *P. sitchensis, Pinus pinaster*,

Pinus sylvestris, and *Pinus taeda* from the PLAZA 3.0 database (Proost et al. 2014) and all

curated plant sequences (Swiss-Prot) from UniProt as protein evidence. This provided a total of

312,953 ESTs and 37,954 proteins for use as evidence alignments. Although we initially

attempted to use the SNAP and AUGUSTUS *ab initio* gene predictors, we found the number of

high-confidence genes was insufficient for training. Therefore, annotations are based on

empirical evidence alone. Coding sequences were manually curated to improve identification of

reading-frames and gene structure based on sequence homology. Hypothetical proteins were

compared to an RNA-Seq dataset obtained from needles (PRJEB26398, Schneider et al., in

preparation), in which transcripts were extracted using a ribominus kit and subjected to *de novo*

transcriptome assembly using Trinity v. 2.8.4 using default parameters (Grabherr et al. 2011).

The *de novo* reconstructed transcripts were aligned to the *P. abies* genome v. 1.0 retrieved from

ConGenIE.org (Sundell et al. 2015) using the GMAP v. 2015.11.20 (Wu & Watanabe 2005). The

alignment coordinates were intersected with the ab-initio prediction to validate the predicted

hypothetical proteins using an ad-hoc R script.

Identification of repeat-mediated recombination and rearrangements

For each inverted repeat pair, we extracted the ± 2,000 bp single-copy flanking regions

and constructed their expected recombination products for use as an alternative genome

reference (Day & Madesis 2007, Fig. S2). We used minimap2 to map the enriched PacBio reads

against these four sequences, with the expectation that reads spanning the repeat and flanking

regions of the alternative configurations with robust mapping quality (MQ > 50) represent actual

structural variations present within the sequenced individual tree (e.g. Park et al. 2014;

Skippington et al. 2015; Cole et al. 2018; Dong et al. 2018).

We aligned the mitogenomes of *P. abies* and *P. glauca* using the rearrangement-aware

progressiveMAUVE aligner (Darling et al. 2010), which identifies locally-collinear blocks

(LCBs) internally free of recombination and reorders them against a selected genome. Then, we

used GRIMM 2.01 (Tesler 2002) to infer a minimum, optimum rearrangement history, assuming

signed undirected chromosomes. Apparent rearrangements introduced by the draft state of the

genomes were corrected following Muñoz & Sankoff (2010) and Cole et al. (2018). Shared

sequence between *P. abies* and *P. glauca* was estimated using blastn and the same parameters as

Guo et al. (2016) for consistency. To put the re-arrangement rates inferred in *Picea* into a

broader phylogenetic context, we re-aligned and estimated minimum numbers of rearrangements

from the species pairs in Guo et al. (2016) with similar levels of divergence (Table S6).

We searched manuscripts associated with all 127 tracheophyte mitogenomes in the NCBI

Genome resource (accessed March 11, 2019) for estimations of recombination rates. In addition,

we exhaustively searched Google Scholar for 'plant mitochondrial genomes' for publications

since 2011, the year of the last extensive review of plant mitogenome recombination (Alverson,

Zhuo, et al. 2011). To be included in our analysis, we required 1) genomes to be assembled using

high-throughput sequencing at a depth of coverage sufficient to allow detection of AGCs

comprising $\geq$ 1.6% of the read pool, 2) estimates of repeat-by-repeat recombination rates inferred

from mapping statistics or alignment rates to be clearly reported, either tabulated or in figures

sufficiently detailed to allow precise interpolation of from vectorized figures using Inkscape v.

0.91 (e.g. Sloan et al. 2012, their Fig. S6). Wherever possible, we limited the analyses to repeats

$\geq$ 50 bp with $\geq$ 80% identity, although some studies employed more stringent cut-offs. Relevant

details for each genome, including the sequencing depth of coverage, repeat number, range of

repeat sizes and identities, and other analytical details are summarized in Table S5.

Comparative analysis of gymnosperm mitogenomes

We downloaded the mitogenomes of *Amborella trichopoda, Cycas taitungensis*, *Ginkgo*

biloba, *Welwitschia mirabilis*, *P. glauca*, and *Pinus taeda* from Genbank to compare substitution

rates, gene and repeat content, and genomic structure. For analyses of nucleotide substitution

rates, we also used the previously-sequenced mitogenome coding sequences of *Gnetum gnemon*.

Pinus sylvestris, and *Araucaria heterophylla* from NCBI Genbank. Given the draft state of the *P.*

glauca mitogenome, we retained only contigs with conserved mitochondrial genes, which

resulted in a 5.2 Mb assembly. We re-annotated repeats and intergenome sequence transfers as

described for *P. abies* for consistency.

Nucleotide substitution rates were estimated from the 41 protein-coding genes in

Pinaceae. Coding sequences were extracted, aligned with MUSCLE v. 3.8.42 (Edgar 2004), and

manually verified to ensure correct reading frames and complete codons. RNA editing sites were

predicted using the PREP-mt webserver (Mower 2005) using a cutoff value of 0.2 as in Guo et

al. (2016). Edited nucleotide sequences were re-aligned using the translation alignment option in

Geneious v. 11.1.4 and poorly aligned blocks of codons were removed using Gblocks v. 0.91b

webserver (Castresana 2000). First and second codon positions were extracted using DnaSP v. 6

(Rozas et al. 2017) and the maximum likelihood phylogeny of the concatenated alignment was

inferred using RAxML v. 8.2.12 (Stamatakis 2014) under the GTR-GAMMA model of

nucleotide evolution. This phylogeny agreed with the prevailing consensus of gymnosperm

evolution (Lu et al. 2014), so we estimated estimate branch lengths in units of synonymous (dS)

and nonsynonymous (dN) substitution rates under the free-ratio branch model in PAML on this

constrained topology (Yang 2007).

Intraspecific mitogenome variation in Picea abies

We selected two putatively ancient trees from Kullman (2008) and collected newly-

flushed buds. Buds were immediately placed in a solution comprising 10 mM MOPS, 5% (v/v)

dimethylsulfoxide and 5% (w/v) glycerol at pH 7.3 and stored on dry ice and then at -80°C until

isolation. In brief, intact mitochondria were isolated using centrifugation and extraneous DNA

was degraded using DNase, and then mitochondrial DNA was isolated using a gram-negative

bacteria genomic DNA purification kit (Supplementary Methods 1). DNA library construction

and PacBio RS II sequencing was performed at DUKE Center for Genomic and Computational

Biology according to manufacturer's instructions.

PacBio sub-reads were checked for adapter contamination and then aligned to the whole

P. abies v. 1.0 assembly containing the new *de novo* mitogenome using minimap2 (Li 2018).

Scaffolds identified as mitochondrial-like by the SVM in the *P. abies* v. 1.0 assembly but were

not contained within the *de novo* assembly were left in place, as they may represent numts. Bases

were called using GATK v. 3.8.0 Haplotype Caller in gvcf mode followed by joint-genotyping in

`includeNonVariantSites` mode (Van der Auwera et al. 2013). As GATK does not assign

confidence metrics to non-variant sites, we considered those covered by 10 to 35 reads to be

represented. Sites with higher coverage represent unresolved repeats. Structural variants were

called with Sniffles v. 1.0.11 using default filtering settings (Sedlazeck et al. 2018). Each

structural variant was then visually validated in IGV v. 2.4.14. (Thorvaldsdóttir et al. 2012).

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