

1 **Single-Copy Orthologs (SCOs) improve species discrimination: A case study in**  
2 **subgus *Jensoa* (*Cymbidium*)**

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11

12 **Abstract**

13 Standard barcodes and ultra-barcodes face challenges in delimitation and discrimination  
14 of closely related species with deep coalescence, hybrid speciation, gene flow or low sequence-  
15 variation. Single copy orthologs (SCOs) have been recommended as standardized nuclear  
16 markers in metazoan DNA taxonomy. Here, we assessed the performance of SCOs in identifying  
17 recently diverged species in subgenus *Jensoa* (*Cymbidium*) which has been poorly settled by  
18 ultra-barcode. More than 90% of target 9094 reference SCOs inferred from three genomes of  
19 *Cymbidium* were successfully retrieved for all 11 representative species in subg. *Jensoa* by  
20 ALiBaSeq from as low as 5× depth whole genome shotgun sequences. Species tree reconstructed  
21 from multiple refined SCO matrices under multispecies coalescent model successfully  
22 discriminated all species and discerned wrongly identified or labeled species. Plentiful and

23 refined SCOs matrices obtained by implementing our pipeline facilitate not only phylogenetic  
24 study, but also high-resolution species diagnosing. Biparentally inherited SCOs as multi-locus  
25 marker not only advances the force of DNA barcoding, but also facilitates an eventual transition  
26 to species-tree-based barcoding strategies.

27

28 **Keywords**

29 Single-Copy Orthologs (SCOs), Ultrabarcoding (UBC), species discrimination, closely related  
30 species, *Jensoa*, pipeline

31

32 **1 | INTRODUCTION**

33 Species recognition is paramount for science and society. DNA barcoding, a tool  
34 proposed by Hebert 20 years ago (Hebert et al., 2003), has proven instrumental in plant species  
35 identification and discovery based on genetic variations of DNA sequences (Hollingsworth et al.,  
36 2016). Four easily amplified gene regions, *rbcL*, *matK*, *trnH-psbA*, and ITS (internal transcribed  
37 spacers), have been agreed upon as the standard plant DNA barcodes (Hollingsworth et al., 2009;  
38 Kress et al., 2005; Li et al., 2011). However, traditional standard barcodes failed in many  
39 evolutionarily young species for lacking sequence divergence (Li et al., 2015; Spooner, 2009; van  
40 Velzen et al., 2012). Ultrabarcoding (UBC), using whole chloroplast genome (Kane & Cronk,  
41 2008) or ribosomal DNA (rDNA) repeat unit (Kane et al., 2012) as extended barcodes, has  
42 overcome the inherent limitations of the traditional single- or multi-locus DNA barcodes by  
43 offering sufficient variable characters (Coissac et al., 2016). By assembling plastomes and rDNA  
44 clusters from low-coverage shotgun sequencing of genomic DNA, universal primers and loci  
45 preference is not annoyance any more (Kress et al., 2005; Straub et al., 2012). Ultrabarcoding has

46 become more highly discriminating and efficient plant DNA barcode to resolve some difficult  
47 taxa (Ji et al., 2019; Kane et al., 2012; Parks et al., 2009; Šlipiko et al., 2020; Yang et al., 2013;  
48 Zeng et al., 2018). However, plastomes and rDNA repeats could not address the limitations in  
49 discrimination species involving introgression, hybridization, incomplete lineage sorting (ILS) or  
50 recent divergence (RuhSAM et al., 2015; Weitemier et al., 2014). Species level polyphyly or  
51 paraphyly are common in closely related species, especially for groups that diverged recently (Z.  
52 F. Liu et al., 2021; van Velzen et al., 2012; Yu et al., 2022).

53 Nuclear genes, which have a preponderance of biparental inheritance over organelle  
54 genes, could considerably improve the accuracy and robustness of DNA barcoding (David et al.,  
55 2021; Huang et al., 2022; Small et al., 2004; Wang et al., 2019; Zimmer & Wen, 2012). ITS and  
56 rDNA do not always track both parents' genome in hybrids and allopolyploids due to lack of  
57 intragenomic uniformity and complex evolutionary fates (Álvarez & Wendel, 2003; Bailey et al.,  
58 2003). Ultra-conserved elements (UCEs) and restriction site-associated DNA (RAD) are also  
59 problematic because of insufficient intraspecific variation or non-homologous flanking region  
60 sequences (Eberle et al., 2020). The compromise between cost and accuracy of the barcoding  
61 results has been broken by progress in sequencing technologies. Whole transcriptome, DNA  
62 target enrichment and whole genome sequencing have become affordable for sampling hundreds  
63 of single copy target loci from nuclear genome (Lemmon et al., 2012; Weitemier et al., 2014;  
64 Wen et al., 2013; Xi et al., 2013). Single copy orthologs (SCOs) are protein-coding genes under  
65 strong selection to be present in one single copy, and they allow a more reliable assessment of  
66 homology to serve as highly suitable and universal makers (Waterhouse et al., 2011). The  
67 number of SCOs increases with increasing relatedness of the species chosen so the number of  
68 inferred SCOs of lower taxonomic levels are larger than higher lineages (Emms & Kelly, 2019;

69 Smith & Hahn, 2021). Putative SCOs could be recovered by two ways, a) to identify  
70 corresponding reads of reference SCOs and then to assemble each putative SCO, b) to assemble  
71 the whole genome and then to extract each putative SCO by querying them to the whole assemble  
72 (Knyshov et al., 2021). SCOs have successfully improved and homogenized species delimitation  
73 and discrimination in Metazoa (Dietz et al., 2021; Joshi et al., 2022). SCOs have been used as  
74 molecular markers in plant phylogenetics for several year (Hu et al., 2023; Huang et al., 2022;  
75 Johnson et al., 2018; B. B. Liu et al., 2021; Liu et al., 2022; G. Zhang et al., 2023; Zhang et al.,  
76 2012), but no report on species identification yet.

77 Subgenus *Jensoa* (Raf.) Seth & Cribb (Orchidaceae; Epidendroideae; Cymbidieae;  
78 Cymbidiinae; *Cymbidium*) consisting of about 20 species, are mostly terrestrial growing in  
79 tropical and subtropical Asia (Liu et al., 2006; Zhang et al., 2021). The well-known Asian  
80 Cymbidiums cultivated more than 2000 years in China are all from this subgenus and comprise  
81 thousands of artificial hybrids (Du Puy et al., 2007; Hew, 2001). Subgenus *Jensoa* diverged less  
82 than 4 Ma (G. Zhang et al., 2023), and species from this subgenus had little morphological  
83 variation before flowering. Hybridization is as common as poaching in *Jensoa*, therefore,  
84 accurate identification of this subgenus is essential to breeding and trade (Liu et al., 2006).  
85 Previous effort has failed by using standard barcodes, plastomes and un-assembled reads (L.  
86 Zhang et al., 2023). As an example of how SCOs could be applied, we will here examin the  
87 power of SCOs on discriminating *Cymbidium* subgenus *Jensoa* (Orchidaceae), recently diverged  
88 species with frequently hybridization. Lineage specific reference SCOs were firstly inferred from  
89 three annotated whole genomes of species in *Cymbidium*. Putative SCOs were then recovered  
90 from deep genome skimming data of 11 *Jensoa* species with multiple samples. We aim to address  
91 these three questions: (i) Is it possible to recover the vast majority of SCOs from genomic

92 sequencing data with lower than 10× depth? (ii) How to achieve convincing SCoS matrices and  
93 subsequent species tree by a convenient pipeline? (iii) To assess the feasibility of SCoS in plant  
94 species identification using low-pass sequencing data.

95

## 96 **2 | MATERIALS AND METHODS**

### 97 **2.1 | Plant material and data collection**

98 According to our previous study (L. Zhang et al., 2023), 11 species of *Cymbidium* subg.  
99 *Jensoa* were chosen for their nonmonophyly except *C. omeiense* and *C. qiubeiense*.  
100 Each species with four individual representatives were sequenced at first to output about 100 Gb  
101 genomic sequencing data. 33 of these 44 vouchers were identical to our previous study (L. Zhang  
102 et al., 2023). *Cymbidium mannii* (subg. *Cymbidium*) (Fan et al., 2023), *Cymbidium tracyanum*  
103 (subg. *Cyperorchis*) from our project of comparative genomics of *Cymbidium* were included as  
104 the closely related outgroup. Three species from the same tribe Cymbidieae were chose as the  
105 distantly related outgroup, two from subtribe Cymbidiinae (*Grammatophyllum scriptum*,  
106 *Thecopus maingayi*), one from subtribe Acriopsidinae (*Acriopsis javanica*). Three additional  
107 collections of *C. ensifolium* (H3204, ZL442, ZL443) and another published collection (Vocher  
108 RL0671, accession SRR7121924) (Liu et al., 2019) were further added to verify the intraspecific  
109 genetic variation of *C. ensifolium* (Table 1). DNA extraction and genomic sequencing methods  
110 are same as previously described (L. Zhang et al., 2023). Raw data were filtered by Fastp v0.22.0  
111 with default parameters (Chen et al., 2018).

112

### 113 **2.2 | genome size estimation**

114                   Genome size estimates for all samples were obtained using flow cytometry (FCM). About  
115                   20mg fresh young leaf tissue was chopped by scalpel in a Petri dish containing ice-cold Modified  
116                   Gitschier Buffer (45 mM MgCl<sub>2</sub>·6H<sub>2</sub>O, 20 mM MOPS, 30 mM Trisodium citrate, 1% (W/V)  
117                   PVP 40, 0.2% (V/V) Triton X-100, 10 mM Na<sub>2</sub>EDTA, pH 7.0). Homogenate was filtered through  
118                   a 42-mm nylon mesh and stained with propidium iodide (50 mg/ml) and analyzed using a BD  
119                   FACSCalibur Flow Cytometer (Table S1).

120                   44 clean pair-end genomic data were submitted to JellyFish v2.3.0 (Marçais & Kingsford,  
121                   2011) to compute histogram of k-mer frequencies of each sample using sub-command 'jellyfish  
122                   count -C -m21' and 'jellyfish histo -h 3000000'. GenomeScope v2.0(Ranallo-Benavidez et al.,  
123                   2020) were then employed to estimate the genome size of each sample with default parameters.  
124                   Because GenomeScope2 failed in some samples, original data of all individuals were sub-  
125                   sampled to 0.5~4X by seqtk v1.3-r106 (Li, 2012) and merged by BBMerge v39.01 (Bushnell et  
126                   al., 2017). Genome sizes of all individuals were then estimated by RESPECT v1.3.0 (Sarmashghi  
127                   et al., 2021) (Table S1).

128                   Table 1. Species information of all materials used in this study  
129

Species	Voucher	Locality	Clean data (Gbp)	Genome Size (Gb)	Sequencing Depth
<i>C. tortisepalum</i>	18HT2037	Lijiang, Yunnan, China	115.80	3.64	31.81
	ZL55	KBG, Yunnan, China	118.75		32.62
	ZL56	Baoshan, Yunnan, China ‡	113.30		31.13
	ZL70	Dali, Yunnan, China	114.29		31.40
<i>C. goeringii</i>	15043	Enshi, Hubei, China	132.81	4.88	27.21
	16264	Chongqing, China	102.04		20.91
	16266	Chongqing, China ‡	110.48		22.64
	16280	Chongqing, China ‡	115.07		23.58

<i>C. serratum</i>	H4001	Baise, Guangxi, China	100.78	3.70	27.22
	H4002	Baise, Guangxi, China ‡	104.15		28.13
	H4003	Baise, Guangxi, China ‡	125.60		33.92
	ZL453	Qianxinan, Guizhou, China ‡	109.45		29.56
<i>C. omeiense</i>	15002	Zhangjiajie, Hunan, China ‡	120.23	3.82 †	31.46
	15009	Zhangjiajie, Hunan, China ‡	102.52		26.83
	15032	Enshi, Hubei, China ‡	130.21		34.07
	15034	Enshi, Hubei, China ‡	107.39		28.10
<i>C. kanran</i>	18HT1428	Honghe, Yunnan, China ‡	147.91	4.22	35.05
	18HT1873	Lijiang, Yunnan, China ‡	123.40		29.24
	H3602	Qianxinan, Guizhou, China ‡	95.08		22.53
	H3605	Qianxinan, Guizhou, China ‡	99.29		23.53
<i>C. faberi</i>	15019	Enshi, Hubei, China ‡	125.75	3.12	40.30
	15020	Enshi, Hubei, China ‡	149.05		47.77
	15030	Enshi, Hubei, China ‡	112.40		36.03
	ZL39	KBG, Yunnan, China	107.26		34.38
<i>C. sinense</i>	ZL3	Honghe, Yunnan, China ‡	102.69	4.62	22.22
	ZL4	Honghe, Yunnan, China ‡	110.76		23.96
	ZL444	Honghe, Yunnan, China ‡	114.16		24.70
	ZL445	Yunnan, China ‡	107.57		23.27
<i>C. qiubeiense</i>	19HT2776	Qianxinan, Guizhou, China ‡	160.75	6.19	25.97
	ZL13	Qianxinan, Guizhou, China ‡	170.09		27.48
	ZL14	Qianxinan, Guizhou, China ‡	135.91		21.96
	ZL457	Qianxinan, Guizhou, China	105.37		17.02
<i>C. cyperifolium</i> var. <i>szechuanicum</i>	ZL19	Qianxinan, Guizhou, China ‡	131.17	4.41	29.72
	ZL20	Qianxinan, Guizhou, China ‡	154.88		35.09
	ZL64	Qianxinan, Guizhou, China ‡	112.20		25.42
	ZL65	Baise, Guangxi, China ‡	146.16		33.12
<i>C. cyperifolium</i>	14942	Hechi, Guangxi, China	90.55	4.09	22.14
	16268	Chongqing, China ‡	102.68		25.10
	ZL21	Qianxinan, Guizhou, China ‡	103.54		25.32
	ZL22	KBG, Yunnan, China	105.78		25.86
<i>C. ensifolium</i>	13553	Baise, Guangxi, China	107.19	3.18	33.76
	18HT2190	Linzhi, Xizang, China ‡	144.94		45.65
	H3201	Baise, Guangxi, China ‡	138.53		43.63

	H3202	Baise, Guangxi, China	112.05		35.29
	H3204 †	KBG, Yunnan, China	30.45		9.59
	ZL442 †	Wenshan, Yunnan, China ‡	31.95		10.06
	ZL443 †	Nujiang, Yunnan, China ‡	31.46		9.91
	RL0671	Ruili, Yunnan, China §	73.50		23.15
<b>Outgroup</b>					
<i>C. manii</i>	YYL1809	KBG, Yunnan, China	Chromosome-level assembly	2.75	/
<i>C. tracyanum</i>	ZL1	KBG, Yunnan, China ‡	Chromosome-level assembly	3.95	/
<i>Grammatophyllum scriptum</i>	Cymw4 †	Taiwan, China ‡	56.09	/	/
<i>Acriopsis javanica</i>	Cymw6 †	Thailand ‡	62.02	/	/
<i>Thecopus maingayi</i>	Cymw7 †	Thailand ‡	31.30	/	/

130

131 Note: †, 6 additional individuals sequenced more than 25 Gb genomic data; KBG, Kunming Botany Garden; ‡,  
132 vouchers same to our previous study (L. Zhang et al., 2023); §, accessions (Liu et al., 2019); ¶, estimated genome  
133 size according to RESPECT result, not by flow cytometry.

134

### 135 2.3 | Genome assembling and Single-Copy Orthologs retrieval

136 To efficiently assemble to the approximately 5 TB clean genomic data, ultrafast, memory-  
137 efficient short read assemblers were chosen. Clean pair-end reads were assembled by  
138 SOAPdenovo v2.04 (Luo et al., 2012) with command `SOAPdenovo-63mer all -K 41` or  
139 MegaHit v1.2.9 with default parameters. Protein annotations of our three *Cymbidium* genomes  
140 (*C. tortisepalum*, *C. manii*, *C. tracyanum*) were subject to OrthoFinder v2.3.8 (Emms & Kelly,  
141 2019) to obtain 9094 single copy orthologues. These 9094 protein sequences used as queries to  
142 TBLASTN against all short-read assemblies and two chromosomal level assemblies. ALiBaSeq  
143 v1.2 (Knyshov et al., 2021) was employed to extract these 9094 single copy orthologs from the  
144 TBLASTN results with parameters ` -x a -e 1e-10 --is --amalgamate-hits --ac aa-tdna`. To

145 eliminate the introns extracted by ALiBaSeq, the default scoring matrix of TBLASTN were  
146 modified to PAM30. To test the performance of ALiBaSeq at lower sequencing depth, i.e.,  
147 below 10× coverage recommended by previous study (B. B. Liu et al., 2021), 25% subsampling  
148 was imposed on all clean genomic data of all 44 individuals.

149

150 **2.4 | chloroplast genomes and nrDNA assembling**

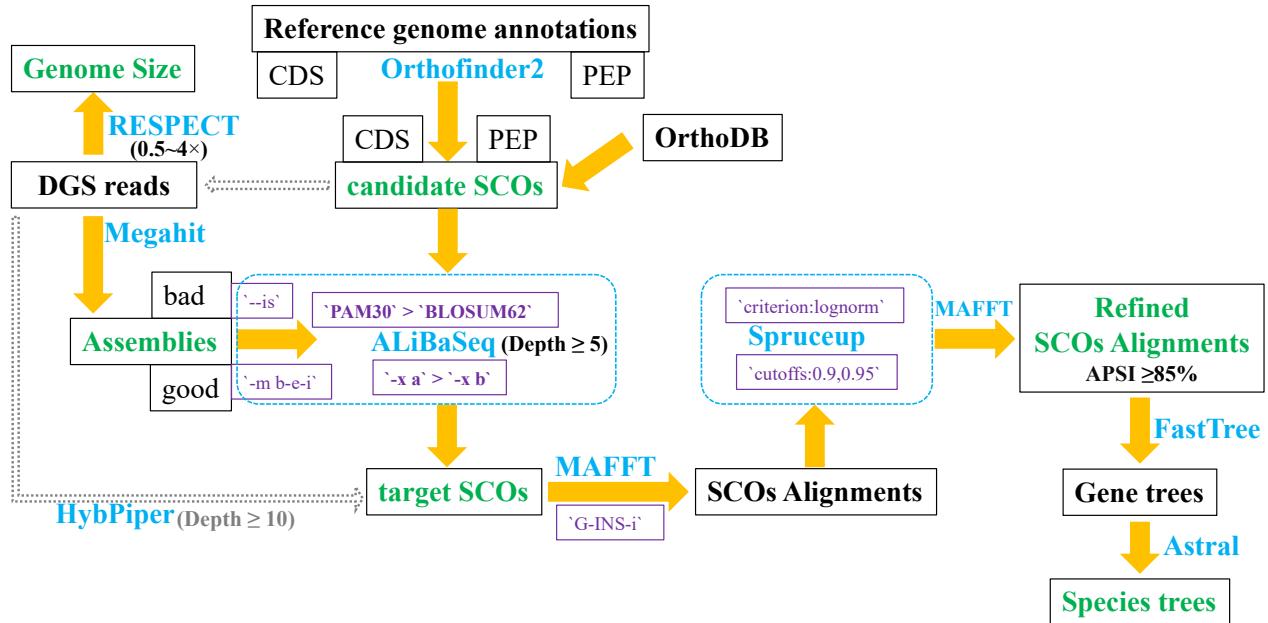
151 Chloroplast genomes and nuclear ribosomal DNA (nrDNA) clusters were de novo  
152 assembled using GetOrganelle v1.7.5 (Jin et al., 2020) and/or NOVOPlasty v4.3.1 (Dierckxsens  
153 et al., 2016). Plastome of *C. sinense* (accession: NC\_021430) and nrDNA of *C. macrorhizon*  
154 (accession: MK333261) were chosen as references. SSCs of all assembled plastomes were  
155 adjusted to the same direction when necessary. nrDNA sequences of each individual were  
156 manually stitched according to the mapping results if they were not complete in Geneious R9  
157 (Biomatters).

158

159 **2.5 | Alignment filtering and tree building**

160 The single copy homologs matrix recovered by ALiBaSeq were aligned by MAFFT v7.508 with  
161 parameters `--globalpair` (Katoh & Standley, 2013). Average pairwise sequence identity (APSI)  
162 of each alignment, a measure for sequence homology computed with ALISTAT v1.9g from the  
163 squid package (Eddy, 2005). To reduce the hazard of non-homologous region, Spruceup  
164 v2022.2.4 (Borowiec, 2019) was used to filter. Only alignments with no missing data and APSI  
165 larger than 85% were chosen for subsequent analysis. Approximately-maximum-likelihood gene  
166 trees were built by FastTree v2.1.10 (Price et al., 2010) with parameters `--gtr --gamma --nt` using

167 the refined alignments. Species trees were inferred using ASTRAL v5.7.8 and normalized quartet  
 168 scores were retrieved from logfiles (Mirarab et al., 2014). (FIGURE 1)



169  
 170 **FIGURE 1.** Graphical overview of the pipeline of this study. Softwires names were depict by  
 171 blue color, and key parameters were in purple. Dashed gray arrows indicate another way to  
 172 recover putative SCOs which is not fully testified in this study. APSI, Average pairwise sequence  
 173 identity.

174

### 175 3. | RESULTS

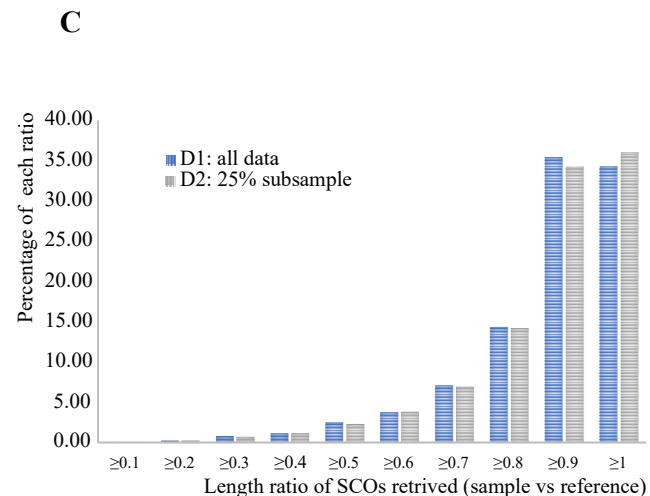
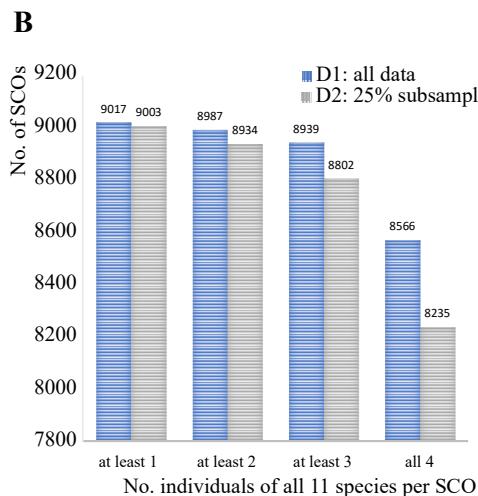
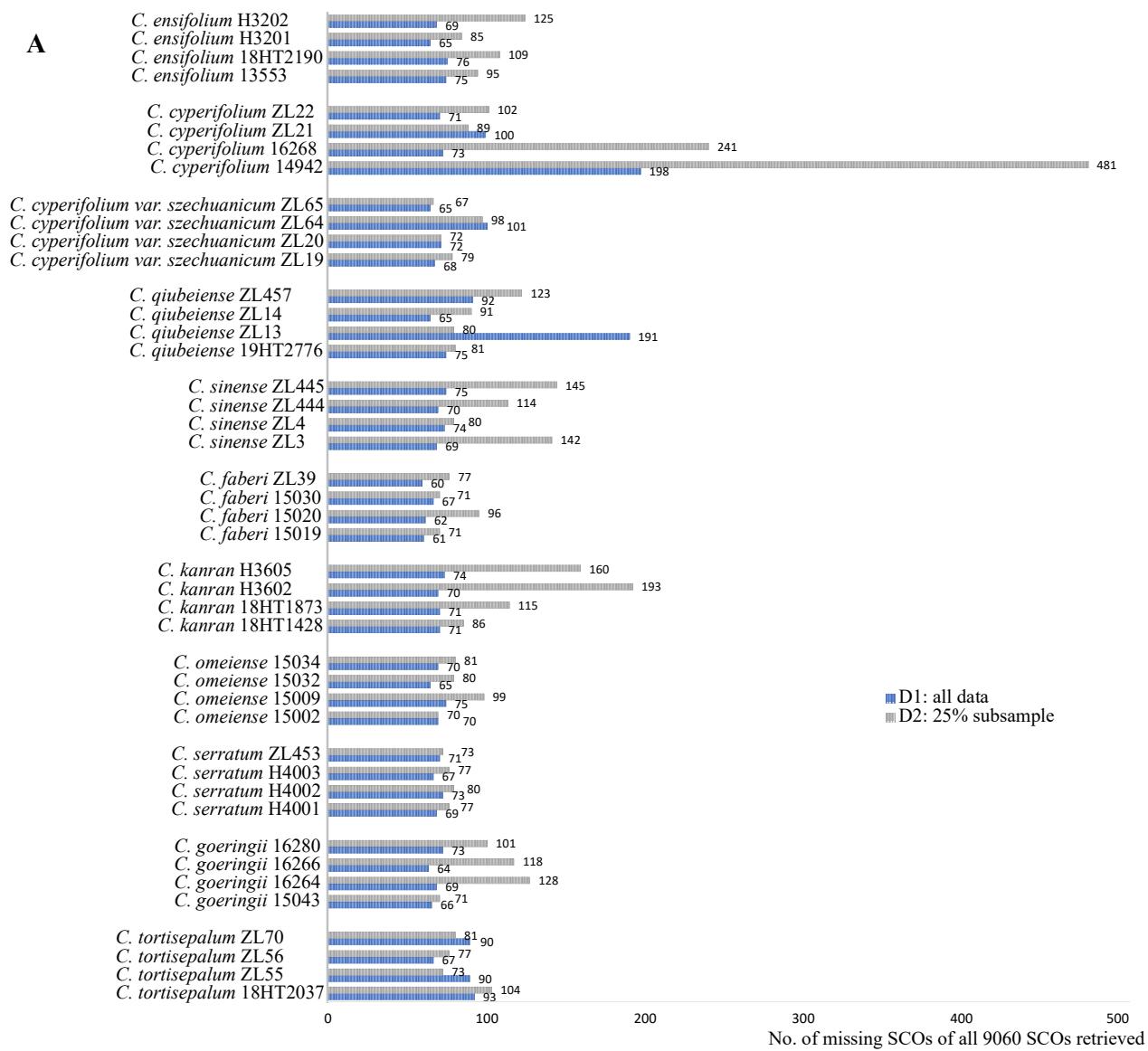
#### 176 3.1 | Genome sizes of species in *Cymbidium* subg. *Jensoa*

177 To accurately estimate the sequencing depth of each species, genome size were measured  
 178 firstly. According to the flow cytometry results, the average genome size of all 11 species in  
 179 subg. *Jensoa* was 4.1 Gb, which is same to the mean value of *Cymbidium* in plant DNA C-values  
 180 database (Leitch et al., 2019). *C. qiubeiense* has the largest genome (6.19Gb), while *C. faberi* and  
 181 *C. ensifolium* have the smallest genome (about 3.1 Gb) (Table 1). Genome sizes estimated by  
 182 GenomeScope2 are not always close to the flow cytometry, which may be caused by insufficient

183 sequencing depth or wrong k-mer peaks chosen by GenomeScope2. Genome sizes calculated by  
184 RESPECT are slightly larger (about 1.19-fold) than flow cytometry (Table S1). According to the  
185 genome sizes of each species, the sequencing depth of all 44 individuals is between 17.02× and  
186 47.77× (average 29.46×), and the depth of 25% subsampled of the 44 individuals and 3 additional  
187 added *C. ensifolium* is between 4.26× and 11.94× (Table 1).

188 **3.2 | putative Single-Copy Orthologs recovery**

189 The average assembly sizes of all 44 individuals with about 100 Gb data (**D1**) and 25%  
190 subsampled (**D2**) were 7.18 Gb and 3.75 Gb, respectively. The abnormal smallest assembly size  
191 of *C. cyperifolium* 14942 (1.56Gb and 0.4Gb for D1 and D2, respectively), was probably caused  
192 by extremely high duplication rate when genomic sequencing. The actual depth of voucher 14942  
193 could be much smaller than 22.14× (Table S1). ALiBaSeq succeeded to retrieve 9060 SCOs from  
194 each dataset (D1 and D2), with only 2 SCOs different from each other. For each species, 98.95%  
195 and 98.06% of all 99660 SCOs (9060 multiplied by 11) were obtained in its all four individuals  
196 from dataset D1 and D2, respectively (Table S2). On average, 99.5% and 99.2% SCOs were  
197 successfully retrieved from each individual in both dataset (D1 and D2), with the lowest  
198 efficiency from *C. cyperifolium* 14942 (FIGURE 2A). From the perspective of SCO, 9017 and  
199 9003 of 9060 SCOs were acquired from at least one individual of each species in dataset D1 and  
200 D2 respectively. 8566 and 8235 of 9060 SCOs were retrieved from all 4 individuals of each  
201 species in dataset D1 and D2 respectively (FIGURE 2B). The ratios of mean length of retrieved  
202 SCOs to the mean length of corresponding reference SCOs were mostly bigger than 0.9 (the  
203 accumulative frequencies were 69.8% and 70.3% in D1 and D2, respectively) (FIGURE 2C,  
204 Table S3). Overall, ALiBaSeq performed great in both recovering efficiency and  
205 representativeness of recovered SCOs.



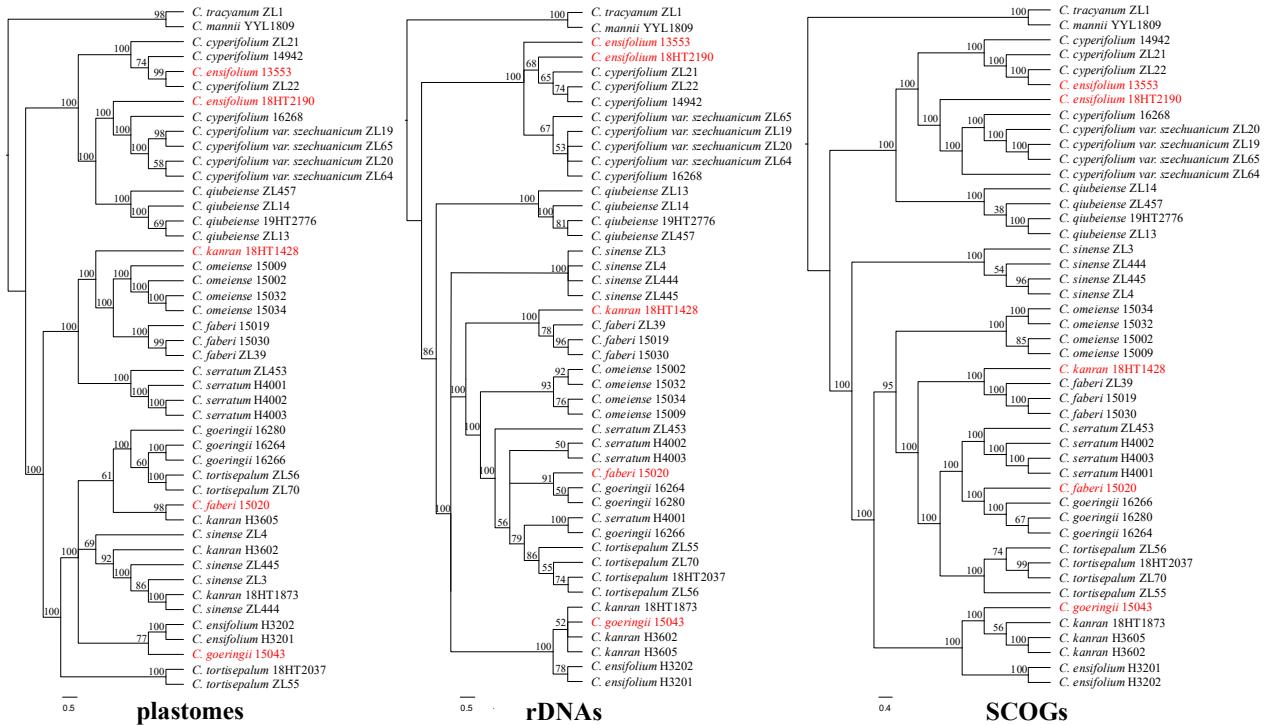
207

208 **FIGURE 2.** Performance of ALiBaSeq. **(A)** The number of missing SCoS of all 9060 SCoS  
209 extracted in each individual in dataset D1 and D2; **(B)** The Number of SCoS extracted in all  
210 species per SCoS. **(C)** Frequency distribution of ratio of mean length of retrieved SCoS to the  
211 mean length of corresponding reference SCoS.

212

213 **3.3 | SCoS perform better than plastomes and rDNA**

214 Our previous study had showed that the identification rate of *C. subg. Jensoa* was the  
215 lowest in genus *Cymbidium* by using plastome as barcode (L. Zhang et al., 2023). After curation  
216 of the plastomes of 44 individuals of 11 species in this study, *C. cyperifolium var. szechuanicum*  
217 and *C. serratum* were successfully identified. rDNA clusters succeeded to identify *C.*  
218 *tortisepalum* and *C. sinense* other than plastomes did, but failed to identify *C. cyperifolium var.*  
219 *szechuanicum* and *C. serratum*. SCoS (extracted from dataset D1 and two outgroup)  
220 outperformed rDNA clusters and plastomes, only *C. ensifolium*, *C. kanran*, *C. faberi*, and *C.*  
221 *goringii* failed to form monophyletic clade (FIGURE 3). Species trees reconstructed by SCoS  
222 recovered from dataset D1 (all data) and D2 (25% subsample) had the same topology and branch  
223 support value (Supplementary FIGURE 1). It strongly foretold that, deep genome skimming  
224 (DGS) with as low as 4 - 5 $\times$  coverage sufficed ALiBaSeq to recover abundant SCoS to  
225 reconstruct robust species tree. ALiBaSeq outperformed HybPiper taking advantage of half  
226 sequencing depth (B. B. Liu et al., 2021). It's worth noting that, the four species which SCoS  
227 failed to identified also occurred abnormally in trees reconstructed by plastomes and rDNA  
228 clusters. These may be vouchers mis-identified or disorder during DNA extraction or genomic  
229 sequencing, especially these three vouchers, 18HT1428, 15020 and 15034 (FIGURE 3,  
230 Supplementary FIGURE 1). Additional vouchers need to include to address these issues.



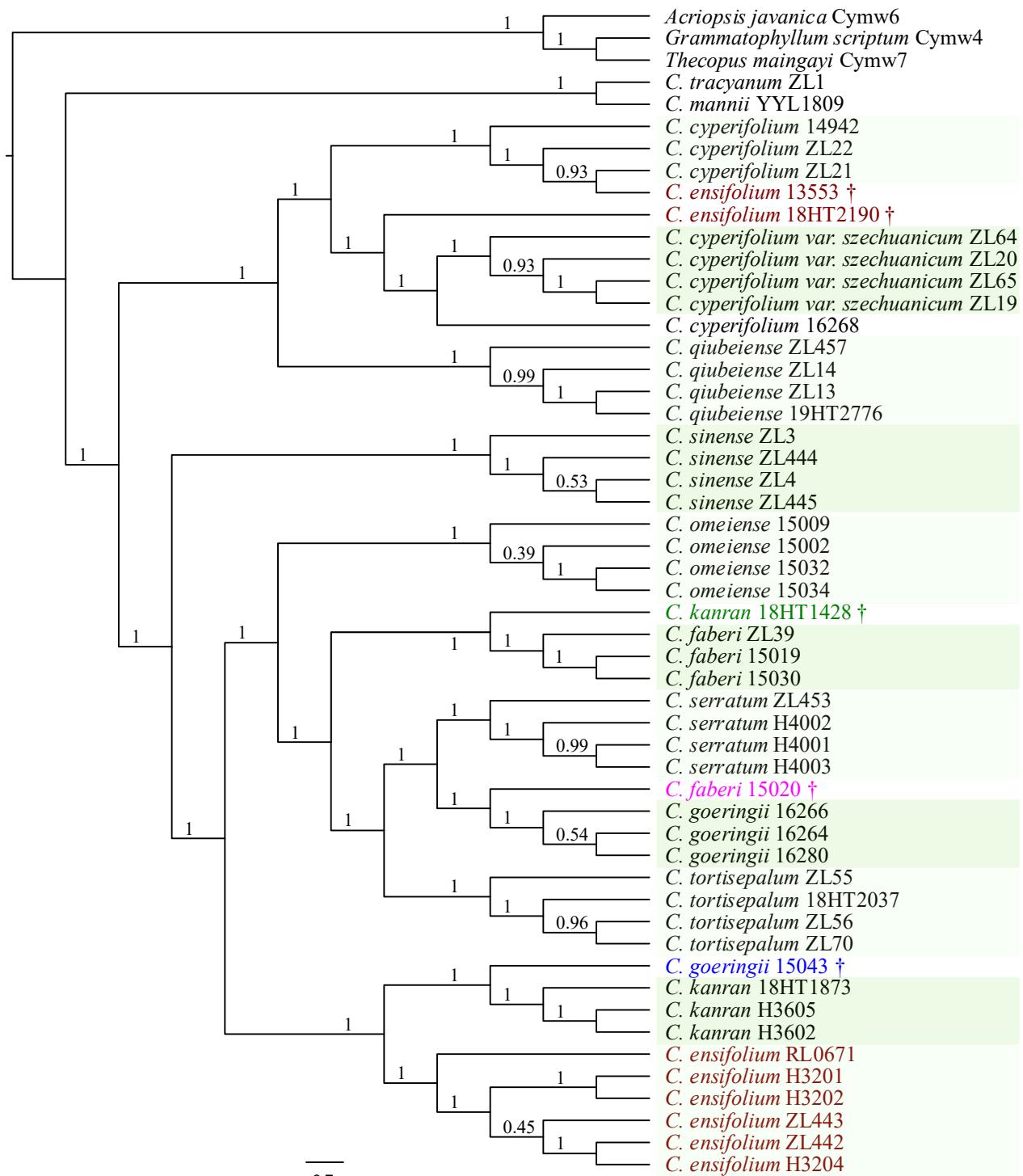
231 **FIGURE 3.** Cladogram tree-based species discrimination of *Jensoa* reconstructed by different  
 232 dataset. Vouchers which are possibly wrong identified are indicated in red. Numbers above each  
 233 brancher expressed as percentage are SH-like (Shimodaira-Hasegawa) local support value in  
 234 plastomes and rDNA trees, and LPP (local posterior probability) in SCOs tree (reconstructed by  
 235 6083 SCOs with  $APSI \geq 85\%$ ).  
 236

237  
 238 **3.4 | Adding individuals to validate the efficacy of SCOs as the barcode**

239 After adding four vouchers of *C. ensifolium* and three vouchers as distantly related  
 240 outgroups to the dataset D2, the performance of SCOs were proved. The two vouchers of *C.*  
 241 *ensifolium*, 13553 and 18HT2190, were both misidentified. They should be *C. cyperifolium* or *C.*  
 242 *cyperifolium var. szechuanicum*. 4 individuals of *C. cyperifolium var. szechuanicum* formed a  
 243 monophyletic clade rather than *C. cyperifolium* (FIGURE 4). In this study, we re-produced the  
 244 genomic data of vouchers by redoing all the molecular experiments including the vouchers used  
 245 in our previous study (L. Zhang et al., 2023). The three vouchers which confused with each other,

246 18HT1428, 15020 and 15034, could be incorrectly identified or distributed before their molecular  
247 materials were sent to us. These two vouchers, 18HT1428, 15020, also clustered around *C. faberi*  
248 and *C. kanran* respectively in our previous study (L. Zhang et al., 2023). If we removing these 5  
249 vouchers, all conspecific samples would be reciprocally monophyletic except *C. cyperifolium*  
250 (voucher 16268). It should be noticed that, SCoS had the power to discriminate all species of *C.*  
251 subg. *Jensoa*, and SCoS may be the most powerful barcode to identification of lower taxonomic  
252 levels where recent divergence or ancient rapid radiation have resulted in limited sequence  
253 variations.

254



255

0.7

256 **FIGURE 4.** Species tree of 11 species of reed grasses (Cyperaceae) reconstructed by 5732 SCOs with APSI  $\geq 85\%$ .

257 Numbers above each brancher expressed as decimal are LPP (local posterior probability). Species  
258 in color contains misidentified vouchers which are marked with dagger symbol (†).

259

260 **4. | Discussion**

261 **4.1 | Choosing of reference SCOs**

262 We hooked the 9094 baits (reference SCOs) needed by ALiBaSeq by OrthoFinder using  
263 the annotated representative protein sequences as the input in this study. Afterward we found that  
264 by chance, the default software used by OrthoFinder was DIAMOND, which gave 1-2% accuracy  
265 decrease but with a runtime of approximately 20 $\times$  shorter (Emms & Kelly, 2019). When using  
266 BLASTP instead of DIAMOND, we got 9104 SCOs, similar total number, but 629 SCOs missing  
267 in DIAMOND result. 619 SCOs in DIAMOND also missed in BLASTP result vice versa. When  
268 using the annotated CDS sequences as the input of OrthoFinder with parameters ` -d -f cds `,  
269 9995 DNA SCOs were produced, much more than protein SCOs. Among these 9995 DNA SCOs,  
270 1736 and 1780 SCOs were absent in BLASTP and DIAMOND results, respectively. 844 and 880  
271 protein SCOs from BLASTP and DIAMOND, respectively, were also absent in DNA results.  
272 There were only 7785 SCOs present in all three results. BLAST should be top priority when  
273 computation resources were rich. To get the whole sequences from chromosomal level genome  
274 assemblies by ALiBaSeq, DNA SCOs as baits were also tested. It turned out that, more exons  
275 were recovered using DNA SCOs as the baits by ALiBaSeq. We didn't test the performance of  
276 DNA bait, which may be a worthwhile choice.

277 What if there are no close related genomes (more than three) available? Could we choose  
278 the pre-determined orthologous gene sets? OrthoDB v5 is a database that catalogs groups of  
279 orthologous genes in a hierarchical manner, from more general lineage to more fine-grained  
280 delineations (Kriventseva et al., 2019). We also test the performance of 1614 SCOs from  
281 embryophyta\_odb10 (inferred from 50 land plants genomes) by using the same workflow as the

282 9094 baits. The final species tree reconstructed by 709 SCOs from 1614 SCOs set was nearly the  
283 same with the tree reconstructed by 5648 SCOs from 9094 SCOs in this study, except the  
284 collection *C. ensifolium* RL0761 (Supplementary FIGURE 2). OrthoDB was another reliable  
285 resource to offer SCOs when there were no close related genomic annotation resources. Other  
286 SCOs set, like Angiosperms353 gene set (Johnson et al., 2018), or strictly/mostly single copy  
287 OGs used by MarkerMiner (Chamala et al., 2015; De Smet et al., 2013), should be also  
288 considered.

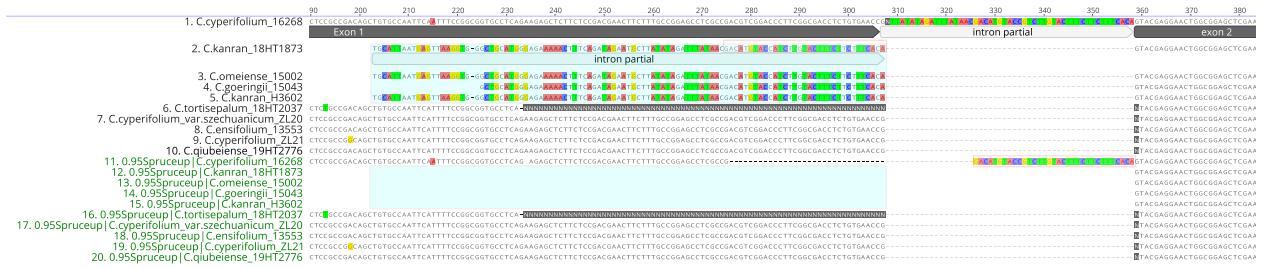
289

#### 290 **4.2 | Introns could create nonhomologous alignments**

291 The accuracy of phylogenetic reconstruction depends on the correct identification of  
292 homologous sites by sequence alignment. Only homologous alignments produced believable  
293 trees. The nucleotides of orthologous introns are difficult to align, especially the sample  
294 examined are relatively distant from each other (Creer, 2007; Sverdlov et al., 2005). Introns could  
295 create nonhomologous alignment, that is, intron residual sequences aligned with neighboring  
296 exon sequences. This phenomenon could be eased after filter by Spruceup, which could reduce  
297 the Shannon entropies of the alignments (FIGURE 5). And the results of Spruceup may still need  
298 to re-align to obtain the eventual refined alignments (Supplementary FIGURE 3). Our study also  
299 demonstrate that protein coding regions of SCOs are enough for high resolution species trees, and  
300 introns of SCOs are not necessary to keep.

301

302



303 **FIGURE 5.** Intron caused nonhomologous alignment could be relieved by Spruceup. The blue  
 304 shadows indicated the mis-aligned intron residual sequences mixed up with exon sequences. The  
 305 red border rectangle indicated the nucleotides that still needed to re-align after Spruceup filter.  
 306  
 307

#### 308 **4.3 | Much lower depth than 10 $\times$**

309 The numbers of SCOs recovered by HybPiper decrease dramatically when genomic  
 310 sequencing depth lower than 10 $\times$  with an average nucleotide coverage cutoff value of 5 (B. B.  
 311 Liu et al., 2021). This could due to the integrated assembling software SPAdes, which is designed  
 312 to assemble small genome like microorganism. By default, HybPiper performs per-sample/gene  
 313 assemblies using SPAdes with the parameter `--cov-cutoff 8` to generate less/short length contigs  
 314 with high base-level accuracy (Johnson et al., 2016). Lower the `--cov-cutoff` value to 5 still  
 315 screw up at coverage lower than 10 $\times$  (B. B. Liu et al., 2021). ALiBaSeq didn't assemble the reads  
 316 mapped to reference SCOs, ALiBaSeq hands whole genome assembling over professional  
 317 software designed to assemble complicated genomes regarding of large genome size and rich  
 318 repetitive elements. The actual depth of 25% subsampled *C. cyperifolium* 14942 could be less  
 319 then 3 $\times$  due to its extremely high PCR duplication rate (59.5%) (Table S1), but only 481 of 9060  
 320 SCOs failed to recovered (Figure 2A). Lower sequencing depth costs less money and relieves  
 321 computation burden too.

#### 322 **4.4 | Convenient, fast and convincing pipeline**

323 To achieve convincing SCOs matrices to reconstruct species tree, lots of software were  
324 investigated and compared. Unlike GenomeScope2 (Ranallo-Benavidez et al., 2020) or FindGSE  
325 (Sun et al., 2017), RESPECT only need  $0.5\times$  to  $4\times$  sequencing depth to estimate the genome sizes  
326 of samples (Sarmashghi et al., 2021). One can just gradually down-sample the genomic  
327 sequencing data to get relatively stable value calculated by RESPECT to determine genome size  
328 of sampled specie. We also recommend Megehit for its stable performance and less memory  
329 usage after comparing it with several other light whole genome assembling software, like  
330 SOAPdenovo2 (Luo et al., 2012), Minia3 (<https://github.com/GATB/minia>), SH-assembly (Shi &  
331 Yip, 2020). HybPiper could not directly extract SCOs from available genome assembly, but  
332 ALiBaSeq can retrieve SCOs from existing genome assembly whether annotations available or  
333 not. However, assembling whole genome needs huge computing resources. We could not run  
334 HybPiper v1.3 successfully on *Jensoa* dataset, but we test it on *Arabidopsis* (unpublished data).  
335 The results showed that ALiBaSeq performed much better than HybPiper when genome  
336 sequencing depth were lower than  $10\times$ , which was similar to the findings by previous research  
337 (B. B. Liu et al., 2021). However, HybPiper v2 released recently, its performance needs to re-  
338 evaluate. Another similar software, Easy353 (Zhang et al., 2022), is also worth investigating. At  
339 the step of alignment refining, Spruceup outperforms other popular software, like Gblocks  
340 (Castresana, 2000), trimAl (Capella-Gutiérrez et al., 2009), MACSE (Ranwez et al., 2018).  
341

#### 342 **4.5 | Kept most SCOs alignments with stringent percent identity**

343 A common rule of thumb is that two sequences are homologous if they are more than  
344 30% identical over their entire lengths (Pearson, 2013). Sequence identity of 60% was

345 recommended together with encoded proteins  $\geq 300$  amino acids when low-copy nuclear genes  
346 were chosen to conduct phylogenetic analyses (Zhang et al., 2012). To reconstruct the correct  
347 species tree by ASTRAL, SCOs should be kept as more as possible (Warnow, 2015). In our  
348 study, stringent identity of SCOs alignments were required. We found that about half of all  
349 recovered SCOs meet the standard of average pairwise sequence identity (APSI)  $\geq 80\%$ . We also  
350 tested using all SCOs with no percent identity filtering, and SCOs with APSI more than 90% and  
351 95%, topologies of species trees were nearly same, with LPP support value slightly down.

352

#### 353 4.6 | Perspectives

354 Organellar genomes are mostly inherited uniparentally, and rDNA genes have high copy  
355 number and are subject to incomplete homogenization. Only low copy orthologous nuclear genes  
356 provide a biparental record of the evolutionary history. More nuclear genes, including both genes  
357 with relatively slow and rapid evolutionary rates, should be used to accurately resolve  
358 relationships among close related species (Li et al., 2017; Zhang et al., 2012). Comparing to  
359 targeted sequencing, deep genome sequencing could promise large datasets of SCOs *in silico*  
360 without laborious baits synthesizing and complicated target enrichment. Predefined  
361 embryophyte\_odb10 with only 1614 SCOs derived from 50 genomes had showed sufficient  
362 resolution at lower taxonomic levels in this study as well as 9094 SCOs inferred from three  
363 *Cymbidium* genomes (Supplementary FIGURE 2). Are there SCOs serve as new universal  
364 barcodes in the whole plant kingdom like traditional standard barcode (Li et al., 2015) ?  
365 OrthoDB-like SCOs (USCOs, universal single-copy orthologs) which could be inferred from  
366 thousands of available genomes of deferent-level plant, may be a huge resource to screen easy-to-  
367 use barcodes applying to both high- and low- rank taxonomic hierarchies (Eberle et al., 2020).

368 More recently diverged species and more vouchers per species need to be addressed to exploit  
369 and validate the power of SCOs as the next generation of DNA barcodes. Additionally, numerous  
370 issues related to phylogenetics, molecular evolution and population genetics, would benefit  
371 greatly by resources of putative SCOs. Furthermore, the bioinformatic tools and computational  
372 resources continue to improve rapidly, we believe that SCOs will soon be prevalent in species  
373 identification, hybrid speciation, infra-species structure and other applications.

374

## 375 **AUTHOR CONTRIBUTIONS**

376 J.B.Y. and D.Z.L designed the study, Z.S.H collected, analyzed the data, and wrote the  
377 manuscript. All authors revised the manuscript.

378

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390

391 **CONFLICT OF INTEREST**

392 The authors declare no conflict of interest.

393

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