1Stepwise large genome assembly approach: A case of Siberian larch 2(*Larix sibirica* Ledeb.)

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28Abstract

29**Background**: *De novo* assembling of large genomes, such as in conifers (~12-30 Gbp) that also 30consist of ~80% of repetitive DNA is a very complex and computationally intense endeavor. One 31of the main problems in assembling such genomes lays in computing limitations of nucleotide 32sequence assembly programs (DNA assemblers). As a rule, modern assemblers are usually 33designed to assemble genomes with a length not exceeding the length of the human genome 34(3.24 Gbp). Most assemblers cannot handle the amount of input sequence data required to 35provide the sufficient coverage needed for a high-quality assembly.

36**Results**: An original stepwise method of *de novo* assembly by parts (sets), which allows to 37bypass the limitations of modern assemblers associated with a huge amount of data being 38processed is presented in this paper. The results of numerical assembling experiments conducted 39using the model plant *Arabidopsis thaliana*, *Prunus persica* (peach) and four most popular 40assemblers, ABySS, SOAPdenovo, SPAdes and CLC Assembly Cell, showed the validity and 41effectiveness of the proposed stepwise assembling method.

42**Conclusion**: Using the new stepwise *de novo* assembling method presented in the paper the 43genome of Siberian larch, *Larix sibirica* Ledeb. (12.34 Gbp) was completely assembled *de novo* 44by the CLC Assembly Cell assembler. It is the first genome assembly for larch species in 45addition to only five other conifer genomes sequenced and assembled for *Picea abies*, *Picea* 46*glauca*, *Pinus taeda*, *Pinus lambertiana* and *Pseudotsuga menziesii* var. *menziesii*. 47**Keywords:** *de novo* genome assembly, Siberian larch, *Larix sibirica*

48

49Background

50The *de novo* assembling of large genomes, such as in conifers, that have a length of 12 to 30 Gbp 51and consist of about 80% of highly repetitive elements (repeats), is a rather complex task [1-12]. 52The main problem of assembling such genomes is the limitations of assembler programs. As a 53rule, modern assemblers are designed to assemble genomes shorter or equal to the length of the

54human genome (3 Gbp). Most assemblers cannot handle the amount of input sequence data 55required to provide the coverage needed for a high-quality assembly or take too much time and 56computer resources. This prompts the development of new approaches in assembling large 57genomes, including Siberian larch (*Larix sibirica* Ledeb.), which together with Siberian stone 58pine (*Pinus sibirica* Du Tour) are the main objects of the genome project "Genomics of the key 59boreal forest conifer species and their major phytopathogens in the Russian Federation" funded 60by a research grant No. 14.Y26.31.0004 from the Government of the Russian Federation.

61

62Methods

63A stepwise approach to assembling large genomes

64High sequence coverage is always needed for high-quality *de novo* genome sequencing and 65assembly. For a given average genome coverage, the coverage of individual genome regions is 66approximately described by the Poisson distribution according to the Lander-Waterman theory 67[13]. Insufficient coverage increases the probability of zero coverage of some genome regions. 68Meanwhile, even a single coverage of genome regions is sufficient for their assembling using De 69Bruijn graph based methods [14] assuming no errors and repeats.

To solve the problem a new stepwise approach to assembling large genomes "in parts" was 71developed. The idea of partitioning data to perform assembly is not new. For example, in the 72article [15] it was proposed to apply a similar two-step hierarchical approach with the aim of 73improving the quality of assembly of bacterial genomes with very high coverage. However, the 74approach presented in [15] does not solve the problems of assembling large and super-large 75genomes, especially if DNA was obtained from diploid tissue.

In our case the assembly is also done in two steps. At the first step, the entire input pool of the 77sequence reads is divided into several sets (parts). The size of each set is within a limit for the 78number of reads that can be handled by the assembler program. Each set is assembled separately,

79then contigs obtained for each part are combined and used as the input data for the second step of 80assembling.

81 With this approach in the second step of assembling, the genome coverage by the input 82contigs no longer obeys the Poisson distribution. However, the level of coverage will not be 83greater than the number of parts by which the original pool of reads has been partitioned, which 84allows to bypass the limitation for the maximum amount of input data in the second step.

The challenge of the approach is the lower tolerance to sequencing errors and polymorphisms. 86The ambiguity in the input sequences in the second step could lead to generating duplications in 87the output. Therefore, the pipeline for the assembly with this approach should also include a 88verification of the assembly for redundancy to exclude potential duplicates. We used the 89UCLUST package [16] and self-blasting for this task.

It should be noted that not all assembly programs allow generating contigs with a coverage 91below the threshold value. To overcome this obstacle in the second step of the stepwise assembly 92either the program codes should be changed or software that does not have these limitations, 93such as the CLC Assembly Cell (QIAGEN, Hilden, Germany) should be used. This software 94takes into account possible sequencing errors during assembling. Thus, if there are sequencing 95errors in the input reads, most of them will not be incorporated in the contigs generated in the 96first step for each part of the pool. However, the problem of the stepwise assembling could be 97insufficient coverage for each part, which can lead to shorter contigs. Since there is a restriction 98on the minimum length of contigs in the assembling programs, such short contigs with 99insufficient lengths will be excluded from the assembly. Therefore, to reduce the probability of 100gaps due to excluding short contigs in the second step, one of the sets in the first step included all 101reads from the original data pool, but to make computing possible they were used as single end 102reads, and they were also multiplied. All steps are presented as a workflow chart in Fig. 1.

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104Testing of the proposed stepwise approach on the model plant species Arabidopsis thaliana

105To test the applicability of the proposed method of stepwise assembling for *de no*vo assembling 106of large genomes, such as in *L. sibirica* (12.03 Gbp), a genome assembly of the model plant 107species *Arabidopsis thaliana* obtained by the proposed method was compared with the standard 108*de novo* assembly of this species genome. A relatively small subset of *A. thaliana* genomic reads 109was selected to get a genome coverage comparable to *L. sibirica*.

110 As an additional argument supporting applicability of the method, the histograms of genome 111coverage obtained for *A. thaliana* and *L. sibirica* were compared for similarity. To construct the 112histograms, the genomic reads used for assembling were mapped to the assembled genomes 113using the bowtie software [17] for *A. thaliana* and the CLC read mapper for *L. sibirica*.

114 The *A. thaliana* genome contains 5 chromosomes and 135 Mbp [18]. We used the SPAdes 115[19], AbySS [20], CLC Assembly Cell (https://www.qiagenbioinformatics.com/products/clc-116assembly-cell), and SOAPdenovo [21] assemblers for the traditional *de novo* assembly of the *A*. 117*thaliana* genome. The genomic paired-end reads of *A. thaliana* were downloaded from the 118Genbank SRA database (accession number SRR492411 [22]). The results of assembly at the 119level of contigs by different assemblers are presented in Fig. 2 and Additional file 1.

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123 The result of assembling repetitive regions of the genome depends on the number and 124similarity of copies of a particular type of repeat. With a small and divergent number of copies, 125the assembler program, as a rule, is able to separate individual copies, so that all variants of this 126repeat will be presented in the final contigs. With a large number of identical or nearly identical 127copies of the same type, it would be difficult for an assembler to separate them. The number of 128repeats in the genome of *A. thaliana* represents quite a significant part - according to different 129estimates from 23 to 32 % [23, 24]. As a result, in the final assemblies, identical repeats of the

130same type can be represented by a single contig. This was reflected in the histogram of the contig131coverage based on the distribution of mapped reads used for assembling and presented in Fig. 3.132

133 <Fig. 3 location>

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135 It should also be noted that in the area of maximum coverage its distribution is more accu-

136rately described by the corrected Poisson distribution expressed by the formula $\frac{bL^{bx}e(-bL)}{\Gamma(bx+1)}$, 137where *L* - average coverage, *x* - coverage value, *b* - correction parameter (inversed value of ex-138tended variation) (Fig. 3, dotted line, *b* = 0.3).

The observed coverage histogram followed the Lander-Waterman theory in general, and the 140degree of coverage can be approximately described by the Poisson distribution for most of the 141genome with the left side maximum peak equaled to 16 reads (Fig. 3). The exact fitting of the 142coverage histogram to the Poisson distribution and the corrected (over-dispersed) Poisson 143distribution was estimated using an iterative maximum likelihood-based procedure implemented 144in the R statistical package. The results of these tests confirmed the fitting of the histogram to the 145over-dispersed Poisson distribution around the peak value, with the reservations about semi-146qualitative description of the distribution. The left and right tails of the distribution do not obey 147the provided model and should be described using another approaches. Because of this, the 148goodness of the fitting depends on selection of limits around a peak value of distribution. In 149reasonable limits between 0.5X and 2X of peak value, the match to over-dispersed Poisson 150distribution was significant based on the Kolmogorov-Smirnov (KS) test (*P* <0.01), but the 151estimated values of parameters should be anyway considered as approximate to avoid an excess 152of accuracy.

153 The clearly observed "heavy tail" in the right part of the distribution for contigs with high 154coverage (more than 100 reads) could be explained by the highly repetitive elements that

155represented different parts in the original genome, but were aligned and mapped together to the 156same single contigs. Therefore, the observed coverage histogram can be divided into two parts, 157with a coverage less or more than 100 reads, respectively. The key observation was that the 158observed coverage histogram for the *L. sibirica* genome followed the same trend that further 159confirms the applicability of the proposed method (respective larch data and figures are 160presented and discussed below in Results). The "heavy tails" were also observed in the coverage 161histograms in metagenomics [25] and medical DNA sequencing [26].

162 The number of copies of different types of repeats in the genome is governed by different 163evolutionary factors, and the simplest way to explain the heavy tail of the distribution is to use 164the Zipf's law to describe the frequencies of different types of repeats [27]. According to the 165Zipf's law, the frequencies of different types of repeats, sorted by the degree of occurrence, 166should be distributed in proportion to 1/n, where n is a consecutive number of the type of repeat 167in the list of observed types.

168 The number of repeats with a given degree of coverage can be expressed as the derivative of 169this dependence, that is, in proportion to $1/n^2$, where n is the degree of coverage. If the value of

 $Z = \frac{1}{\sqrt{Y}}$ is calculated for a coverage histogram same as in Fig. 3, where *Y* is the percentage of 171the genome with a given degree of coverage, then according to the Zipf's law, the value of *Z* 172should directly and proportionality depend on the degree of coverage. This dependence is 173demonstrated in Fig. 4 for the histogram of the observed coverage presented in Fig. 3.

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177 As it can be seen from Fig. 4, the Zipf's law is approximately satisfied for the coverage of 178more than 200 reads per site, which agrees with the abovementioned conclusion about 179assembling repeats that occurred with different frequency in the genome. For a more accurate

180description of the observed dependence, it is recommended to use a distribution based on the

181Zipf-Mandelbrot law formulated as $\frac{1}{n^k}$, where *k* is generally different from unity [27]. 182Nevertheless, the applicability of this law to genomic nucleotide sequences requires further 183study.

184 There are a few studies of the *A. thaliana* genome that identified different types of repeats, 185using, in particular, the method of clustering repeat sequences (for example, [23, 24]). According 186to these studies, while there was a general tendency to meet the Zipf's law for regions with a high 187degree coverage, individual peaks also appeared in the coverage distributions, such as in our case 188(Fig. 4), which can be interpreted as a manifestation of the similarity between individual types of 189repeats.

190 As shown in Fig. 3, the *A. thaliana* genome coverage was mostly described by a Poisson 191distribution with an average value of about 16 reads. To test the suggested stepwise assembling 192method, four sets were generated from the original pool of about 13 million reads. The first three 193sets included the first, second and third thirds of the original pool of reads, respectively. The 194fourth set also included one third of the original pool of reads, but was generated by random 195sampling from the mixed original pool of reads.

196 Thus, four sets of reads were generated from the original pool of reads used in the tests 197presented in Fig. 2 and Additional file 1. Fig. 5 and Additional file 2 presents the results of the 198stepwise assembly by four assemblers when each of the sets (parts) was assembled separately in 199the first step and then finally assembled by pooling all contigs from all four sets. It can be seen 200from the table that the CLC Assembly Cell demonstrated the best performance.

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202 <Fig. 5 location>

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Table 1 presents the results of assembly of each of the sets (parts) separately (the first step), as 205well as based on the pooling of contigs obtained respectively from two, three, and four sets 206(parts) using the CLC Assembly Cell software.

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<Table 1 location>

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Table 1 shows that insufficient coverage led to a significant decrease in the average contig 210length compared to the data in Fig. 2 and Additional file 1, but in the second step of assembling 211this parameter was corrected, and with the increase in the number of parts was stabilized at the 212level of values close to the values obtained by the different assemblers used to assemble the 213entire pool of reads simultaneously.

The identity of assembly obtained using parts and the stepwise method with assembly based 214 215on assembling simultaneously all reads was tested bv the NUCmer software 216(http://mummer.sourceforge.net), and the highest similarity was for alignments generated by the 217CLC Assembly Cell (90.14%) and Abyss (95.24%) software, respectively (Fig. 5 and Additional 218file 2), but the former software computed the assembly with the less number of contigs and more 219realistic total length, and seven times faster than the latter one with the same computer hardware 220resources (31 vs. 217 minutes, Fig. 5 and Additional file 2).

Fig. 6 compares the genome coverage histograms for the *A. thaliana* genome assembly based assembling the entire pool of reads simultaneously, such as in Fig. 3, and assembly based on exactly stepwise assembling in two steps of four parts (Table 1). It is clearly seen in Fig. 6 that the exactly stepwise assembled genome was adequately covered by the original set of reads.

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<Fig. 6 location>

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The ambiguous positions in the *A. thaliana* sequencing data were estimated by aligning229original *A. thaliana* reads to the assembly by Bowtie2. They represented 0.7% of genome size.

230The duplications of contigs were not detected in the final assembly, thus indicating a low level of 231ambiguity for the assemblyobtained by the suggested method.

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233The stepwise approach for the *Larix sibirica* genome assembly

234For the assembly of the *L. sibirica* genome, four PE and three MP libraries with different insert 235size were used (Fig. 7 and Additional file 3). At the first step, MPE libraries were decoupled and 236used as single reads to complete a pool of reads. The pool of reads was split to four parts and 237four sets of conigs were obtained, respectively. The CLC Assembly Cell software was selected 238for assembling the larch genome as the best performing software.

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Also, a fifth set of reads was added to the analysis. This set included all reads, but the PE and 243MPE reads were decoupled and used as single reads. This set was generated because we found 244experimentally that the CLC Assembly Cell assembler was able to process the entire volume of 245the *L. sibirica* sequence data, but only if the information about the length of the insertion was not 246indicated. In this case the "Optimization of the graph using paired reads" step is skipped. In this 247step long repeats are allowed, and scaffolding is not performed which turns out to be too much 248computationally intense and practically prohibitive for large volume data. Therefore, this set 249increased the representation of all reads, but they all could be used only as the single end reads at 250this step.

Unlike the inbred highly homozygous plant used for the genome sequencing and assembly, 252such as *A. thaliana*, the *L. sibirica* tree used for genome sequencing in our study represented a 253common forest tree with a relatively high level of individual heterozygosity and, respectively, 254high within individual biallelic variation. The number of ambiguous positions in the *L. sibirica* 255sequencing data was estimated at level 3.0% of genome size. The presence of duplicate contigs 256was detected in the preliminary draft assembly of *L. sibirica* obtained in the second step, thus 257revealing the higher data ambiguity in the *L. sibirica* sequencing data compared to the *A.* 258*thaliana* data. To resolve the ambiguities in the second stage, the total number of all contigs 259resulted from the fifth set was increased by 16 folds by multiplying each contig 16 times, 260respectively. This trick allowed the CLC assembler to apply the majority rule when picking one 261of the alternative alleles, using the alleles selected in the fifths set in the first step of assembly. 262The same approach was used also for the *Arabidopsis thaliana* genome stepwise assembly by 263four different assemblers (Fig. 8 and Additional file 4). The CLC Assembly Cell again 264demonstrated the best performance.

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266 <**Fig. 8** location>

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268In addition, to verify the accuracy of the stepwise CLC Assembly Cell assembly the medium size 269genome (265 Mb, 2n =16) of *Prunus persica* (peach) was also assembled by both the traditional 270method using 24324216 sequence reads (~15X coverage) available on 271<u>https://www.ncbi.nlm.nih.gov/bioproject/PRJNA31227</u> and the same stepwise approach that was 272used for the larch genome assembly and based on the five parts (Fig. 9 and Additional file 5). 273The traditional and stepwise assemblies were similar by 95.64% based on the NUCMER 274comparison.

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<Fig. 9 location>

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277**Results**

278Stepwise assembly of the Larix sibirica genome in parts

279The length of the *L. sibirica* genome is about 12.03 Gb [28], and about 82 % of which consists of 280repeats [6-8]. The volume of the larch sequencing data obtained (11 billion paired 100 bp long 281reads) was hardly manageable by the available genome assemblers and more than twice the 282maximum amount of data that the best performing software in our test with the *Arabidopsis* data

283CLC Assembly Cell can handle. Therefore, we developed a new stepwise assembly method for 284assembling this and other large genomes and demonstrated its consistency in computer 285experiments on assembling the model plant *A. thaliana* genome.

The original Siberian larch sequencing data were partitioned into five sets following mainly 287the procedure described for *A. thaliana* in Methods with an additional fifth set. Each set was 288separately assembled by the CLC Assembly Cell program. The assembly results are presented in 289Table 2 for each set. Only contigs with a minimum length of 200 bp were included in the final 290assembly.

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<Table 2 location>

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Total length of contigs assembled separately for each of the five sets varied from ~2.5 to ~6 294Gb. The N50 parameter varied from ~300 to ~1300 bp. At the second step, individual assemblies 295were combined by specifying them as unpaired reads and changing the *k*-mer parameter length 296from 35 to 60. In addition, the mate pair (MP) reads generated from the MP libraries with 2000-29710,000 bp long inserts were added to the CLC Assembly Cell input data. These reads were used 298at the stage of scaffolding (joining contigs into scaffolds with gaps of known expected length).

Additional scaffolding was done using BESST [29], and 228,571 additional scaffolds were 300generated. The scaffolding was also improved by using larch transcriptome reads and RaScaf + 301Bowtie2 software [30]. About 92% reads were mapped to the genome assembly and allowed us 302to connect 3,622 contigs into scaffolds. The assembly was finished with gap-closing using the 303Sealer program implemented in the last part of the Abyss pipeline [31], and 61,037 gaps were 304closed.

305 Thus, the contigs of the all five assemblies were processed and the obtained statistics 306presented in Table 2.

307 Adding the last two assemblies based on the 4 and 5 sets (Table 2) improved the final 308parameters (Table 3), and increased the total contig and scaffold lengths from 7.18 to 7.99 Gb

309and from 11.04 to 12.34 Gb, respectively. The N50 parameter remained unchanged compared to 310the best values of partial assemblies. This is inconsistent with the results for *A. thaliana* 311assembly tests, but could be explained by the additional scaffolding procedure with the MP reads 312for the *L. sibirica* assembly.

- 313 <Table 3 location>
- 314

The assembly was tested for redundancy using a custom pipeline specially developed for this 316task, which checks for duplication taking into account possible erroneous nucleotide 317substitutions and indels. As a result, 74851 scaffolds were excluded. The assembly was 318additionally checked for vector contamination and redundancy using the UniVec database 319(https://www.ncbi.nlm.nih.gov/tools/vecscreen/univec) and the BLAST program, and as a result, 32010681 sequences were deleted.

321 Finally, after scaffolding, a complete Siberian larch genome of 12.34 Gb was assembled *de* 322*novo*. The computing time taken to assemble larch genome using 40 cores is presented in Fig. 10 323and Additional file 6. In total it took about 529 hours or 22 days. Therefore, the larch genome 324computing using the next best assembler SOAPdenovo could predictively take more than 100 325days.

326

<Fig. 10 location>

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The histogram of the coverage for the obtained genome corresponded to the Poisson distribu-329tion with extended variation in the regions with low coverage (Fig. 11A) and to the Zipf's law in 330the region of high coverage (Fig. 11B) and was similar to the one obtained for *A. thaliana* (Fig. 3313). The values for the inversed over-dispersion parameter were nearly the same for both genomes $332(0.3 \pm 0.1)$, as it was confirmed by likelihood-based parameter estimates.

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<**Fig. 11** location>

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The correlation presented in Fig. 11B was completely linear in the region of sufficient 337coverage, as expected from the Zipf's law, in contrast to the correlation for *A. thaliana*, in which 338many individual peaks were observed (Fig. 4). This is consistent with the results of the analysis 339of genomic repeats in Norway spruce [1], where it was difficult to cluster repeats and separate 340some types of repeats, as it can be done for many other genome sequences of eukaryotes. It also 341followed from our results that distribution of repeats is continuous in conifers. The presence of a 342large number of repeats and associated with them discontinuities in assembling can explain the 343smaller average contig length in comparison with the results of the *A. thaliana* genome 344assembling.

The accuracy of the stepwise CLC Assembly Cell assembly was also verified by assembling 346the medium size genome (265 Mb, 2n =16) of *Prunus persica* (peach) using both methods. The 347assembly parameters are presented in Table 3 and the histogram of the coverage in Fig. 12. Both 348observed and expected distributions of the peach genome coverage were similar to those for 349*Arabidopsis* (Figs 2, 3, 5) and Siberian larch (Fig. 11) genomes.

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The negative binomial distribution or the over-dispersed Poisson distribution is often used to 354describe genome coverage histograms, but, to our best knowledge, the effect of overdispersion 355was not systematically studied in the context of genome assemblies (but see [25, 32]). However, 356the similar values of the over-dispersion parameter for three assembled genomes confirmed by 357the KS tests could serve as an additional argument that the proposed method could be adequately 358scaled to the assembly of large genomes.

359

360Discussion

361Testing of the proposed stepwise approach for assembling genomes in parts on the model plant 362species *A. thaliana* showed that despite some deterioration of the distribution parameters of the 363contig lengths in the final assembly compared to normal assembling using the CLC Assembly 364Cell, the result of the stepwise assembling was comparable with the results of assembling all data 365simultaneously using different assemblers. Comparison of the lengths of the obtained genomes 366and histograms of the coverage obtained by different methods also allows us to state that the 367stepwise assembling by parts allows obtaining a consistent and reliable genome assembly 368corresponding to the original biological material.

The analysis of the coverage histograms carried out for *A. thaliana, Prunus persica* (peach) 370and larch showed a tendency to satisfy the Zipf's law for the frequency of repeats and provided 371additional grounds for concluding that the stepwise assembly approach by parts is applicable for 372assembling large genomes, such as the Siberian larch genome. The interpretation of the coverage 373histograms using the Zipf's law made it possible also to clarify the idea of the statistical 374regularities characterizing the evolutionary mechanisms of multiplication of repeats in different 375plant species.

376

377Conclusion

378Using the new stepwise *de novo* assembling method presented in the paper the genome of 379Siberian larch, *Larix sibirica* Ledeb. (12.34 Gbp) was for the first time completely assembled *de* 380*novo* by the CLC Assembly Cell assembler. It is the first genome assembly for any larch species 381in addition to only five other conifer genomes sequenced and assembled for *Picea abies* [1], 382*Picea glauca* [2], *Pinus taeda* [3-5, 9, 11], *Pinus lambertiana* [10] and *Pseudotsuga menziesii* 383var. *menziesii* [12]. Presented approach makes assembling feasible for very large genomes with a 384reasonable computing time and without engaging huge computing resources. The assemblies 385produced by this approach are still of reasonable quality allowing their annotation and further 386use.

387

388 Declarations

389

390Abbreviations

391Gb: Giga Base; bp: base pair; HPC: High Performance Computing

392

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396

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402Availability of data and materials

403This manuscript describes published software and a new developed pipeline (source code is 404available from the authors on request). The sequence reads and obtained scaffolds are publicly 405available under the NCBI Genbank BioProject accession number PRJNA393226.

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407Authors' contributions

408KVK & DAK conceived and developed original idea. SIF, VVS, ANC, SVM & YAP wrote the 409codes, developed the pipeline and implemented parallelization. YAP and NVO processed original 410sequencing reads and generated the original data. DAK, VVS, SIF & KVK together wrote the 411first draft manuscript. All authors read, revised and approved the final manuscript.

413Competing interests

414The authors declare that they have no competing interests.

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419Ethics approval and consent to participate

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540**Table 1** Results of the *Arabidopsis thaliana* genome stepwise assembling in four sets (parts)

Assembly part	Total length, bp		Contigs	
		N50, bp	number	mean length, bp
1 ^a	101200000	1586	110067	919
2 ^a	101200000	1601	109903	920
3ª	101.2	1595	110119	919
4 ^b	101.2	1586	110384	916
1+2	113.2	3225	64543	1753
1+2+3	116.6	3861	60606	1923
1+2+3+4	113.7	4325	52576	2161

541 using the CLC Assembly Cell software

^aRepresents approximately 1/3 of all original reads; ^bRepresents also approximately 1/3 of all original reads, but randomly selected

542Table 2 The assembly results of the five sets generated from the original Larix sibirica genome

543sequencing data

Set	Number of contigs	N50, bp	Maximum length, bp	Total length, Gbp
1	7870837	310	56157	2.566
2	5469129	535	65362	2.549
3	5449065	1383	157662	4.319
4	4677717	1092	91349	3.117
5	13244672	475	46203	5.937

544Table 3 The final stepwise Larix sibirica genome assembly based on five sets and the MP reads

Assembly*	Number, mln	N50, bp	Maximum length, bp	Total length, Gbp
Contigs	12.40	1074	128642	7.99
Scaffolds	11.33	6443	354326	12.34

545*Minimum contig length used for assembling was 200 bp.

547Additional files

Additional file 1: Table S1: The results of the traditional *de novo Arabidopsis thaliana* genome 549assembly generated by four different assemblers

Additional file 2: Table S2: Results of the *Arabidopsis thaliana* genome stepwise assembly by 551different assemblers using raw reads partitioned into four sets

Additional file 3: Table: S3 Sequencing libraries and generated sequence data used for the *Larix* 553*sibirica* genome assembly

Additional file 4: Table S4 Results of the *Arabidopsis thaliana* genome stepwise assembly by 555four different assemblers using raw reads partitioned into five sets following approach used for 556assembling of the *Larix sibirica* genome

Additional file 5: Table S5 The traditional and stepwise CLC Assembly Cell genome assembly 558parameters for peach (*Prunus persica*)

Additional file 6: Table S6 The computing time taken to assemble each set and the complete 560*Larix sibirica* genome using 40 cores

563Figure titles (max 15 words) and legends (max 300 words)

565Fig. 1 Stepwise assembly workflow chart

Fig. 2 The results of the traditional *de novo Arabidopsis thaliana* genome assembly generated by 568four different assemblers. Minimum contig length used for assembling was 200 bp 569

Fig. 3 Histogram of the *Arabidopsis thaliana* genome coverage by the mapped reads used for the 571genome assembly generated by the CLC Assembly Cell software (solid line). Expected and 572corrected Poisson distributions are represented by dashed and dotted lines, respectively. The 573number of reads (degree of the genome coverage) is on the horizontal axis; the logarithmic 574proportion of the genome with such degree of coverage is on the vertical axis 575

Fig. 4 Dependence of the transformed value of the fraction of the genome coverage Z on the 577level of coverage. Solid line represents linear dependency calculated by the least square fit 578

Fig. 5 Results of the *Arabidopsis thaliana* genome stepwise assembly by different assemblers 580using raw reads partitioned into four sets. Minimum contig length used for assembling was 200 581bp

Fig. 6 Comparison of the *Arabidopsis thaliana* genome coverage histograms obtained for the 584genome assembly assembled by the CLC Assembly Cell using all reads simultaneously (solid 585line) and the stepwise method with two steps and four parts (dotted line) 586

Fig. 7 Sequence coverage for seven sequencing libraries used for the *Larix sibirica* genome 588assembly

Fig. 8 Results of the *Arabidopsis thaliana* genome stepwise assembly by four different 591assemblers using raw reads partitioned into five sets following approach used for assembling of 592the Larix sibirica genome. Minimum contig length used for assembling was 200 bp 593

Fig. 9 The traditional and stepwise CLC Assembly Cell genome assembly parameters for peach 595(*Prunus persica*). Minimum contig length used for assembling was 200 bp. 596

Fig. 10 The computing time (number of hours) taken to assemble each set and the complete 598*Larix sibirica* genome using 40 cores 599

Fig. 11 A: Observed distribution of Siberian larch genome coverage (solid line) and expected 601from the corrected Poisson distribution (dotted line) with average coverage value equaled 7 and 602correction parameter b=0.3. **B**: Dependence of the transformed degree of genome coverage Z on 603the Siberian larch genome coverage (solid line). The dashed line represents linear dependency 604calculated by the least square fit and fully coincides with the solid line. 605

Fig. 12 A: Observed distribution of *Prunus persica* (peach) genome coverage (solid line) and 607expected from the corrected Poisson distribution (dotted line) with average coverage value 608equaled 15 and correction parameter b=0.3. **B**: Dependence of the transformed degree of genome 609coverage Z on the peach genome coverage (solid line). The dashed line represents linear 610dependency calculated by the least square fit and fully coincides with the solid line.