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Title

Hybrid *de novo* assembly of the draft genome of the freshwater mussel *Venustaconcha ellipsiformis* (Bivalvia: Unionida).

Authors

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29 **Abstract**

30 Freshwater mussels (Bivalvia: Unionida) serve an important role as aquatic ecosystem
31 engineers but are one of the most critically imperilled groups of animals. An
32 assembled and annotated genome for freshwater mussels has the potential to be
33 utilized as a valuable resource for many researchers given their ecological value and
34 threatened status. In addition, a sequenced genome will help to answer more
35 fundamental questions of sex-determination and genome evolution in bivalves
36 exhibiting a unique “doubly uniparental inheritance” mode of mitochondrial DNA
37 transmission through comparative genomics approaches. Here, we used a combination
38 of sequencing strategies to assemble and annotate a draft genome of the freshwater
39 mussel *Venustaconcha ellipsiformis*. The genome described here was obtained by
40 combining high coverage short reads (65X genome coverage of Illumina paired-end
41 and 11X genome coverage of mate-pairs sequences) with low coverage Pacific
42 Biosciences long reads (0.3X genome coverage). Briefly, the final scaffold assembly
43 accounted for a total size of 1.54Gb (366,926 scaffolds, N50 = 6.5Kb, with 2.3% of
44 "N" nucleotides), representing 93% of the predicted genome size of 1.66Gb. Over one
45 third of the genome (37.5%) consisted of repeated elements and more than 85% of the
46 core eukaryotic genes were recovered. Finally, we reassembled the full mitochondrial
47 genome and found six polymorphic sites with respect to the previously published
48 reference. This resource opens the way to comparative genomics studies to identify
49 genes related to the unique adaptations of freshwater mussels and their distinctive
50 mitochondrial inheritance mechanism.

51 **Keywords:** genome assembly, annotation, High Throughput Sequencing, Freshwater
52 Mussels, Unionida

53 **Introduction**

54 Through their water filtration action, freshwater mussels (Bivalvia: Unionida) serve
55 important roles as aquatic ecosystem engineers (Gutiérrez et al. 2003; Spooner &
56 Vaughn 2006), and can greatly influence species composition (Aldridge et al. 2007).
57 From a biological standpoint, they are also well known for producing obligate parasitic
58 larvae that metamorphose on freshwater fishes (Lopes-Lima et al. 2014), for being
59 slow-growing and long-lived, with several species reaching >30 years old and some
60 species >100 years old (see Haag & Rypel 2011 for a review), and for exhibiting an
61 unusual system of mitochondrial transmission called Doubly Uniparental Inheritance
62 or DUI (see Breton et al. 2007; Passamonti & Ghiselli 2009; Zouros 2013) for
63 reviews). From an economic perspective, freshwater mussels are also exploited to
64 produce cultured pearls (Haag 2012). Regrettably however, habitat loss and
65 degradation, overexploitation, pollution, loss of fish hosts, introduction of non-native
66 species, and climate change have resulted in massive freshwater mussel decline in the
67 last decades (reviewed in Lopes-Lima et al. 2017; 2018). For example, more than 70%
68 of the ~300 North American species are considered endangered at some level (Lopes-
69 Lima et al. 2017).

70

71 While efforts are currently underway to sequence and assemble the genome of
72 the marine mussel *Mytilus galloprovincialis* (Murgarella et al. 2016), genomic
73 resources for mussels in general are still extremely scarce. In addition to *M.*
74 *galloprovincialis*, the genomes of two other marine mytilid mussel species, i.e. the
75 deep-sea vent/seep mussel *Bathymodiolus platifrons* and the shallow-water mussel

76 *Modiolus philippinarum* have recently been published (Sun et al. 2017). In all cases,
77 genomes have proven challenging to assemble due to their large size (~1.6 to 2.4Gb)
78 and widespread presence of repeated elements (~30% of the genome, and up to 62% of
79 the genome for the shallow-water mussel *Modiolus philippinarum*, Sun et al. 2017).
80 For example, the *Mytilus* genome remains highly fragmented, with only 15% of the
81 gene content estimated to be complete (Murgarella et al. 2016). With respect to
82 freshwater mussels (order Unionida), no nuclear genome draft currently exists. An
83 assembled and annotated genome for freshwater mussels has the potential to be
84 utilized as a valuable resource for many researchers given the biological value and
85 threatened features of these animals. Such studies are needed to help identifying genes
86 essential for survival (and/or the genetic mechanisms that led to decline) and
87 ultimately for developing monitoring tools for endangered biodiversity and plan
88 sustainable recoveries (Pavey et al. 2016; Savolainen et al. 2013). In addition, a
89 sequenced genome will help answer more fundamental questions of sex-determination
90 (Breton et al. 2011; 2017) and genome evolution through comparative genomics
91 approaches (e.g. Sun et al. 2017).

92

93 Given the challenges in assembling a reference genome for saltwater mussels
94 (Sun et al. 2017; Murgarella et al. 2016), we used a combination of different
95 sequencing strategies (Illumina paired-end and mate pair libraries, Pacific Biosciences
96 long reads, and a recently assembled reference transcriptome (Capt et al. 2018) to
97 assemble the first genome draft in the family Unionidae. Hybrid sequencing
98 technologies using long read–low coverage and short read–high coverage offer an

99 affordable strategy with the advantage of assembling repeated regions of the genome
100 (for which short reads are ineffective) and circumventing the relatively higher error
101 rate of long reads (Koren et al. 2012; Miller et al. 2017). Here, we present a *de novo*
102 assembly and annotation of the genome of the freshwater mussel *Venustaconcha*
103 *ellipsiformis*.

104

105 **Methods**

106 To determine the expected sequencing effort to assemble the *Venustaconcha*
107 *ellipsiformis* genome, i.e., the necessary software and computing resources required,
108 we first searched for C-values from other related mussel species. C-values indicate the
109 amount of DNA (in picograms) contained within a haploid nucleus and is roughly
110 equivalent to genome size in megabases. Two closely related freshwater mussel
111 species (*Elliptio* sp., c-value = 3; *Unio* sp., c-value = 3.2), in addition to two
112 other well studied mussel groups (*Mytilus* spp., c-value = 1.3-2.1; *Dreissena*
113 *polymorpha*, c-value = 1.7) were identified on the Animal Genome Size Database
114 (<http://www.genomesize.com>). As such, we estimated the *Venustaconcha* genome size
115 to be around ~1.5-3.0Gb, and this originally served as a coarse guide to determine the
116 sequencing effort required, given that when the sequencing for *Venustaconcha* was
117 originally planned, no mussel genome had yet been published.

118

119 ***Mussel specimen sampling, genomic DNA extraction and library preparation***

120 Adult specimens of *Venustaconcha ellipsiformis* were collected from Straight River
121 (Minnesota, USA; Lat 44.006509, Long -93.290899) and sexed by microscopic

122 examination of gonad smears. Gills were dissected from a single female individual and
123 genomic DNA was extracted using a Qiagen DNeasy Blood & Tissue Kit (QIAGEN
124 Inc., Valencia, CA, USA) using the animal tissue protocol. The quality and quantity of
125 DNA, respectively, were assessed by electrophoresis on 1% agarose gel and with a
126 BioDrop mLITE spectrophotometer (a total of 15 µg of DNA was quantified using the
127 spectrophotometer). For whole genome shotgun sequencing and draft genome
128 assembly, we used two sequencing platforms: Illumina (San Diego, CA) HiSeq2000
129 and Pacific Biosciences (Menlo Park, CA) PacBio RSII. First, three paired-end
130 libraries with insert size of 300b were constructed using Illumina TruSeq DNA Sample
131 Prep Kit. One mate pair library with insert sizes of about 5Kb was constructed for
132 scaffolding process using Illumina Nextera mate-pair library construction protocol. For
133 high-quality genome assembly, Pacific Biosciences system was employed for final
134 scaffolding process using long reads. Pacific Biosciences long reads (>10Kb) were
135 generated using SMRT bell library preparation protocol (ten SMRT cells were
136 sequenced). Construction of sequencing libraries and sequencing analyses were
137 performed at the Genome Quebec Innovation Centre (McGill University, Qc, Canada).
138

139 *Pre-processing of sequencing reads*

140 We quality trimmed paired-end and mate-pair reads using TRIMMOMATIC 0.32 (Bolger
141 et al. 2014) with the options ILLUMINACLIP:TruSeq3-PE.fa:2:30:10 LEADING:3
142 TRAILING:3 SLIDINGWINDOW:6:10 MINLEN:36. This allowed removal of base
143 pairs below a threshold Phred score of three at the leading and trailing end, in addition
144 to removing base pairs based on a sliding window calculation of quality (minimum

145 Phred score of ten over six base pairs). Finally, if trimmed reads fell below a threshold
146 length (36b), both sequencing pairs were removed. We verified visually the quality
147 (including contamination with Illumina paired-end adaptors) before and after trimming
148 using FASTQC (Andrews 2010). This allowed us to only keep high quality reads prior to
149 the assembly steps.

150

151 Following quality trimming, we used BFC (Li & Durbin 2009) to perform error
152 correction for the Illumina paired-end sequencing data. BFC suppresses systematic
153 sequencing errors, which helps to improve the base accuracy of the assembly and
154 reduce the complexity of the *de Bruijn* graph based assembly, described below.

155

156 Corrected paired-end reads were subsequently used to identify the optimal K
157 value that provides the most distinct genomic k-mers using KMERGENIE v1.7016
158 (Chikhi & Medvedev 2014). We tested k = 10 to 100, in incremental steps of 10, and
159 we then refined the interval from 20 to 40, in incremental steps of 2 to get a more
160 precise estimate of K. Based on the best K value (k=42), KmerGenie was also used to
161 estimate genome size.

162

163 ***Genome assembly strategy***

164 We used ABYSS 2.0 (Jackman et al. 2017), a modern genome assembler specifically
165 built for large genomes and reads acquired by different sequencing strategies. ABYSS
166 2.0 works similarly to ABYSS (Simpson et al. 2009), by using a distributed *de Bruijn*
167 graph representation of the genome, therefore allowing parallel computation of the

168 assembly algorithm across a network of computers. In addition, the software makes
169 use of long sequencing reads (Illumina mate-pair libraries and Pacific BioSciences
170 long reads) to bridge gaps and scaffold contigs. Yet, as memory requirements and
171 computing time scale up exponentially with genome size, for large genomes (>1Gb),
172 these rapidly become very large (>100GB of RAM) and unpractical. Consequently,
173 Jackman et al. (2017) introduced ABYSS 2.0, which employs a probabilistic data
174 structure called a Bloom filter (Bloom 1970) to store a *de Bruijn* graph representation
175 of the genome and, consequently, greatly reduces memory requirements and
176 computing time. The Bloom filter allows removing from memory the majority of
177 nearly identical k-mers likely caused by sequencing errors, as k-mers with an
178 occurrence count below a user-specified threshold are discarded. The caveat is that it
179 can generate false positive extension of contigs, but through optimization, this can be
180 kept well below 5%, and in fact, false positives can be corrected later on in the
181 assembly step (Jackman et al. 2017).

182

183 In the current study, we combined different types of high throughput
184 sequencing to aid in assembling the genome (**Table 1**). ABYSS 2.0 (Jackman et al.
185 2017) performs a first genome assembly step without using the paired-end information,
186 by extending unitigs until either they cannot be unambiguously extended or come to an
187 end due to a lack of coverage (*uncorrected unitigs*). This first *de Bruijn* graph
188 representation of the genome is further cleaned of vertices and edges created by
189 sequencing errors (*unitigs*). Paired-end information is then used to resolve ambiguities
190 and merge *contigs*. Following this, mate-pairs are mapped onto the assembly to create

191 *scaffolds*, and finally long reads (Pacific Biosciences long reads) and the
192 *Venustaconcha* reference transcriptome from Capt et al. (2018) were also mapped onto
193 the assembly to create *long-scaffolds*. This reference transcriptome was assembled
194 from a pool of sequences coming from four different male and female individuals and
195 further details are provided in Capt et al. (2018). Although ideally sequencing
196 information would all come from a single individual, the current study design did not
197 allow for this. In addition, given that coding sequences are conserved compared to
198 non-coding regions, it remains highly valuable to use a transcriptome in a *de novo*
199 genome assembly.

200

201 We ran the ABYSS 2.0 assembly stage (abyss-bloom-dbg) with a k-mer size of
202 41 (ABYSS requires an odd number k-mer), a Bloom filter size of 24GB, 4 hash
203 functions and a threshold of k-mer occurrence set at 3. These parameters were chosen
204 after performing several test assemblies, in order to minimize the false positive rate
205 (<5%), maximize the N50 of the assembly and keep the virtual memory (95GB) and
206 CPU (24 CPUs) requirements within a reasonable computational limit for our
207 resources. In addition, we adjusted parameters at the mapping stage to create contigs,
208 scaffolds and long-scaffolds to maximize N50 (overlap required in re-alignments,
209 distance between mate-pairs, nb reads aligned to support assembly, see pipeline
210 available at https://github.com/seb951/venustaconcha_ellipsiformis_genome).

211

212 Genome completeness was assessed using BUSCO 3.0.2 (Benchmarking
213 Universal Single-Copy Orthologs, Simao et al. 2015). Briefly, BUSCO uses curated lists

214 of known core single copy orthologs to produce evolutionarily-informed quantitative
215 measures of genome completeness (Simao et al. 2015). Here, we tested both the
216 eukaryotic (303 single copy orthologs) and metazoan (978 single copy orthologs) gene
217 lists to assess the completeness of our genome assembly.

218

219 *Characterization of repetitive elements*

220 Given that repetitive elements can occupy large proportions of a genome, the
221 characterization of their proportion and composition is an essential step during genome
222 annotation. RepeatModeler open-1.0.10 (Smit & Hubley 2015) was used to create an
223 annotated library of repetitive elements contained in the *Venustaconcha* genome
224 assembly (excluding sequences <1Kb). Then, with RepeatMasker open-4.0.7 (Smit et
225 al. 2015), we extracted libraries of repetitive elements for the taxa “Bivalvia” and
226 “Mollusca” from the RepeatMasker combined database (comprising the databases
227 Dfam_consensus-20170127 and RepBase-20170127) using built-in tools. Sequences
228 classified as “artefact” were removed from the last two libraries before the subsequent
229 steps. The three libraries were used alone and/or in combination (except for the
230 Mollusca+Bivalvia combination) to mask the cut-down assembly again with
231 RepeatMasker, specifying the following options: -nolow (to avoid masking low
232 complexity sequences, which may enhance subsequent exon annotation), -gccalc (to
233 calculate the overall GC percentage of the input assembly), -excln (to exclude runs of
234 ≥ 20 Ns in the assembly sequences from the masking percentage calculations). Option -
235 species was used to specify the taxon for the runs with Bivalvia and Mollusca libraries,
236 while option -lib used to specify the *Venustaconcha* library and the combined ones.

237 Results summaries for the latter three runs were refined with the RepeatMasker built-in
238 tools. Linear model fit for genome size and repeats content for all available bivalve
239 genomes were calculated with R version 3.1.0 (R Core Team 2012), using the highest
240 masking value found for *Venustaconcha* .

241

242 ***Genome annotation***

243 We used QUAST (Gurevich et al. 2013) to calculate summary statistics on the genome
244 assembly. In addition, QUAST uses a Hidden Markov Model to identify putative genes
245 in the final assembly (GLIMMERHMM Majoros et al. 2004). Following this, we
246 translated Open Reading Frames identified in the annotation files into protein
247 sequences using BEDTOOLS v2.27.1 (Quinlan & Hall 2010) and EMBOSS TRANSEQ
248 v6.6.0 (Rice et al. 2000) bioinformatics pipelines. These were then compared against
249 the manually curated UniProt database (556,388 reference proteins, downloaded
250 January 11th 2018, e-value cut-off of 10^{-5}) using BLASTp (Altschul et al. 1990). These
251 steps were done on the long-scaffolds assembly, the masked long-scaffolds assembly
252 (with low complexity regions replaced with N), in addition to the broken long-
253 scaffolds assembly (scaffolds broken into smaller contigs by QUAST, based on long
254 stretches of N nucleotides).

255

256 ***Mitochondrial genome***

257 Given the rare mode of mitochondrial inheritance of freshwater mussels and therefore
258 its evolutionary importance, we first aimed to check if the mitochondrial female
259 genome had been properly assembled. Using BLASTn (Altschul et al. 1990) with high

260 stringency (E value $<1e-50$), we identified a fragmented mitochondrial genome. We
261 then created a mt specific dataset containing 1,396,004 sequence reads by aligning
262 paired-end reads to the reference mt genome of Breton et al. (2009) (GenBank Acc.
263 No. FJ809753) using SAMTOOLS v1.3.1 and BEDTOOLS v2.27.1 (Li et al. 2009; Quinlan
264 & Hall 2010). We then rebuilt the mt genome *de novo* using ABYSS 2.0, testing
265 different k-mers (17-45). In addition, we aligned reads to the reference transcriptome
266 using BWA v0.7.12-R1039 (H Li & Durbin 2009) and identified Single Nucleotide
267 Polymorphisms (SNPs) with respect to the reference mt genome using SAMTOOLS and
268 BCFTOOLS v1.3.1 (Li et al. 2009).
269

270

271 **Results and Discussion**

272 We generated 564M paired-end reads (2 X 100b) representing an average 65X
273 coverage of the genome (**Table 1**). This was complemented by 98M mate-pairs (5Kb
274 insert, 11X average genome coverage) and 103,000 Pacific Biosciences long reads
275 (0.3X average genome coverage), and a recently published reference transcriptome
276 comprised of 285,000 contigs (Capt et al. 2018). Filtering and trimming the raw
277 paired-end and mate-pair sequences removed about 5% of the total base pairs from
278 further analyses, indicating that the quality of the raw sequences was high (**Table 1**).
279 K-mer analysis indicated that the number of unique k-mers peaked at 42 and predicted
280 a genome assembly size of 1.66Gb (**Figure 1**), smaller than predicted genome size
281 according to C-value for other Unionida, but in general agreement with the recent draft
282 genome of the marine mussel *Mytilus galloprovincialis* (1.6Gb) and the deep-sea
283 vent/seep mussel (*Bathymodiolus platifrons*, 1.64Gb).

284

285 Running the ABySS 2.0 assembly stage (abyss-bloom-dbg) led to a low False
286 Positive Rate (<0.05%). The N50 for the contig assembly was 3.2Kb with 551,875
287 contigs (discarding contigs <1Kb, given that small contigs likely represent artefacts
288 and provide little information for the overall genome assembly (Pavey et al. 2016;
289 Murgarella et al. 2016, see **Table 2**). Once these were corrected and paired-end, mate-
290 pairs and long read information were added, the scaffolds N50 increased to 5.5Kb,
291 with 2.3% of nucleotides represented as “N” (see **Table 2** for the summary statistics
292 and **Table 3** for overall genome assembly statistics acquired from QUAST analysis).
293 Adding the Pacific Biosciences long reads only slightly improved the scaffolds N50

294 (from 5.5 to 5.7Kb, **Table 2**) and slightly decreased the number of *long-scaffolds*
295 >1Kb (from 423,853 to 410,237), likely because our long read coverage was quite low
296 (0.3X, **Table 1**). In addition, it is also possible that the more error prone Pacific
297 Biosciences sequences, compared to Illumina paired-end reads, reduced their usability
298 (Miller et al. 2017). Once the reference transcriptome was added, it improved the N50
299 to 6.5Kb, and substantially decreased the number of long-scaffolds to 366,926. This
300 final long-scaffold assembly accounted for a total size of 1.54Gb (with 2.3% of "N"
301 nucleotides) and represented 93% of the predicted genome size of 1.66Gb. Yet, it
302 remained highly fragmented (366,926 scaffolds, **Table 2**). Genome annotation
303 statistics can also be viewed in html format and downloaded here:
304 [https://github.com/seb951/venustaconcha_ellipsiformis_genome/tree/master/annotatio](https://github.com/seb951/venustaconcha_ellipsiformis_genome/tree/master/annotation_quast_v3)
305 [n_quast_v3](https://github.com/seb951/venustaconcha_ellipsiformis_genome/tree/master/annotation_quast_v3)

306

307 While assembly numbers (N50, number of scaffolds, etc.) are not directly
308 comparable with other recently published genomes given the diversity of sequencing
309 approaches (Illumina, 454, Sanger, PacBio), library types, sequencing depth and
310 unique nature of the genome themselves, they can give a broad perspective of the
311 inherent difficulties of assembling large genomes. The best comparison is probably
312 with the saltwater mussel, *Mytilus galloprovincialis*, giving their similar genome size
313 (1.6Gb for *Mytilus* vs 1.66Gb for *Venustaconcha*) and Illumina paired-end sequencing
314 approaches (32X for *Mytilus* vs 65X for *Venustaconcha*). While the *Mytilus* genome
315 project (Murgarella et al. 2016) did not utilize mate-pair libraries or Pacific Bioscience
316 long reads, they did make use of sequencing libraries with varying insert sizes (180,

317 500 and 800b). As such, they obtained a genome assembly quality relatively similar to
318 ours and consisting of 393 thousand scaffolds (>1Kb), with however a substantially
319 lower N50 (2.6Kb compared to 6.5Kb for *Venustaconcha*). The recently reported
320 genome for the deep-sea vent/seep mussel *Bathymodiolus platifrons* (1.64Gb) made
321 use of nine Illumina sequencing libraries with varying insert sizes (180 to 16Kb) and
322 an overall coverage of >300X. With this very thorough sequencing approach, the
323 scaffold N50 obtained was substantially higher (343.4Kb), but again the genome
324 remained highly fragmented, into >65 thousands scaffolds. As exemplified here, high
325 coverage sequencing libraries with varying insert sizes have become a broadly used
326 approach for large and complex genomes. In fact, it is implemented by default in many
327 genome assembly platforms (e.g. Soapdenovo2, Luo et al. 2012, ALLPATHS-LG, Gnerre
328 et al. 2011). In the future, these libraries will likely be useful to further assemble the
329 *Venustaconcha* genome, at least until these approaches are superseded by affordable,
330 error free, single molecule long read sequencing (Gordon et al. 2016; Badouin 2017)
331 or mapping approaches that allow reaching chromosome level assemblies such as
332 optical mapping (e.g. Bionano Genomics, San Diego, CA).

333

334 Results of the BUSCO (Simao et al. 2015) analyses showed that 664 (68%) of
335 the 978 core metazoan genes (CEGs) were considered complete in our assembly.
336 When the BUSCO analysis was extended to include also fragmented matches, 871
337 (89%) proteins aligned. Results were similar when compared against the 303 core
338 eukaryotic genes (61% complete, 86% complete or fragmented, **Table 4**). When
339 compared to the previously published reference transcriptome for *Venustaconcha*

340 *ellipsiformis* (Capt *et al.* 2018), we found fewer complete genes, but also fewer
341 duplicated genes (97.5% complete, and 24% duplicated in the reference transcriptome,
342 compared to 68.1% complete and 1% duplicated here). This likely reflects the fact that
343 the reference transcriptome is nearly complete, while the current reference genome is
344 still fragmented. However, the reference transcriptome also likely contains multiple
345 isoforms of the same genes, in addition to possible nematode contaminating sequences,
346 despite the authors' best efforts to minimize these problems. Previously analysed
347 molluscan genomes of similar size (Murgarella *et al.* 2016; Sun *et al.* 2017) have found
348 that 16% (*Mytilus galloprovincialis*, 1.6Gb), 25% (pearl oyster *Pinctada fucata*,
349 1.15Gb), 36% (California sea hare *Aplysia californica*, 1.8Gb) of the core eukaryotic
350 genes were complete. For their part Sun and collaborators (2017), identified 96% of
351 the core metazoan genes to be partial or complete in the deep-sea vent/seep mussel
352 *Bathymodiolus platifrons* (1.6Gb), again reflecting that the depth and type of
353 sequencing, in addition to the idiosyncrasies of each genome, can have considerable
354 influence on the end results.

355

356 The custom *Venustaconcha* repeat library created *de novo* with RepeatModeler
357 contained 2,068 families, the majority of them (1,498, 72.44% of the total) classified
358 as “unknown”. The genome masking performed with the Bivalvia and Mollusca
359 libraries had scarce performances (masking 2.38% and 2.59%, respectively; details in
360 **Supplementary Table RM1**), possibly because of the phylogenetic distance between
361 *V. ellipsiformis*, which belongs to the early-branching bivalve lineage of
362 Palaeoheterodonta, and the other bivalve and mollusk species represented in the

363 database as well as their relative number of sequences. The custom *Venustaconcha*
364 library masked 37.17% of the genome, while the combined *Venustaconcha*+Bivalvia
365 masked 37.69% of the genome and the *Venustaconcha*+Mollusca reached 37.81%, the
366 highest masking percentage (**Supplementary Table RM2**). After refining, these raw
367 values slightly decreased to respectively 36.29%, 36.80%, and 36.91%
368 (**Supplementary Table RM3**). All these latter values of repeat content fall in the 32-
369 39% range (the median for all species is 37%) where six out of the nine sequenced
370 bivalve species lie, irrespective of their genome size (*M. philippinarum* and *R.*
371 *philippinarum* are the furthest from this interval) (**Table 5** and **Supplementary Figure**
372 **1**). Although the number of species sequenced up to now is still low, this observation
373 indicates that repetitive elements may contribute differently to the total genome size
374 among the different bivalve taxa: indeed, the correlation between genome size and
375 repeats content is weak (**Supplementary Figure 1**). In both the *ab initio* masking with
376 the *Venustaconcha* library and the two combined ones, most of the identified repeats
377 are categorized as “unknown” (22.8% of the assembly), followed by retroelements
378 (LINEs 2.9%, LTR elements 2.3-2.4%, and SINEs 1.7%, for a total of 6.9% of the
379 assembly) and DNA elements (5.4-5.6% of the assembly) (**Supplementary Table**
380 **RM3**). Direct comparisons of these values with other species should be performed
381 with caution, as the usually large “unclassified” portion of repeats might contain
382 species-specific variants of known elements (Murgarella et al. 2016) that may
383 therefore change the relative weight of each category on the total.
384

385 QUAST was used to calculate summary statistics and identify putative genes in
386 the final assembly using a hidden markov model (**Table 3**). Following this, 29,031;
387 14,195 and 25,544 Open Reading Frames were annotated using BLASTp against
388 UniProt database in the long-scaffolds, broken and masked long-scaffolds assemblies,
389 respectively.

390

391 Freshwater mussels, marine mussels, as well as marine clams are the only
392 known exception in the animal kingdom with respect to the maternal inheritance of
393 mitochondrial DNA (see Breton et al. 2007 for a review). Their unique system,
394 characterized by the presence of two gender-associated mitochondrial DNA lineages,
395 has therefore attracted studies to better understand mitochondrial inheritance and the
396 evolution of mtDNA in general. Using BLASTN, we recovered 53 contigs matching to
397 the 15,975b female reference mt genome from Breton *et al.* (2009), indicating that the
398 mt genome was highly fragmented and likely improperly assembled with our current
399 approach, much like what was found in the *Mytilus galloprovincialis* genome draft of
400 Murgarella (Murgarella et al. 2016). As such, we created a dataset of mt specific
401 sequences that could be aligned to the mt genome (1,396,004 reads). This mt specific
402 dataset was then re-assembled *de novo*, using different k-mers (17-45). Using a k-mer
403 similar or larger to the one used in the overall assembly ($k \geq 41$) resulted in a failed
404 assembly (no contigs created, data not shown), while using a k-mer < 21 generated a
405 highly fragmented mt genome (data not shown). Using a k-mer between 21 and 39
406 generated one large contig of 16,024b comprising the entire mitogenome, with a 42b
407 insertion in the 16S ribosomal RNA. Given the different rate of evolution of mtDNAs,

408 it is likely that assembly parameters we used for the whole genome were not
409 appropriate for the *V. ellipsiformis* female mt genome. Finally, we also re-aligned the
410 mt specific dataset to the original mt genome of Breton et al. (2009) and found high
411 coverage (mean = 7,256X, SD = 682) for most positions, while for three regions
412 coverage dropped below 300X (**Figure 2**). Six SNPs with respect to the reference were
413 also identified, indicating possible polymorphism, or sequencing error in the original
414 mt reference genome (**Figure 2**).

415

416 **Conclusion**

417 High throughput sequencing has the power to produce draft genomes that were only
418 reserved to model systems ten years ago. Here we report the first *de novo* draft
419 assembly of the *Venustaconcha ellipsiformis* genome, a freshwater mussel from the
420 bivalve order Unionida. Our assembly covers over 93% of the genome and contains
421 nearly 90% of the core eukaryotic orthologs, indicating that it is nearly complete.
422 However, as for other mussel genomes recently published, our genome remains
423 fragmented, showing the limits of high throughput sequencing and the necessity to
424 combine different sequencing approaches to augment the scaffolding and overall
425 genome quality, especially when a large fraction of the genome is comprised of
426 repetitive elements. In the future, the *Venustaconcha* genome will benefit from a larger
427 number of long read sequences, varying library size for paired-end sequencing, and the
428 use of genetic, physical or optimal maps to subsequently order scaffolded contigs into
429 pseudomolecules or chromosomes.

430

431 **Abbreviations**

432 BLAST: Basic Local Alignment Search Tool

433 b: base pairs

434 Kb: Kilobases

435 M: Million

436 Gb: Gigabases

437 GB: gigabytes

438 CPU: Central Processing Unit

439 DNA: Deoxyribonucleic acid

440 LINEs: Long interspersed elements

441 LTR: Long terminal repeats

442 ORF: Open Reading Frames

443 N80/50/20: weighted median statistic such that 80/50/20% of the entire assembly is

444 contained in contigs/scaffolds equal to or larger than this value.

445 L50 = minimum number of sequences required to represent 50% of the entire assembly

446 RAM: Random Access Memory

447 SINEs: Short interspersed elements

448

449 **Data availability**

450 Supporting data for this Genome Report will be made available on datadryad.org

451 Raw sequences are available in the SRA database with number SRP132483

452 (submission SUB3624229 to be release upon publication) and Bioproject accession

453 PRJNA433387. All scripts used in the analyses are available on github

454 (https://github.com/seb951/venustaconcha_ellipsiformis_genome).

455

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Table 1: DNA sequencing strategy.

Type	Raw		Raw		Trimmed reads (%)			SRA		
	Insert size (bp)	Read Length (bp)	No. Reads (paired)	Total length (mb)	No. Reads (paired)	Total length (mb)	Total length (% raw)	read length	coverage	accession
Paired-end	300	2X100	189,876,842	37,975	185,721,156	36,274	95.5	97.6		
Paired-end	300	2X100	195,394,768	39,079	191,002,987	37,319,	95.5	97.7		
Paired-end	300	2X100	178,820,287	35,764	174,954,230	34,224	95.6	98.9		
Total			564,091,897	112,818	551,678,373	107,818	95.6	98.1	65X	

Mate pair	5000	2X100	97,801,148	19,560	94,350,168	18,717	95.7	99.3	11X
Pacific Bioscience Long reads assembled transcriptome		4,406.4 (average)	103,096	454					0.27X
	1,170.9 (average)		285,260	334					
	301-50,048 (min-max)								

584

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586

587 **Table 2:** Assembly statistics (ABYSS2.0).

588

assembly	n (x10e6)	n:1000	L50	min	N80	N50	N20	max	sum (x10e6)
raw unitigs	39.8	347,879	101,624	1,000	1,361	2,181	3,891	25,883	707
unitigs	18.5	444,734	127,617	1,000	1,485	2,452	4,273	25,944	984
contigs	14.0	551,875	141,012	1,000	1,704	3,117	5,817	39,408	1,449
scaffolds	13.7	423,853	92,607	1,000	2,303	5,477	9,099	45,260	1,539
long scaffolds (PacBio)	13.7	410,237	86,661	1,000	2,391	5,708	9,893	47,610	1,548
long scaffolds (PacBio + transcriptome)	13.6	366,926	58,906	1,000	2,534	6,523	16,660	298,135	1,549

589

590 n = number of contigs, n:1,000 = number of contigs of minimum length of 1,000, L50 = minimum number of sequences required to
 591 represent 50% of the entire assembly, min = minimum length of sequences analysed, N80, N50, N20 = weighted median statistic such
 592 that 80/50/20% of the entire assembly is contained in contigs equal to or larger than this value, max = maximum size of contig, sum =
 593 sum of all contigs of size > min, assembly stage (*raw unitigs* = raw assembly, not taking into account paired-end information, *unitigs*
 594 = filtering, merging and popping bubbles in *De Bruijn* graph, *contigs* = unitigs with paired-end information mapped, *scaffolds* =
 595 contigs with mate-pairs information mapped, *long scaffolds* = scaffolds with PacBio / transcriptome information integrated).

596

597

598 **Table 3:** Assembly and annotation statistics for the long scaffold assembly.

599

QUAST Assembly statistics	long_scaffolds	long_scaffolds (> 1kb scaffolds broken based on N streches)	long_scaffolds (> 1kb scaffolds, masked assembly)
Number of contigs (>= 0 b)	13,635,758	821,266	374,245
Number of contigs (>= 1 kb)	371,706	549,364	374,245
Number of contigs (>= 5 kb)	94,238	50,209	95,019
Number of contigs (>= 10 kb)	26,952	5,151	27,030
Number of contigs (>= 25 kb)	5,073	23	4,976
Number of contigs (>= 50 kb)	1,456	0	1,427
Total length (>= 0 b)	2,638,723,663	1,554,026,338	1,596,234,060
Total length (>= 1kb)	1,590,292,198	1,425,294,273	1,596,234,060
Total length (>= 5 kb)	1,000,983,904	360,423,103	1,003,000,325
Total length (>= 10 kb)	541,545,133	64,766,821	538,648,016
Total length (>= 25 kb)	231,252,884	687,249	226,147,564
Total length (>= 50 kb)	107,178,666	0	104,739,660
Number of contigs	371,706	821,266	37,4245
Largest contig	313,274	44,597	31,3274
Total length	1,590,292,198	1,554,026,338	1,596,234,060

Estimated reference length	1,660,000,000	1,660,000,000	1,660,000,000
GC (%)	34.19	34.19	33.49
N50	6,656	2,812	6,627
number of N's per 100 kb	2,293.33	13.17	39,200.22
number of predicted genes (unique)	201,068	277,765	123,457
number of predicted genes (>= 300 b)	74,820	82,359	41,697
number of predicted genes (>= 1.500 kb)	18,539	14,338	11,897
number of predicted genes (>= 3 kb)	6,511	3,289	4,375
number of annotated ORF (uniprot)	29,031	14,198	25,544

600

601 All statistics are based on contigs of size ≥ 1 kb, unless otherwise noted (e.g., "# contigs ≥ 0 b)" and "Total length ≥ 0 b)"

602 include all contigs.).

603

604

605 **Table 4:** Analysis of genome completeness using BUSCO 3.0.2 (Benchmarking Universal Single-Copy Orthologs, (Simao et al. 2015)).

606

	metazoa	eukaryota
Complete orthologs (C)	664 (68%)	185 (61%)
Complete and single-copy orthologs (S)	652 (67%)	181 (60%)
Complete and duplicated orthologs (D)	12 (1%)	4 (1%)
Fragmented orthologs (F)	207 (21%)	76 (25%)
Missing orthologs (M)	107 (11%)	42 (14%)
Total ortholog groups searched	978	303

607

608

609 **Table 5:** Gene size and repeat elements

Subclass	Order	Family	Species	Estimated genome size (Gb)	% of repeated elements
Palaeoheterodonta	Unionida	Unionidae	<i>Venustaconcha ellipsiformis</i>	1.66	37.81
Heterodonta	Veneroida	Veneridae	<i>Ruditapes philippinarum</i>	1.37	26.38
Pteriomorpha	Mytiloida	Mytilidae	<i>Bathymodiolus platifrons</i>	1.64	47.90
			<i>Modiolus philippinarum</i>	2.38	62.00
			<i>Mytilus galloprovincialis</i>	1.60	36.13
	Ostreoida	Ostreidae	<i>Crassostrea gigas</i>	0.55	36.00
		Pectinidae	<i>Chlamys farreri</i>	0.95	32.10
			<i>Patinopecten yessoensis</i>	1.43	38.87
Pterioda	Pteriidae	<i>Pinctada fucata</i>	1.15	37.00	
Pteriomorpha mean (s.d.)				1.39 (0.58)	41.43 (10.29)
	Mytiloida mean (s.d.)			1.87 (0.44)	48.68 (12.95)
	Ostreoida mean (s.d.)			0.98 (0.44)	35.66 (3.40)
		Pectinidae mean (s.d.)		1.19 (0.34)	35.49 (4.79)
all subclasses mean (s.d.)				1.41 (0.51)	39.35 (10.23)

610 Estimates of genome size and percentage of repeated elements in the currently available bivalve nuclear genomes. Data for each single
 611 species was retrieved from the literature (Takeuchi et al. 2012; Zhang et al. 2012; Murgarella et al. 2016; Mun et al. 2017; Yuli Li et
 612 al. 2017; Wang et al. 2017). The genome size for *V. ellipsiformis* was based on k-mer analysis (see methods and Fig. 1). Mean and

613 standard deviation (s.d.) values are also shown for the taxa comprising more than one species and for all subclasses, i.e. the class

614 Bivalvia.

615

616

617

618 **Figure Legends**

619

620 **Figure 1:** KmerGenie report for best k + predicted genome size.

621

622 **Figure 2:** Mitochondrial coverage based on sequence alignment and annotation (from NCBI). Six nucleotide positions were identified
623 in the legend as fixed for an alternative allele compared to the reference of Breton et al. (2009).

624

Predicted best k = 42

Predicted assembly size = 1.657 Gb



