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MitoFinder: Efficient automated large-scale extraction of mitogenomic data in target enrichment phylogenomics

Rémi Allio1 | Alex Schomaker-Bastos2* | Jonathan Romiguier1* | Francisco Prosdocimi2 | Benoit Nabholz1 | Frédéric Delsuc1

Abstract

Thanks to the development of high-throughput sequencing technologies, target enrichment sequencing of nuclear ultraconserved DNA elements (UCEs) now allows routine inference of phylogenetic relationships from thousands of genomic markers. Recently, it has been shown that mitochondrial DNA (mtDNA) is frequently sequenced alongside the targeted loci in such capture experiments. Despite its broad evolutionary interest, mtDNA is rarely assembled and used in conjunction with nuclear markers in capture-based studies. Here, we developed MitoFinder, a user-friendly bioinformatic pipeline, to efficiently assemble and annotate mitogenomic data from hundreds of UCE libraries. As a case study, we used ants (Formicidae) for which 501 UCE libraries have been sequenced whereas only 29 mitogenomes are available. We compared the efficiency of four different assemblers (IDBA-UD, MEGAHIT, MetaSPAdes, and Trinity) for assembling both UCE and mtDNA loci. Using MitoFinder, we show that metagenomic assemblers, in particular MetaSPAdes, are well suited to assemble both UCEs and mtDNA. Mitogenomic signal was successfully extracted from all 501 UCE libraries, allowing us to confirm species identification using CO1 barcoding. Moreover, our automated procedure retrieved 296 cases in which the mitochondrial genome was assembled in a single contig, thus increasing the number of available ant mitogenomes by an order of magnitude. By utilizing the power of metagenomic assemblers, MitoFinder provides an efficient tool to extract complementary mitogenomic data from UCE libraries, allowing testing for potential mitonuclear discordance. Our approach is potentially applicable to other sequence capture methods, transcriptomic data and whole genome shotgun sequencing in diverse taxa. The MitoFinder software is available from GitHub (https://github.com/RemiAllio/MitoFinder).

KEYWORDS
bioinformatics, DNA barcoding, insects, invertebrates, metagenomics, systematics
Next-generation phylogenomics, in which phylogenetic relationships are inferred from thousands of genomic markers gathered through high-throughput sequencing (HTS), is on the rise. More specifically, targeted enrichment or DNA sequence capture methods are becoming the gold standard in phylogenetic analyses because they allow subsampling the genome efficiently at reduced cost (Lemmon & Lemmon, 2013; McCormack et al., 2013). The field has witnessed the rapid parallel development of exon capture from transcriptome-derived baits (Bi et al., 2012), anchored hybrid enrichment techniques (Lemmon, Emme, & Lemmon, 2012), and the capture of ultraconserved DNA elements (UCEs; Faircloth et al., 2012). All hybridization capture methods target a particular portion of the genome corresponding to the defined probes plus flanking regions. Previous knowledge is required to generate sequence capture probes, but ethanol-preserved tissues, old DNA extractions and museum specimens can be successfully sequenced (Blaimer et al., 2015; Faircloth et al., 2012; Guschanski et al., 2013). The first UCEs were identified by Bejerano et al. (2004) in the human genome and have been shown to be conserved in mammals, birds and even ray-finned fish (Stephen, Pheasant, Makunin, & Mattick, 2008). Thanks to their large-scale sequence conservation, UCEs are particularly well suited for sequence capture experiments and have become popular for phylogenomic reconstruction of diverse animal groups (Blaimer et al., 2015; Esselstyn, Olveros, Swanson, & Faircloth, 2017; Guschanski et al., 2013). Initially restricted to a few vertebrate groups such as mammals (McCormack et al., 2012) and birds (McCormack et al., 2013), new UCE probe sets have been designed to target thousands of loci in arthropods such as hymenopterans (Blaimer et al., 2015; Branstetter, Danforth, et al., 2017; Faircloth, Branstetter, White, & Brady, 2015), coleopterans (Baca, Alexander, Gustafson, & Short, 2017; Faircloth, 2017) and arachnids (Starrett et al., 2017).

It has been shown that complete mitochondrial genomes could be retrieved as by-products of sequence capture/enrichment experiments such as whole exome capture in humans (Picardi & Pesole, 2012). Indeed, mitogenomes can in most cases be assembled from off-target sequences of UCE capture libraries in aninotes (do Amaral et al., 2015). Despite its well-acknowledged limitations (Galtier, Nabholz, Glënóm, & Hurst, 2009), mitochondrial DNA (mtDNA) remains a marker of choice for phylogenetic inference (e.g., Hassanin et al., 2012), for species identification or delimitation through barcoding (e.g., Coissac, Hollingsworth, Lavergne, & Taberlet, 2016), and to reveal potential cases of mitochondrial discordance resulting from introgression and/or hybridization events (e.g., Grummer, Morando, Avila, Sites, & Leaché, 2018; Zarza et al., 2016, 2018). mtDNA could also be used to taxonomically validate the specimens sequenced for UCEs using CO1 barcoding (Ratnasingham & Hebert, 2007) and to control for potential cross-contaminations in HTS experiments (Ballenghien, Faivre, & Galtier, 2017). In practice, the few studies that have extracted mtDNA signal from UCEs (e.g., Meiklejohn et al., 2014; Pie et al., 2017; Wang, Hosner, Liang, Braun, & Kimball, 2017; Zarza et al., 2018) have done so manually for only a few taxa. Most studies assembling mitogenomes from UCE libraries have used contigs produced by the Trinity RNAseq assembler (Grabherr et al., 2011) as part of the PHYLUCE pipeline (Faircloth, 2016), which was specifically designed to extract UCE loci. Indeed, RNAseq assemblers such as Trinity allow to deal with the uneven coverage of target reads in sequence-capture libraries, but also multicopy genes such as the ribosomal RNA cluster, and organelles (chloroplasts and mitochondria). However, this strategy probably does not scale well to hundreds of taxa because of the high computational demand required by Trinity. A potential solution to extract mitochondrial signal from UCE libraries could be the use of iterative mapping against a reference mitogenome using MiTOBIN (Hahn, Bachmann, & Chevreux, 2013). However, this tool requires both closely related reference mitogenomes and good coverage to perform well and also requires UCE and mtDNA assemblies to be conducted separately. Metagenomic assemblers could provide a powerful alternative to assemble both UCE loci and mtDNA simultaneously because they have been designed for efficient de novo assembly of complex read populations by explicitly dealing with uneven read coverage and are computationally and memory efficient. Comparisons based on empirical bulk data sets of known composition (Vollmers, Wiegand, & Kaster, 2017) have identified IDBA-UD (Peng, Leung, Yu, & Chin, 2012), MEGAHIT (Li et al., 2016) and MetaSPades (Nurk, Meleshko, Korobeynikov, & Pevzner, 2017) as the most efficient current metagenomic assemblers.

As a case study, we focused on ants (Hymenoptera: Formicidae) for which only 29 mitogenomes were available on GenBank compared to 501 UCE captured libraries as of March 29, 2018 (Appendix S1). This contrasts sharply with the other most speciose group of social insects, termites (Isoptera), for which almost 500 reference mitogenomes have been produced (Bourguignon et al., 2017) and no UCE study has been conducted so far. Sequencing and assembling difficulties stemming from both the AT-rich composition (Foster, Jermiin, & Hickey, 1997) and a high rate of mitochondrial genome rearrangements in hymenopterans (Dowton, Castro, & Austin, 2002) might explain the limited number of mitogenomes currently available for ants. It is only recently that a few ant mitogenomes have been assembled from UCE data (Mecalázaro, Poteaux, Bayona-Vásquez, Branstetter, & Zaldívar-Riverón, 2018; Ströhler et al., 2017; Vieira & Prosdocimi, 2019). Here, we built a pipeline called MitoFinder designed to automatically assemble both UCE and mtDNA from raw UCE capture libraries and to specifically extract and annotate mitogenomic contigs. Using publicly available UCE libraries for 501 ants, we show that complementary mitochondrial phylogenetic signal can be efficiently extracted using metagenome assemblers along with targeted UCE loci.

2 | MATERIALS AND METHODS

2.1 | Data acquisition

We used UCE raw sequencing data for 501 ants produced in 10 phylogenomic studies (Blaimer et al., 2015; Blaimer et al.,...
1. Cell lysis & DNA extraction

2. Target enrichment with probes

3. Sequencing & assembly

4. Find and annotate mitochondrial contigs with MitoFinder

5. Barcode identification

6. UCE annotation

7. Create mtDNA and UCE matrices

8. Phylogeny inferences

**FIGURE 1** Conceptualization of the pipeline used to assemble and extract UCE and mitochondrial signal from ultraconserved element sequencing data.
2.2 | De novo assembly of mitogenomic and UCE data with MitoFinder

To extract mitogenomic data from UCE libraries, we developed a dedicated bioinformatic pipeline called MitoFinder (Figure 1). This pipeline was designed to assemble sequencing reads from target enrichment libraries, extract, and annotate mitochondrial contigs. To evaluate the impact of assembler choice, contigs were assembled with idba-ud version 1.1.1, megahit version 1.1.3, and metaSPAdes version 3.13.0 within MitoFinder, and with Trinity version 2.1.1 within phyUce using default parameters. Mitochondrial contigs were then identified by similarity search using BLAST with e-value ≥ 1e−06 against our ant reference mitogenomic database. Each detected mitochondrial contig was then annotated with tblastx for protein-coding genes (CDSs) and blastn for 16S and 12S rRNAs taking advantage of the geneChecker module of MITObim (Schomaker-Bastos & Prosdocimi, 2018) that we incorporated into MitoFinder. Finally, we used ArwEn version 1.2 (Laslett & Canback, 2007) to detect and annotate tRNA genes.

Considering possible rearrangements in ant mitogenomes, each annotated mitochondrial CDS was first aligned with Mafft version 7.271 (Katoh & Standley, 2013) algorithm FFT-NS-2 with option --adjustdirection. Then, to take into account potential frameshifts and stop codons, mitochondrial CDS alignments were refined with macse version 2.03 (Ranwez, Douzery, Cambon, Chantret, & Deluc, 2018) with option -prog alignSequences, which produces both nucleotide and amino acid alignments. To improve alignment accuracy and reduce calculation time, we used sequences from available ant mitogenomes as references for each CDS (option -seg_lr). Sequences with internal stop codons were excluded to remove incorrectly annotated fragments potentially corresponding to nuclear mtDNA segments (NUMTs) in each protein-coding gene alignment. Then, individual gene alignments were checked by eye to manually remove remaining aberrant sequences. Finally, a nucleotide supermatrix was created by concatenating protein-coding and ribosomal RNA genes. Because the mitochondrial signal might be saturated for inferring deep phylogenetic relationships, an amino acid supermatrix with the 13 mitochondrial CDSs was also assembled.

2.3 | Guided iterative mitogenomic data assembly with MITObim

For comparison purposes, we also ran MITObim (Hahn et al., 2013) to extract mitochondrial sequences from the 501 UCE raw sequencing data. This software is designed to assemble mitochondrial reads using mitochondrial bait such as the CO1 sequence of a related species when available. Then, based on iterative mapping, MITObim extends as much as possible the mitochondrial contig previously obtained. For each library, given the scarcity of closely related complete mitochondrial genomes available for ants, the longest CO1 sequence available for the genus, or the most closely related genus, was used as bait for the initial step of MITObim. As there is no annotation step in MITObim, MitoFinder was used to annotate the resulting MITObim contigs.

2.4 | DNA barcoding

To verify species identification of the 501 ant UCE libraries, CO1 sequences extracted by MitoFinder using metaSPAdes (mtDNA recovered for all species) were compared with species-level barcode records (3,328,881 CO1 sequences including more than 100,000 ants) through the identification server of the Barcode Of Life Data System version 4 (Ratnasingham & Hebert, 2007). The same CO1 sequences were also compared against the NCBI nucleotide database using megablast with default parameters. An identification was considered to be confirmed when the query CO1 sequence had 95% similarity with a reference sequence in BOLD or GenBank with the same identifier.

2.5 | Assembly of UCEs

As recommended by Faircloth (2016), we first relied on Trinity to assemble UCE contigs using the phylice_ assembly_ assembly_trinity module of phyUce. To assess the impact of assembler choice on UCE loci retrieval, we also used the assemblies obtained with idba-ud, megahit, and metaSPAdes as implemented in MitoFinder. PHYLUCE scripts phylice_ assembly_ get_ match_ counts and phylice_ assembly_ get_ fastas_ from_ match_ counts were used to match contigs obtained for each sample to the bait set targeting 2,590 UCE loci for Hymenoptera (Branstetter, Ješovnik, et al., 2017). The resulting alignments were then cleaned using gblocks (Castrasana, 2000) with the phyucle_ align_ get_ gblocks_ trimmed_ alignments_ from_ untrimmed script. Finally, loci found in at least 75% of species were selected to create the four corresponding UCE supermatrices using the phyucle_ align_ get_ only_ loci_ with_ min_ taxa script.

2.6 | Phylogenetic analyses

Phylogenetic relationships of ants were inferred from a total of 16 different supermatrices corresponding to the four supermatrices
constructed from contigs obtained with each of the four assemblers (IDBA-UD, MEGAHIT, MetaSPAdes, and Trinity). The four supermatrices are as follows: (i) a UCE nucleotide supermatrix built from the concatenation of UCE loci retrieved for at least 75% of species, (ii) a mitochondrial nucleotide supermatrix consisting of the concatenation of the 13 protein-coding genes and the two rRNA genes, (iii) a mitochondrial amino acid supermatrix of the 13 protein-coding genes, and (iv) a mixed supermatrix of UCE nucleotides and mitochondrial amino-acid protein-coding genes. For all supermatrices, phylogenetic inference was performed with maximum likelihood (ML) as implemented in IQ-TREE version 1.6.8 (Nguyen, Schmidt, von Haeseler, & Minh, 2015) using a GTR+Γ model for UCE and mitochondrial nucleotide supermatrices, an mtART+Γ model partitioned by gene for mitochondrial amino acids, and a partitioned model mixing a GTR+Γ model for UCE nucleotides and an mtART+Γ model for mitochondrial amino acids for the mixed supermatrices. Statistical node support was estimated using ultrafast bootstrap (UFBS) with 1,000 replicates (Hoang, Chernomor, von Haeseler, & Minh, 2018). Nodes with UFBS values >95% were considered strongly supported. For all supermatrices, the congruence among the different topologies obtained with the four assemblers was evaluated by calculating quartet distances with DQUAD (Ranwez, Criscuolo, & Douzery, 2010).

3 | RESULTS

3.1 | Assembly of UCE data sets

De novo assembly of 501 UCE capture sequencing libraries was performed with four different assemblers: IDBA-UD, MEGAHIT and MetaSPAdes via MitoFinder and Trinity via PHYLUCE. All assemblers provided different numbers of contigs (Table 1) ranging from 30,544 (IDBA-UD) to 114,392 (MEGAHIT) on average. The average computational time per assembly was highly variable among assemblers with Trinity being by far the slowest (35 CPUs, median time per sample: 1 hr:06 min:22 s, total time for all samples: 26.9 days) and IDBA-UD the fastest (5 CPUs, median time per sample: 0 hr:11 min:01 s, total time for all samples: 4.4 days). MEGAHIT (5 CPUs, median time per sample: 0 hr:12 min:35 s, total time for all samples: 4.9 days) being slightly slower, and MetaSPAdes (5 CPUs, median time per sample: 0 hr:25 min:44 s, total time for all samples: 14.9 days) having a median assembly time about twice as slow as the other two metagenomic assemblers (Table 1; Figure 2a).

The UCE supermatrices created by PHYLUCE for each of the four assemblers contained on average 2,580 of the 2,590 UCE loci for Hymenoptera (Table 1). All matrices contained 501 species, but the size of the supermatrix and the percentage of missing data varied depending on the assembler (Table 1). Trinity, which is generally used as the default assembler in PHYLUCE, resulted in the shortest and most incomplete supermatrix with 2,579 loci representing 127,803 sites (40.5% variable) and 17.8% missing data. Among metagenomic assemblers, MetaSPAdes provided the largest and most complete supermatrix with 2,582 loci representing 156,456 sites (44.5% variable) and only 6.0% missing data. IDBA-UD retrieved 2,581 loci representing 132,403 sites (43.9% variable) with only 6.7% missing data, and MEGAHIT resulted in a supermatrix with 2,579 loci representing 147,589 sites (43.2% variable) but with 12.4% missing data. Note that fewer than 30 loci were retrieved for Phalacrocyrmex fugax (between four and 27 loci depending on the assembler). This is congruent with the original publication in which this low-quality library was not included in phylogenetic analyses (Branstetter, Danforth, et al.,

### TABLE 1 Summary statistics on assembly results according to the assembler used

<table>
<thead>
<tr>
<th>Assembler</th>
<th>Assembly time</th>
<th>UCEs</th>
<th></th>
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</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>Number of contigs</td>
<td>Number of loci</td>
<td>Matrix size</td>
<td>% Variable sites</td>
<td>% Missing data</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>IDBA-UD (5 CPU)</td>
<td>0 hr:11 min:02 s</td>
<td>30.544</td>
<td>2,581</td>
<td>132,403</td>
<td>43.9</td>
<td>6.7</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>MEGAHIT (5 CPU)</td>
<td>0 hr:12 min:35 s</td>
<td>114.392</td>
<td>2,579</td>
<td>147,589</td>
<td>43.2</td>
<td>12.5</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>MetaSPAdes (5 CPU)</td>
<td>0 hr:25 min:42 s</td>
<td>113.303</td>
<td>2,582</td>
<td>156,456</td>
<td>44.3</td>
<td>6.1</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Trinity (35 CPU)</td>
<td>1 hr:06 min:22 s</td>
<td>43.481</td>
<td>2,579</td>
<td>127,803</td>
<td>40.5</td>
<td>17.8</td>
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</table>

<table>
<thead>
<tr>
<th></th>
<th></th>
<th>Mitogenomes</th>
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<tbody>
<tr>
<td></td>
<td></td>
<td>Number of contigs</td>
<td>Number of species</td>
<td>Number of genes</td>
<td>AA matrix size</td>
<td>% Missing data</td>
<td>% Variable sites</td>
<td>NT matrix size</td>
<td>% Missing data</td>
</tr>
<tr>
<td>IDBA-UD (5 CPU)</td>
<td>4.2</td>
<td>499</td>
<td>13.04</td>
<td>3,764</td>
<td>20.9</td>
<td>86.7</td>
<td>13.635</td>
<td>26.1</td>
<td>85.8</td>
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<tr>
<td>MEGAHIT (5 CPU)</td>
<td>3.9</td>
<td>499</td>
<td>13.61</td>
<td>3,757</td>
<td>15.3</td>
<td>87.5</td>
<td>13.718</td>
<td>20.6</td>
<td>86.4</td>
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<td>MetaSPAdes (5 CPU)</td>
<td>3.8</td>
<td>501</td>
<td>13.73</td>
<td>3,766</td>
<td>14.6</td>
<td>88.9</td>
<td>13.713</td>
<td>19.8</td>
<td>87.1</td>
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<tr>
<td>Trinity (35 CPU)</td>
<td>4.2</td>
<td>500</td>
<td>13.37</td>
<td>3,760</td>
<td>18.0</td>
<td>86.9</td>
<td>13.648</td>
<td>26.7</td>
<td>86.1</td>
</tr>
</tbody>
</table>

Note: The values are averages over the 501 assemblies, except for the assembly time, which is a median value. The two parts of the table report specific statistics for (a) ultraconserved elements data, and (b) mitochondrial data. Note that 35 CPUs were used for Trinity whereas 5 CPUs were used for other assemblers.
Accordingly, we removed the *Phalacrocyrmex fugax* library (SRR5437956) from the data set.

### 3.2 Extracting mitochondrial sequences from UCE sequencing data

Depending on the assembler used in MitoFinder, mitochondrial reads were recovered in 499, 500 and 501 libraries out of a total of 501 (Table 1; Figure 2b). Overall, mitochondrial signal thus was detected in all libraries but only MetaSPAdes retrieved it in all species (Appendix S2). On average, 3.8 contigs per species were identified (Tables 1 and 2; Figure 2b) and 13.7 genes were annotated with MitoFinder (Figure 2c; Table 2). In 296/501 cases, MitoFinder was able to assemble a contig of more than 15,000 bp containing at least 13 annotated genes that probably represents the complete mitochondrial genome. In 52 of these cases, all 15 genes were annotated. In the remaining cases, the putative mitogenome contigs were missing one or two genes, mostly the short and divergent ATP8 (131/296), the 12S rRNA (29/296) and the 16S rRNA (10/296), which were present but not directly annotated by our BLAST-based procedure. By comparison, MITOBIM produced a mitochondrial contig for only 358 libraries for which an average of 3.51 genes were annotated representing 2,840.24 nucleotides on average.

After alignment and cleaning, mitochondrial genes obtained with MitoFinder were used to create nucleotide and amino acid supermatrices. To be consistent with UCE analyses, and despite the recovery of some mitochondrial signal, we ignored *Phalacrocyrmex fugax* in further analyses. In the nucleotide supermatrices (13 protein-coding + 12S and 16S rRNAs), we obtained 13 genes on average per species, which resulted in supermatrices with 13,679 nucleotide sites (86.4% variable) and 23.3% missing.
TABLE 2 Statistical comparison between the performances of the different assemblers

<table>
<thead>
<tr>
<th>Number of mtDNA contigs</th>
<th>IDBA-UD</th>
<th>MEGAHIT</th>
<th>MetaSPAdes</th>
<th>Trinity</th>
</tr>
</thead>
<tbody>
<tr>
<td>IDBA-UD</td>
<td>**(*)</td>
<td>**<em>(</em>)</td>
<td>NS (-)</td>
<td></td>
</tr>
<tr>
<td>MEGAHIT</td>
<td>**(-)</td>
<td>* (+)</td>
<td>***(-)</td>
<td></td>
</tr>
<tr>
<td>MetaSPAdes</td>
<td>***(-)</td>
<td>* (-)</td>
<td>***(-)</td>
<td></td>
</tr>
<tr>
<td>Trinity</td>
<td>NS (+)</td>
<td>***(+))</td>
<td>***(+))</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Number of coding mtDNA nucleotides</th>
<th>IDBA-UD</th>
<th>MEGAHIT</th>
<th>MetaSPAdes</th>
<th>Trinity</th>
</tr>
</thead>
<tbody>
<tr>
<td>IDBA-UD</td>
<td>***(-)</td>
<td>***(-)</td>
<td>* (-)</td>
<td></td>
</tr>
<tr>
<td>MEGAHIT</td>
<td>**<em>(</em>)</td>
<td>* (+)</td>
<td>***(-)</td>
<td></td>
</tr>
<tr>
<td>MetaSPAdes</td>
<td>**<em>(</em>)</td>
<td>* (+)</td>
<td>***(-)</td>
<td></td>
</tr>
<tr>
<td>Trinity</td>
<td>* (+)</td>
<td>***(-)</td>
<td>***(-)</td>
<td></td>
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</table>

<table>
<thead>
<tr>
<th>Number of mtDNA genes</th>
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<th>MEGAHIT</th>
<th>MetaSPAdes</th>
<th>Trinity</th>
</tr>
</thead>
<tbody>
<tr>
<td>IDBA-UD</td>
<td>***(-)</td>
<td>***(-)</td>
<td>***(-)</td>
<td>***(-)</td>
</tr>
<tr>
<td>MEGAHIT</td>
<td>**<em>(</em>)</td>
<td>* (+)</td>
<td>***(-)</td>
<td>**<em>(</em>)</td>
</tr>
<tr>
<td>MetaSPAdes</td>
<td>**<em>(</em>)</td>
<td>* (+)</td>
<td>***(-)</td>
<td>**<em>(</em>)</td>
</tr>
<tr>
<td>Trinity</td>
<td>**<em>(</em>)</td>
<td>***(+))</td>
<td>***(+))</td>
<td>***(-)</td>
</tr>
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</table>

<table>
<thead>
<tr>
<th>Number of UCE nucleotides</th>
<th>IDBA-UD</th>
<th>MEGAHIT</th>
<th>MetaSPAdes</th>
<th>Trinity</th>
</tr>
</thead>
<tbody>
<tr>
<td>IDBA-UD</td>
<td>***(-)</td>
<td>***(-)</td>
<td>***(-)</td>
<td>***(-)</td>
</tr>
<tr>
<td>MEGAHIT</td>
<td>**<em>(</em>)</td>
<td>* (+)</td>
<td>***(-)</td>
<td>**<em>(</em>)</td>
</tr>
<tr>
<td>MetaSPAdes</td>
<td>**<em>(</em>)</td>
<td>* (+)</td>
<td>***(-)</td>
<td>**<em>(</em>)</td>
</tr>
<tr>
<td>Trinity</td>
<td>**<em>(</em>)</td>
<td>***(+))</td>
<td>***(+))</td>
<td>***(-)</td>
</tr>
</tbody>
</table>

Note: Statistical significance was estimated with a paired nonparametric test (paired Wilcoxon test). *** p < .001; ** p < .01; * p < .05; NS = p > .05; and (+)/(-) indicates the result of the comparison between the row and the column.

3.3 Barcoding analyses

A total of 534 CO1 sequences retrieved from the 501 MetaSPAdes assemblies were used to verify species identification of the UCE libraries (Appendix S3). Similarity searches against BOLD and GenBank allowed us to confirm the species identity in only 312 cases probably because of the limited availability of CO1 barcoding data for these ant species. Moreover, in 42 cases, two or three CO1 sequence fragments were retrieved from the same UCE library. In seven of these cases, the slightly overlapping CO1 fragments probably resulted from poor assembly or erroneous annotation. However, in the 35 remaining cases, the genuine complete CO1 sequence overlapped with shorter contigs assembled from a minority of the reads, suggesting either cross-contaminations, NUMTs, endoparasites or bacterial symbionts. For instance, in Temnothorax sp. mmp11 (SRR5809551), a 391-bp fragment annotated as CO1 by MitoFinder was found to be 98.2% identical to both the Wolbachia pipientis wAlbB and Wolbachia Pel strain wPip genomes, which are bacterial endosymbionts of the mosquitoes Aedes albopictus and Culex quinquefasciatus, respectively. Also, in Sericomyrmex bondari (SRR5044901) and Sericomyrmex mayri (SRR5044856) short CO1 fragments best matched with nematodes. However, in the 312 cases for which CO1 barcoding allowed us to confirm the species identity of the UCE library, we did not detect any obvious cases of cross-contaminations where the CO1 extracted from a given library would have been identical to that of another library (Appendix S3).

3.4 Phylogenetic results

The ML topologies inferred from the different UCE supermatrices were very similar with an average quartet distance of 0.005 among assemblers (Appendix S4). However, the percentage of supported nodes (UFBS > 95) differed depending on the assembler: IDBA-UD (91.37%), MetaSPAdes (89.96%), MEGAHIT (89.56%) and Trinity (85.85%). In the following, we only discuss the phylogenetic results obtained with MetaSPAdes that provides the most comprehensive assemblies for both UCE and mitochondrial data (Table 1). The following 12 well-established subfamilies (with several species in the dataset) were retrieved with maximal UFBS support (100%): Amblyoponinae, Apomyrminae, Dolichoderinae, Dorylinae, Ectatomminae, Formicinae, Heteroponerinae, Myrmeciinae, Myrmicinae, Proceratiinae, Ponerinae, and Pseudomyrmecinae (Figure 3a). The two supergroups Formicoid and Poneroid were also retrieved with maximal UFBS support, as well as consensual phylogenetic relationships among Formicoid subfamilies (Ward, 2014).

For mitochondrial matrices, the percentage of supported nodes (UFBS > 95) with nucleotides also differed depending on the assembler and was higher than with the amino acids: MetaSPAdes (84.5% vs. 80.1%), MEGAHIT (84.0% vs. 79.4%), Trinity (83.3% vs. 80.4%) and IDBA-UD (80.2% vs. 78.0%). However, ML mitogenomic trees inferred from amino acids were more congruent with UCE topologies than those inferred from the mitochondrial nucleotides (average quartet distance = 0.035 vs. 0.063; Appendix S4). Among assemblers, the ML topologies inferred with amino acid matrices were highly congruent with an average quartet distance of 0.007 (Appendix S4). In the ML tree obtained with the MetaSPAdes supermatrix (Figure 3b), all ant subfamilies were retrieved with maximal UFBS support values except for Myrmicinae (93%), Ponerinae (97%) and Proceratiinae (99%) (Figure 3a). However, relationships among subfamilies were not congruent with UCE phylogenomic inferences.

data on average (Table 1). In the amino acid matrices (13 protein-coding genes), we obtained supermatrices with 3,762 amino acid sites (87.4% variable) and 17.2% missing data on average (Table 1).
except for Heteroponerinae + Ectatomminae (UFBS = 100) and Dolichoderinae + Aneuretinae (UFBS = 96; Figure 3a).

Finally, phylogenetic inference carried on mixed supermatrices composed of UCEs and mitochondrial amino acids resulted in ML topologies that were also highly similar among assemblers with an average quartet distance of 0.006 (Appendix S4). The percentages of supported nodes (UFBS > 95) were: IDBA-UD (91.2%), MEGAHIT (92.8%), MetaSPAdes (92.2%) and Trinity (90.4%). As with UCE
matrices, the 12 well-established subfamilies, the two supergroups Formicoid and Poneroid, and consensus Formicoid inter-subfamilies relationships (Ward, 2014) were all retrieved with maximal UFBS support.

4 | DISCUSSION

4.1 | Metagenomic assemblers as powerful tools for assembling UCEs

Currently, genomic and transcriptomic de novo assemblers are commonly used to assemble UCE loci from DNA capture sequencing data (Faircloth, 2016). Because metagenomic assemblers such as iDBA-UD, MEGAHiT and METASPADES have been designed to account for variance in sequencing coverage, they seem to be well adapted for targeted enrichment or DNA sequence capture data. Our results show that metagenomic assemblers are indeed faster at assembling UCE loci than the classically used, but computationally intensive, Trinity transcriptomic assembler. Furthermore, they seem more effective and lead to data sets containing more variable sites, less missing data and increased phylogenetic signal (Tables 1 and 2). Indeed, the topologies obtained with the metagenomic assemblers are very similar to the topology obtained with the Trinity-based supermatrix, contain a higher number of supported nodes (UFBS ≥ 95%) and are consistent with previous studies (Ward, 2014). Furthermore, assemblies obtained with the three metagenomic assemblers provide variable numbers of contigs (ranging from 30,544 to 114,392) resulting in differences in the completeness of the matrices (6.0%-17.8% of missing data for UCE matrices and 29.9%-41.3% for mitochondrial matrices) and in numbers of variable sites (for UCE, 40.5%-44.3%; for mtDNA, 77.2%-79.0%). Interestingly, for both UCE matrices and mtDNA matrices, METASPADES consistently provides more loci, more variable sites and less missing data. In addition, mitochondrial signal was extracted from all libraries only using METASPADES within MitoFinder. Despite a computation time on average twice that of the other two metagenomic assemblers, METASPADES was the more effective assembler for ant UCEs. This software therefore provides a much-needed alternative to Trinity for efficiently assembling hundreds of UCE libraries.

4.2 | MitoFinder efficiently extracts mitochondrial signal from UCE capture data

Ultraconserved elements are key loci exploited as target capture sequences in an increasing number of phylogenomic studies. DNA sequence capture methods are used to efficiently enrich targeted DNA regions in library preparation prior to sequencing, but non-targeted regions are always sequenced in the process, resulting in so called “off-target reads.” Interestingly, off-target reads could represent up to 40% of the sequenced reads in exome capture experiments (Chilamakuri et al., 2014) and many contigs not belonging to targeted UCE loci are typically assembled from UCE capture data (e.g., Faircloth et al., 2015; Smith, Harvey, Faircloth, Glenn, & Brumfield, 2014). Given this high proportion of off-target reads, we can expect that mtDNA could be found as off-target sequences in many target enrichment data. Accordingly, several studies have succeeded in extracting mtDNA from UCE libraries (e.g., do Amaral et al., 2015; Smith et al., 2014). The development of MitoFinder allowed the automatic extraction of mitochondrial signal from all 501 ant UCE libraries. This maximum success rate indicates that this approach is highly efficient at least in Formicidae. However, the success in retrieving mitochondrial sequences ultimately depends on the number of mitochondria contained in the tissue used for DNA extraction and library preparation. As expected, mitochondrial off-target reads are much more common in muscle and heart than in lung tissues in humans (D’Erchia et al., 2015). Similarly, mitochondrial sequences are probably rare or absent in libraries constructed from vertebrate blood, even in birds in which nucleated red blood cells contain mitochondria, but in very low numbers (Reverter et al., 2017). In invertebrates, our case study with a 100% success rate in ant UCEs demonstrates that mitochondrial sequences could probably be easily retrieved for many arthropod taxa as a by-product of target enrichment sequencing experiments. Finally, the comparison between MitoFinder and MiTOBIM emphasizes that the use of de novo assembly instead of iterative mapping is a suitable solution for recovering mitochondrial signal for groups with limited mitogenomic references.

4.3 | The value of complementary mitochondrial signal

Mitochondrial sequences could provide interesting and important complementary information compared to nuclear sequences. First, mtDNA can be used to confirm the identity of the species sequenced for conserved UCE loci. Here, we were able to confirm the identification of 312 ant species out of the 501 UCE libraries using CO1 barcoding without revealing a single case of obvious species misidentification. Given that ant UCE libraries have been constructed from museum specimens, the 501 CO1 sequences we annotated could be used as reference barcoding sequences in future studies. Then, even though we did not detect such cases, the high mutation rate and the absence of heterozygous sites in mtDNA also make it well adapted for cross-contamination detection analyses (Ballenghien et al., 2017).

Nevertheless, mitochondrial markers also have some well-identified limitations (Galtier et al., 2009). First, mtDNA could be inserted in the nuclear genome in the form of NUMTs (Bensasson, Zhang, Hartl, & Hewitt, 2001). NUMTs could potentially be assembled as off-target contigs in DNA capture libraries and we might have indeed extracted some fragments corresponding to NUMTs for the CO1 gene using MitoFinder (Appendix S2). Theoretically, NUMTs could be picked up by analysing the coverage of putative mitochondrial contigs as they are expected to have a coverage comparable to other
off-targets nuclear contigs, whereas genuine mitochondrial contigs should have a higher coverage. A second limitation of mtDNA exists in arthropods where maternally inherited intracellular bacteria are frequent. Among those bacteria, Wolbachia is particularly widespread and could distort the mitochondrial genealogy when a particular strain spreads within the host species, hitchhiking its linked mitochondrial haplotype (Cariou, Duret, & Charlat, 2017). Wolbachia infection is frequent among ants and could therefore be responsible of some mitonuclear discordance (Wenseleers et al., 1998). We indeed discovered such an instance with a Wolbachia CO1 sequence identified in Temnothorax spp. mmp11 (SRR5809551), which was confirmed by several assembled contigs matching to Wolbachia strain genomes in this sample.

Beyond the methodological aspects of species identification and potential cross-contamination detection, mitochondrial sequences could also be useful to tackle fundamental evolutionary questions. UCEs have also proved to be useful genetic markers for phylogeography and for resolving shallow phylogenetic relationships (Musher & Cracraft, 2018; Smith et al., 2014). In this context, mtDNA could also bring complementary information. In most animals, mtDNA has a maternal inheritance without recombination, which means that all mitochondrial genes behave as a single locus. This simplifies the interpretation of the phylogenetic pattern between closely related species or within subdivided populations of a species. Mitonuclear phylogenetic discordance could also reveal interesting phenomena involving hybridization, sex-biased dispersal and introgression (Bonnet, Leflois, Rouset, & Crochet, 2017; Toews & Brelsford, 2012). In practice, hybridization events are often identified using mitonuclear discordance (Li et al., 2016) and, in some cases, the mitochondrial introgression events have proven to be adaptive (Seixas, Boursot, & Melo-Ferreira, 2018). Nevertheless, in our ant case study, a detailed comparison of mitochondrial and UCE phylogenies did not reveal convincing occurrences of such discordances.

### 4.4 Ant phylogenetic relationships from 500 UCE and mitochondrial data

Both nuclear and mitochondrial data retrieved the most consensual phylogenetic relationships in the ant phylogeny (Borowiec et al., 2019; Branstetter, Ješovnik, et al., 2017; Ward, 2014). Twelve Formicidae subfamilies were recovered as monophyletic in all analyses, with both the nuclear and the mitochondrial data sets, confirming their robustness. However, the well-defined inter-subfamily relationships within Formicoids (Borowiec et al., 2019; Branstetter, Ješovnik, et al., 2017; Ward, 2014) were only supported by the UCE data set, but not by the mitochondrial amino acid data set. For example, the army ant subfamily (Dorylineae) was not retrieved as the sister-group of all other Formicoids, but was the closest relative of Pseudomyrmicinae (UFBS = 100). Similarly, contradicting the classical and well-defined relationship of Heteroponerinae + Ectatomminae as the sister-group of Myrmicinae (Borowiec et al., 2019; Branstetter, Ješovnik, et al., 2017; Ward, 2014), the mitochondrial data set supported an alternative relationship with Dolichoderinae + Aneuretinae (UFBS = 96). These differences suggest that mitochondrial data might be not well suited to resolve ancient phylogenetic relationships at the ant inter-subfamily level diverging about 100 million years ago (Moreau, Bell, Vila, Archibald, & Pierce, 2006), even if they look suitable for more recent nodes such as intra-subfamily relationships.

Interestingly, these topological incongruences between UCEs and mitochondrial genes also revealed different topologies regarding the existence of the Poneroid taxa, a controversial clade not always retrieved depending on the studies (Ward, 2014), but that tends to be retrieved in the most recent studies (Borowiec et al., 2019; Branstetter, Ješovnik, et al., 2017; UCE data set in this study) and is not recovered by our mitochondrial amino acid data set (Figure 3b).

The same applies to the phylogenetic placement of Apomyrminae, a subfamily grouped with either Leptanillinae or Amblyoponinae in previous studies (Ward, 2014), but that was grouped with Proceratiinae in our mitochondrial data set (UFBS = 98; Figure 3b).

For such controversial nodes, our study demonstrates that the nature of the phylogenetic markers can provide different results. Such differences between nuclear and mitochondrial data might be due to the substitutional saturation of mitochondrial data even at the amino acid level. This problem may actually be exacerbated in hymenopteran mitochondria that possess high AT content, translating into strongly biased codon usage and potentially leading to phylogenetic reconstruction artefacts (Foster & Hickey, 1999; Foster et al., 1997). Interestingly, such differences between mitochondrial and nuclear inference for ancient phylogenetic relationships are not observed with insects with less AT-rich mitochondrial genomes such as swallowtail butterflies (Allio et al., 2019; Condamine, Nabholz, Clamens, Dupuis, & Sperling, 2018) or tiger beetles (Vogler & Pearson, 1996). This calls for additional studies on both controversial and consensual ant inter-subfamily relationships with more comprehensive genome-wide data sets.

### 5 CONCLUSIONS

In this study, we developed the MitoFinder tool to automatically extract and annotate mitogenomic data from raw sequencing data in an efficient way. For the assembly step of our pipeline, we tested four different assemblers and showed that METASPADES is the most efficient and accurate assembler for both UCE and mitochondrial data.

Applying MitoFinder to ants, we were able to extract mitochondrial signal from 501 UCE libraries. This demonstrates that mitochondrial DNA can be found as off-target sequences in UCE sequencing data. Interestingly, mtDNA extracted from UCE libraries can also be used to: (a) confirm species identification with barcoding methods, (b) highlight potential sample cross-contamination, and (c) reveal potential cases of mitonuclear discordance caused by hybridization events leading to mitochondrial introgression. Finally, MitoFinder was developed with UCE libraries but our approach should also work with data obtained from other capture methods in which numerous
off-target reads are sequenced, as well as with transcriptomic and whole genome sequencing data, in which mitochondrial reads are over-represented.

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AUTHOR CONTRIBUTIONS

R.A. and F.D. conceived the ideas and designed methodology, analysed the data, and led the writing of the manuscript; R.A. implemented the MitoFinder software in part using code previously written by A.S.-B.; J.R., F.P. and B.N. contributed to the writing of the manuscript. All authors contributed critically to the drafts and gave final approval for publication.

DATA AVAILABILITY STATEMENT

The MitoFinder software is available from GitHub (https://github.com/RemiAllio/MitoFinder) and GitLab (https://gitlab.com/RemiAllio/mitofinder). Annotated mitogenomes and partial mitogenomic contigs containing at least two genes and 1,000 bp have been deposited in GenBank and are available in the Third Party Annotation Section of the DDBJ/ENA/GenBank databases under accession numbers TPA: BK012118–BK012857 (Appendix S5). The full analytical pipeline, MitoFinder results including UCE and mtDNA contigs for all assemblers, phylogenetic data sets and corresponding trees can be retrieved from zenodo.org (https://doi.org/10.5281/zenodo.3231390).

We used UCE raw sequencing data for 501 ants produced in 10 phylogenomic studies (Blaimer et al., 2015, Blaimer, LaPolla, Branstetter, Lloyd, & Brady, 2016; Branstetter, Danforth, et al., 2017, Branstetter, Ješovnik, et al., 2017, Branstetter, Longino, et al., 2017; Faircloth et al., 2015; Ješovnik et al., 2017; Pierce et al., 2017; Prebus, 2017; Ward & Branstetter, 2017). (see Appendix S1)

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SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section.