1 Teaching transposon classification as a means to crowd source the

2 curation of repeat annotation – a tardigrade perspective

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- 81 genome assembly
- 82

83 Abstract

84 The advancement of sequencing technologies results in the rapid release of hundreds of new 85 genome assemblies a year providing unprecedented resources for the study of genome 86 evolution. Within this context, the significance of in-depth analyses of repetitive elements, 87 transposable elements (TEs) in particular, is increasingly recognized in understanding genome 88 evolution. Despite the plethora of available bioinformatic tools for identifying and annotating 89 TEs, the phylogenetic distance of the target species from a curated and classified database of 90 repetitive element sequences constrains any automated annotation effort. Manual curation of 91 raw repeat libraries is deemed essential due to the frequent incompleteness of automatically 92 generated consensus sequences. However, manual curation and classification are time-93 consuming processes that offer limited short-term academic rewards and are typically 94 confined to a few research groups where methods are taught through hands-on experience. 95 Crowd sourcing efforts could offer a significant opportunity to bridge the gap between 96 learning the methods of curation effectively and empowering the scientific community with 97 high-quality, reusable repeat libraries. Here, we present an example of such crowd sourcing 98 effort developed through both in-person and online courses built around a collaborative peer-99 reviewed teaching process that can be used as teaching reference guide for similar projects. 100 The collaborative manual curation of TEs from two tardigrade species, for which there were 101 no TE libraries available, resulted in the successful characterization of hundreds of new and 102 diverse TEs: A hidden treasure awaits discovery within non-model organisms.

Background

104 The importance of in-depth analyses of repetitive elements, particularly transposable elements 105 (TEs), is becoming more and more fundamental to understand genome evolution and the 106 genetic basis of adaptation [1]. While there is a wealth of bioinformatic tools available for the

107 identification and annotation of TEs (<u>https://tehub.org/en/resources/repeat_tools</u>), any 108 automated annotation effort is limited by the phylogenetic distance of the target species to a 109 database of curated and classified repetitive element sequences [2]. For example, in birds 110 where zebra finch and chicken have well-characterized repetitive elements because their 111 genomes were first sequenced in large consortia during the pre-genomics era [3,4], automated 112 annotation of other bird genomes will render most repeats as correctly classified [5,6]. On the 113 other hand, in taxa as diverse and divergent as insects, up to 85% of repetitive sequences can 114 remain of "unknown" classification in non-Drosophila species [7]. This is problematic. 115 Inferences about the mobility and accumulation of TEs, as well as their potential effects on 116 the host, are not feasible for unclassified repeats, as well as for incorrectly classified repeats if 117 the automated classification is based on short, spurious nucleotide sequence similarity [8,9].

118 The reference bias in TE classification reflects the history of the TE field in the genomics era: 119 In the 1990s and 2000s, there were usually multiple people tasked with TE identification, 120 classification, and annotation for each genome project, yielding manually curated consensus 121 sequences (namely representative sequences which quality was controlled and improved) and 122 fully classified TE libraries deposited in databases such as Repbase [2]. Over the last ten years, 123 however, the number of genome projects both of individual labs as well as large consortia has 124 increased exponentially and so have speed and number of automated TE annotation efforts 125 [10–12], while time and personnel have remained limited for curated TE annotation efforts. 126 Similar to taxonomic expertise required for identifying and classifying organisms, TE 127 identification and classification need hands-on experience with manual curation for months or 128 even years per genome [1] which is usually taught through knowledge passed within genome 129 projects and research groups. Recent efforts [13–15] have started to make manual curation 130 accessible to a broader scientific audience, with the aim to increase reproducibility and 131 comparability. However, what cannot be changed is that there are hundreds if not thousands

of genomes per TE-interested researcher with more or less pressing priority for time-consuming manual curation.

134 Low scalability and people power are major obstacles that need to be overcome by the many 135 facets of computational biology where curation is essential. Annotation efforts of other 136 genomic features have shown that crowd sourcing through teaching [16–22], or "course 137 sourcing" as we call it, has the benefit of providing participants with hands-on skills for 138 curation and experience on how to reconcile biology with technical limitations, while 139 simultaneously sharing the workload of time-consuming curation across multiple people 140 working on different parts at the same time. Thus, we argue that a TE curation effort that 141 would take months or years for a single person may fit into a few days or weeks of teaching, 142 of course as long as reproducibility and comparability are ensured throughout course duration.

143 Here, we present our "course sourcing" experience from two iterations of a Physalia Course 144 on TE identification, classification, and annotation. We focused on two species of tardigrades 145 as a case study to motivate student-centered learning through direct contribution to scientific 146 knowledge: Tardigrades are, to our knowledge, the most high-ranking animal phylum without 147 curated TE annotation, very clearly illustrated by the fact that in previous genome analyses, 148 almost all repeats remained of "unknown" classification [23]. Tardigrades are a diverse group 149 of aquatic and terrestrial animals which show extraordinary ability to survive extreme 150 environments by entering the state of cryptobiosis [24]. This animal clade comprises almost 151 1,200 described species belonging to Panarthropoda [25] and the two species used in the 152 courses are closely related and belong to the Hypsibildae family [23].

The first course took place in person in June 2018 in Berlin across five full-time work days: The first three days familiarized the 13 participants with the biology of TEs, concepts for classification, and methods for annotation using the tardigrade *Hypsibius dujardini*, while the last two days had a student-centered learning format where each participant was able to 157 deepen knowledge where needed and curate as many TEs as possible from the target species. 158 The second course took place virtually in June 2021 due to the Covid-19 pandemic and 159 comprised five afternoons in the Berlin time zone to minimize Zoom fatigue. The overall 160 format was similar to the prior in-person course but with 24 participants and focusing on 161 another tardigrade, *Ramazottius varieornatus*, which the participants identified to have not a 162 single shared TE family with the tardigrade *H. dujardini* curated in the 2018 course. Between 163 the two courses, the participants were able to uncover a vast diversity of TEs and successfully 164 curate almost 500 consensus sequences. We demonstrate therefore that a collaborative 165 approach is a valuable means to achieve significant results for the scientific community and 166 we hope to share with the community a teaching reference for future similar efforts, because: 167 A hidden treasure always awaits discovery in non-model organisms.

168 Results and Discussion

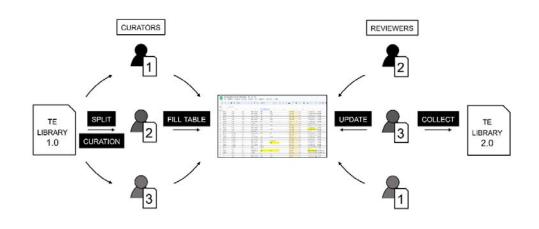
169 Incorporating crowd sourcing efforts within a classroom setting ("course sourcing") can 170 represent an invaluable opportunity for teaching, while simultaneously contributing to the 171 scientific community. However, course sourcing also presents its own unique challenges, particularly in terms of minimizing errors, maximizing reproducibility and student 172 173 engagement. Drawing from our experience in both in-person and virtual settings, we 174 identified several crucial factors in teaching TE manual curation that must be considered 175 during the organization and supervision of such course, like: a) establishing a standardized 176 approach for curation and classification of TE consensus sequences; b) implementing a peer-177 review process between participants to check on the quality of the curated consensus sequence; 178 c) maintaining meticulous version control of the libraries. Here, we describe how we 179 addressed these points. First, to establish a standard approach to manual curation, we 180 implemented methods widely used in the TE community that have been recently reviewed in 181 detail [13,14]. The approach, briefly, consists in producing and inspecting multi-sequence

182 alignments for each of the consensus sequences automatically generated by RepeatModeler 183 [10]. Each nucleotide position of the "alignable part" of the alignment is carefully inspected to 184 identify the correct termini of the TE while correcting for any ambiguous base or gap. To 185 correct for ambiguous bases, we applied a majority rule and assigned the most representative 186 IUPAC nucleotide character for each position in the alignment (see Methods). To correct the 187 consensus sequences where gaps of different lengths are present, we considered each 188 insertion/deletion length as independent events so that a majority rule was applicable to these 189 regions as well. When very complex regions could not be unambiguously solved, stretches of 190 10 N nucleotides were inserted as placeholder (gap) in the consensus sequence. The TE 191 classification followed the nomenclature used by RepeatMasker to ensure direct compatibility 192 with the tool and its suite of scripts for downstream analysis. Second, when participants 193 completed the curation of their consensus sequences, then their results would go through a 194 peer-review process where both the quality of the sequence and its classification were revised 195 by other participants (or course faculty). During the in-person edition, a random set of 196 consensus sequences curated by one participant was assigned to another participant, while in 197 the second online edition, all sequences were reviewed by the two instructors and one 198 participant (Figure 1). The review of the TE sequences continued after the official conclusion 199 of the course. To ensure reproducibility and the documentation of the entire decision-making 200 process for classification, all steps and details of classification were recorded in a shared 201 Google Sheet. The tables would include the changes in consensus sequence names, names of 202 the curators and reviewers and additional comment (Figure 1, Table S1). Whenever a change 203 was introduced in a consensus sequence (either in the nucleotide sequence itself or in the 204 classification), the new version was directly added to the multi-sequence alignment file used 205 for curation together with the original one. Keeping all the versions of a consensus in the 206 same alignment file and respective notes in the tables allows the implementation of a basic 207 version control useful to check on the steps leading to a particular decision. From the re-

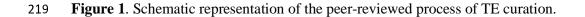
208 iteration of the course, we noticed three particularly challenging points for beginners that need 209 an extra supervision effort. The most challenging points are the identification of the correct 210 termini, target site duplications (a hallmark of transposition for the vast majority of TEs) if 211 any, and the correct spelling of the TE categories for classification in accordance with the 212 RepeatMasker nomenclature rules. The last point is of particular importance especially if the 213 repeat annotation is visualized as a landscape using the RepeatMasker scripts (e.g., 214 calcDivergence.pl and createRepeatLanscape.pl) to not cause computing errors and 215 downstream misinterpretations.

216 Finally, all the tutorials to obtain and curate a TE library are available on the GitHub

217 repository linked to this paper: <u>https://github.com/ValentinaPeona/TardigraTE</u>.



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220 Improvement of the transposable element libraries

To generate the TE libraries, we first ran RepeatModeler and RepeatModeler2 on both species and obtained 489 and 900 consensus sequences for *H. dujardini* and *R. varieornatus* respectively (**Table 1**). Then the course participants manually curated as many consensus sequences as possible. In about three course days plus voluntary efforts by some participants after each course, the participants were able to curate 286 consensus sequences (58%) of the

226 *H. dujardini* library and 145 consensus sequences (16%) of the *R. varieornatus* library (**Table**) 227 **S1-3**). Given the lack of previously curated libraries from closely related species, most of the 228 consensus sequences were automatically classified as "unknown" by RepeatModeler, but the 229 thorough process of manual curation successfully reclassified 305 unknown consensus 230 sequences (out of a total of 431 curated sequences, 71%) into known categories of elements. 231 After manual curation, we found that most of the two species' libraries are comprised of DNA 232 transposons and a minority of retrotransposons (Table 1). Since many consensus sequences 233 remained uncurated and unclassified, it is possible that the relative percentages of the 234 categories change in the future, but we expect, especially from the composition of the H. 235 dujardini library, to mostly find additional (non-autonomous) DNA transposons among the 236 unclassified.

237

Table 1: Overview of classification of tardigrade repeats in the curated libraries. The librarieshere described contain both curated and uncurated consensus sequences.

Species	DNA	LINE	LTR	SINE	Unknown
Hypsibius dujardini	247	12	29	2	199
Ramazzottius varieornatus	203	35	11	-	651

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The process of manual curation improved the overall level of TE classification of the libraries but also the quality of the individual consensus sequences by correctly identifying their termini and in general by extending their sequence. Indeed, by comparing the lengths of the consensus sequences for the same element, we can notice a marked increase in length after curation (**Figure 2**).

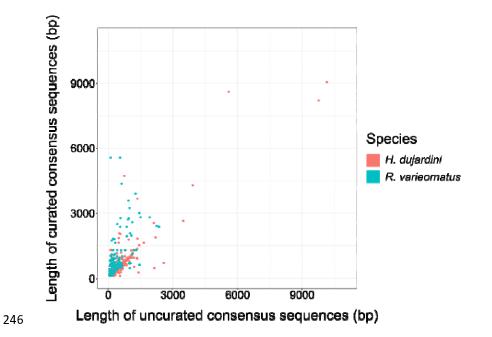


Figure 2. Comparison of the length of the consensus sequences before and after manualcuration.

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250 Diversity of transposable elements

251 When looking at the diversity of repeats in the partially curated libraries (libraries comprising 252 both curated and uncurated consensus sequences), we identified a total of 419 Class II DNA 253 consensus sequences belonging to the superfamilies/clades CMC, MULE, TcMar, Sola, 254 PiggyBac, Tc4, PIF-Harbinger, Zator, hAT, Maverick, and P. Many of these elements are 255 non-autonomous and show a remarkable diversity of internal structures (Figure 3). For Class 256 I retrotransposons, we found 40 LINEs belonging to the superfamilies/clades CR1, CRE, R2, 257 R2-NesL, L2, RTE-X and RTE-BovB and other 35 LTRs belonging to the 258 superfamilies/clades DIRS, Gypsy, Ngaro and Pao.

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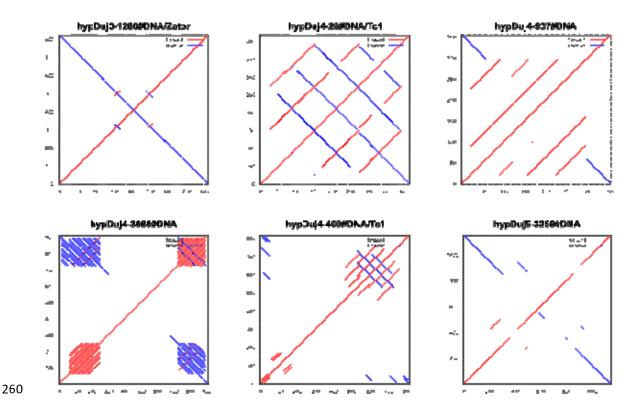


Figure 3. Dotplots of six DNA transposons from the library of *Hypsibius dujardini* produced with the MAFFT online server. These elements were selected by course participants for aesthetic reasons.

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265 To highlight the importance of generating and using custom repeat libraries for the organisms 266 of interest as well as their curation, we masked the two tardigrade genomes and compared 267 how the annotation and accumulation patterns change when using general repeat libraries (in 268 this case the Repbase library for Arthropoda) and species-specific ones before and after 269 curation (Figure 4, Table 2). The use of the known repeats for Arthropoda available on 270 Repbase provided a poor and insufficient annotation for both species (all the following 271 percentages are given for *H. dujardini* and then for *R. varieornatus*) where only 1.95% and 272 0.26% of the assemblies were annotated as interspersed repeats and the accumulation patterns 273 were characterized only by likely old insertions. Then the use of species-specific, albeit 274 uncurated, libraries completely changed the percentage of TEs annotated (16.38% and 275 15.66%) and their accumulation patterns that showed many recently accumulated insertions.

276 While the shape and percentages of the repeat landscapes did not drastically change after the 277 manual curation of the libraries, the curated libraries clearly highlighted a large accumulation 278 of DNA transposons in recent and ancient times alike that were either not present in the other 279 landscapes or were hidden among the "unknown" repeats. Especially for *R. varieornatus*, the 280 curation highlighted a higher accumulation of repeats in the very recent times (1-5% of 281 divergence). This higher accumulation of DNA transposons in recent times is also in line with 282 the finding of multiple putatively active transposable element subfamilies (**Table 3**). Finally, 283 the use of the repeat library of one species to annotate the other species (reciprocal masking) 284 resulted to be almost as insufficient as the use of the Repbase library for Arthropoda stressing 285 once again how important it is to have a capillary knowledge of the repeatome for correct 286 biological interpretations.

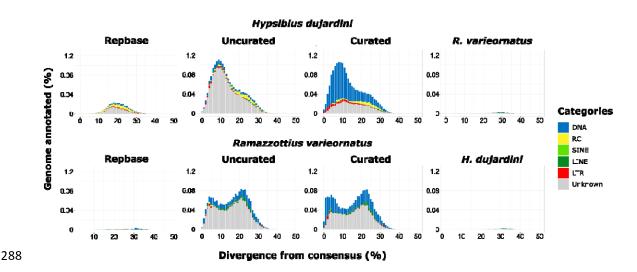


Figure 4. Repeat landscapes of the genomes of *H. dujardini* and *R. varieornatus* annotated with the Repbase (Arthropoda clade), uncurated and curated of both tardigrades combined libraries, and with libraries of the reciprocal species (only species-specific repeats). The divergence from consensus calculated with the Kimura 2-parameter distance model is shown on the x-axis. The percentage of genome annotated is shown on the y-axis.

Table 2. Number of base pairs annotated and percentages of the main TE categories.

S peci es	Library	DNA (bp)	DNA (%)	LINE (bp)	LINE (%)	SINE (bp)	S INE (%)	LTR (bp)	LT R (%)	Unknown (bp)	Unknown (%)	Total(bp)	Total (%)
	Repbase Arthropoda	347033	0.34	75334	0.07	264	0	2 00462	0.2	1370894	1.34	1993987	1.95
Hypsibius dujardini	Uncurated	1681052	1.65	3 10 2 3 9	0.3	5 16 6	0.01	5 14564	0.5	14199202	13.92	16710223	16.38
	C ura te d	111495 52	10.93	290632	0.28	2 42 4	0	868156	0.85	4658887	4.57	16969651	16.63
	R. varieornatus	62676	0.06	60480	0.06	0	0	8437	0.01	609 17	0.06	192510	0.19
R ama zzotti us varieornatu s	Repbase Arthropoda	689 02	0.12	33938	0.06	266	0	16972	0.03	23959	0.04	144037	0.26
	Uncurated	1753754	3.16	4 13 647	0.75	4 48 6	0.01	134451	0.24	6375274	11.5	8681612	15.66
	C ura te d	3385077	6.11	454742	0.82	1320	0	145257	0.26	4880857	8.81	8867253	16
	H dujardini	45939	0.08	40575	0.07	132.0	0	6334	0.01	49444	0.09	143612	0.26

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Table 3: List of repeat subfamilies with putatively ongoing activity, i.e., at least 10 copies

with 0% distance to consensus.

TE category	Hypsibius dujardini	Ramazzottius varieornatus
DNA transposon	7	3
LTR retrotransposon	3	0
Unknown	0	2

300

301 As a demonstrative example of the contribution of the collaborative curation process in 302 providing novel insights into TEs diversity, taxonomic distribution and biology, we decided to 303 deeply characterize consensus sequences that we classified as Tc4. These elements have a 304 rather limited taxonomic distribution, few references in bibliography exist, and they 305 incompletely duplicate the target site upon transposition [26] which can impose challenges for 306 their classification. The Tc4 transposons are DDD elements firstly discovered in 307 Caenorhabditis elegans [26] where they recognize the interrupted palindrome CTNAG as 308 target site for insertion, and cause duplication of only the central TNA trinucleotide. 309 Regarding their taxonomic distribution, consensus sequences for Tc4 elements are known and 310 deposited only for nematodes and arthropods in RepeatPeps, Repbase and DFAM. 311 Phylogenetic analyses based on DDD segments confidently placed the four tardigrade Tc4

312 consensus sequences identified in *R. varieornatus* within the Tc4 clade in a sister relationship 313 with arthropod elements and with a branching pattern that reassemble the Panarthropoda 314 group (tardigrades + onychophorans + arthropods) within Ecdysozoa [27] (Figure 5A). The 315 DDD catalytic domain resulted to be highly conserved between different phyla (Figure 5B) 316 and the target site of tardigrades mirror what was previously observed in nematodes (i.e., 317 C|TNA|G where "]" marks the transposase cut site Figure 5C-D). We could therefore 318 hypothesize that these elements first originated during the diversification of Ecdysozoa. 319 However, broader comparative analyses involving more early-diverging Metazoa clades are 320 necessary to confirm this lineage-specific origin.

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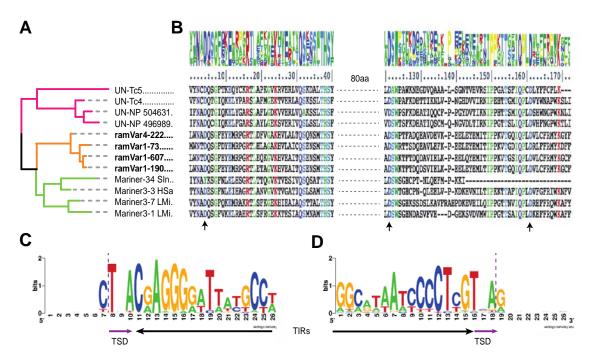


Figure 5. Characterization and phylogenetic analyses of Tc4 elements. (A) Phylogenetic tree of Tc4 consensus sequences based on DDD catalytic domains identified in the *R. varieornatus* consensus sequences, highlighted in bold and orange, together with representative sequences extracted from the RepeatPeps library from nematodes (pink) and insects (green). All nodes received maximal support value. (B) Alignment of DDD catalytic domains of sequences

included in phylogenetic analyses. Residues conserved in more than 80% of the sequences are
colored. Arrows highlight catalytic DDD residues. Sequence logos of 5' (C) and 3' (D) ends
of Tc4 elements used to curate the *R. varieornatus* consensus sequences. Black and purple
arrows denote terminal inverted repeats (TIRs) and target site duplications (TSDs),
respectively. The purple dotted line marks the transposase cut on the CTNAG target site.

333

334 Contributions from the course participants

During both editions of the course, participants were free to explore their favorite topics within the scope of the syllabus and we here share two contributions developed by the participants that can be useful for the entire community. First, an additional repeat library of Consensus sequences (119 of which are DNA transposons) was produced with the use of REPET for *R. varieornatus* (**Table S4**). Second, a guide for the classification of TEs from multisequence alignments (**File S1**) that can be a useful starting point for beginners and complementary to more extensive guides [13,14].

342 **Conclusion**

343 As shown here and in many other studies, repeat annotation is key to correctly identify and 344 interpret patterns of genome evolution and proper annotation is based on a thorough curation 345 of the repeat libraries [8,9,28]. However, it is hard for curation efforts to keep up with the 346 sheer amount of genomes released every year as curation done by single laboratories may 347 require months or even years for a single genome. Recent developments of machine learning-348 based tools to automatize the curation and classification processes are promising [29–32] and 349 there are additional tools to facilitate the curation process like TEAid [13] and EarlGrey [33]. 350 Until fully automatized, reliable tools are developed and there are manual curation training 351 sets for understudied taxa, we emphasize the need to implement manual curation for repeat 352 libraries as well as to find alternative ways to deal with the curation of hundreds of new

353 libraries. Here we presented one such alternative approach, namely a peer-reviewed course 354 sourcing effort designed to be as reproducible and comparable as possible and where the 355 hands-on tutorials were designed to be meaningful for the participants because they dealt with 356 real unexplored data and directly contributed to the scientific community. The two iterations 357 of this course sourcing effort resulted in the successful curation of hundreds of new and 358 diverse TEs and we hope that this experience and teaching framework can be of use for the 359 genomic and TE communities and to be applicable to other types of data/analysis that need 360 manual curation (e.g., genome assemblies [21,22] and gene annotations).

361 Materials and Methods

362 Genome assemblies

For this study, we used the genome assemblies of the two tardigrade species: *Hypsibius dujardini* (GCA_002082055.1) and *Ramazzottius varieornatus* (GCA_001949185.1) produced by sequencing a pool of male and female individuals by Yoshida et al. [34]. The *Hypsibius dujardini* genome was assembled using long PacBio and short Illumina reads whereas the *Ramazzottius varieornatus* genome was assembled using a combination of Sanger and Illumina reads [34].

369 Raw repetitive element library

To start the *de novo* characterization of transposable elements, we ran RepeatModeler on *H. dujardini* and RepeatModeler2 on *R. varieornatus* [35] using the option -LTR_struct and obtained a library of raw consensus sequences for each of the genomes. RepeatModeler and RepeatModeler2 automatically named the consensus sequences with the prefix "rnd" that we replaced with the abbreviations of the species names: "hypDuj" for *H. dujardini* and "ramVar" for *R. varieornatus*.

376	The two libraries were then compared to find similar sequences belonging either to the same
377	family or subfamily by using, respectively, the 80-80-80 rule [36] and the 95-80-98 rule [37].
378	The comparison was done by masking the library of R . varieornatus with the library of H .
379	dujardini using RepeatMasker [38].
380	Manual curation of the consensus sequences
381	After the generation of the libraries of raw consensus sequences, we proceeded with the
382	collaborative peer-reviewed manual curation step. The participants of the course were split
383	into ten groups and each group received about 80 consensus sequences to curate.
384	The first step of the curation consisted in the alignment of the raw consensus sequences to the
385	genome of origin using BLAST [39]. The best 20 BLAST hits were selected aligned with
386	their raw consensus sequence with MAFFT [40] which produced a multisequence alignment
387	for each consensus sequence ready to be manually curated (script RepeatModelerPipeline.pl).
388	Each of the multisequence alignment was then inspected to: 1) find the actual boundaries of
389	the repetitive elements; 2) build a new consensus sequence with Advanced Consensus Maker
390	(https://hcv.lanl.gov/content/sequence/CONSENSUS/AdvConExplain.html); 3) fix
391	ambiguous base and gap calls in the new consensus sequence following a majority rule; 4)
392	find sequence hallmarks to define the repetitive elements as transposable elements (e.g., target
393	site duplication, long terminal repeats, terminal inverted repeats or other motifs). Every new
394	consensus sequence was reported in a common Excel table (Table S1). To quantitatively
395	measure the improvement of the repeat libraries after manual curation, we compared the
396	length of consensus sequences before and after curation.
397	In all the figures and tables, the term "curated" indicates that the library mentioned contains
398	manually curated consensus sequences as well as all the consensus sequences that remained

399 uncurated. Finally, we consider each consensus sequence as a proxy for a transposable

400 element subfamily. However, the consensus sequences were not checked for redundancy and

401	not clustered into families and subfamilies using the 80-80-80 or 95-80-98 rules for
402	nomenclature because the focus of the study was on classifying the consensus sequences into
403	superfamilies and orders of transposable elements.

The code used to produce the consensus sequences and their alignments is provided as tutorial
on the GitHub repository https://github.com/ValentinaPeona/TardigraTE.

406 Classification

The new consensus sequences were classified using sequence characteristics retrieved by the alignments (e.g., target site duplications, terminal repeats) and homology information retrieved through masking the sequences with Censor [41,42] following the recommendations from [36] and [43]. When the information retrieved by the alignments and Censor were not enough to provide a reliable classification of the elements, the sequences were further analyzed for the presence of informative protein domains using Conserved Domain Database [44–46].

Since the course participants in general had never curated transposable element alignments before, we decided to implement a peer-review process. For the first course (*H. dujardini*), the results of each participant were sent to another participant to check the curated alignments and independently retrieve key information for the classification. The independent sequences and classifications would be compared and fixed if necessary. In the second course (*R. varieornatus*), all sequences were inspected by the same 3 reviewers only who applied the same process as previously described.

421 Comparative analysis of the repetitive content

The genome assemblies of both tardigrade species were masked with RepeatMasker 4.1.10 using four different types of TE libraries: 1) known Arthropoda consensus sequences from Repbase; 2) raw uncurated consensus sequences from the respective species; 3) curated consensus sequences together with the consensus sequences that were not curated from the respective species; 4) curated consensus sequences together with the consensus sequences that were curated from the other species. The RepeatMasker output files were then used to get the percentages of the genomes annotated as TEs and to visualize the landscapes of the accumulation of repeats.

Finally, we estimated the number of putative active transposable elements in the two genomes
by filtering the RepeatMasker annotation for elements that show at least 10 copies with a 0%
divergence from their consensus sequences.

433 Characterization of Tc4 elements

During the manual curation process, participants found types of DNA transposons that are currently considered to have a rather restricted phylogenetic distribution like Tc4 Mariner elements, therefore more in-depth analyses were run on these elements. The protein domains of known Tc elements were compared to the Tc4 consensus sequences from the tardigrade species and phylogenetic relationships were established.

439 Protein homologies of the partially curated repeat libraries were collected using BlastX (e-440 value 1e-05) [47] against a database of TE-related protein (RepeatPeps library) provided with the RepeatMasker installation. We extracted the amino acid translation of each hit on Tc4 441 442 elements based on the coordinates reported in the BlastX output. Resulting protein sequences 443 were aligned together with all members of the TcMar superfamily present in RepeatPeps 444 library using MAFFT (L-INS-i mode) [48] and the alignment was manually inspected to 445 identify and isolate the catalytic DDD domain. The resulting trimmed alignment was used for 446 phylogenetic inference with IQ-TREE-2 [49], identifying the best-fit evolutionary model with 447 ModelFinder2 and assessing nodal support with 1000 UltraFastBootstrap replicates [50]. The 448 resulting maximum likelihood tree was mid-point rooted and the Tc4 subtree extracted for 449 visualization purposes. The DDD segments of Tc4 elements were re-aligned using T-Coffee 450 in *expresso* mode [51] to produce conservation scores. A sequence logo of 5' and 3'

- 451 boundaries of identified Tc4 elements was produced extracting all sequences used to curate
- 452 the four *R. varieornatus* Tc4 elements and keeping the first 15 bp and 11 bp before and after
- 453 the terminal inverted repeats (TIRs), respectively.
- 454 Additional transposable element library
- Participants ran REPET tool V3.0 [52] to produce a de novo transposable element library for *R. varieornatus* in parallel to the one generated by RepeatModeler2. A custom TE library composed by repeats from Repbase and from *H. dujardini* was used to aid REPET in the classification process. Only consensus sequences that showed two or more full-length copies in the *R. varieornatus* genome were retained in the new library. Furthermore, the consensus sequences were scanned for protein domains and presence of TIRs or long terminal repeats (LTRs).
- 462 **Abbreviations**
- 463 LTR: Long Terminal Repeats
- 464 TE: transposable element
- 465 TIR: Terminal Inverted Repeats

466 **Declarations**

- 467 *Ethics approval and consent to participate*
- 468 Not applicable.
- 469 *Consent for publication*
- 470 Not applicable.
- 471 Availability of data and materials

472	All data generated or analyzed during this study are included in this published article and its
473	supplementary information files. All newly curated repeat consensus sequences were
474	deposited in Dfam. The code for the tutorials used in the course as well as for the analysis of
475	the manuscript can be found on GitHub: <u>https://github.com/ValentinaPeona/TardigraTE</u> .
476	Competing interests
477	Carlo Pecoraro is founder of Physalia Courses (http://www.physalia-courses.org/) but had no
478	role in the design of the study.
479	Authors' contributions
480	AS conceived the project and VP contributed to its development. VP and JM analyzed the
481	data. AS, VP, JM wrote the manuscript, and all authors revised the manuscript. MT, AM, DA,
482	JS, GP provided additional contributions to the teaching material. All authors except CP
483	contributed to the curation of the repeat library. CP provided and maintained the
484	computational infrastructure during the courses. Authors are listed in alphabetical order.
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489	

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