1 CNEr: a toolkit for exploring extreme

² noncoding conservation

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22 Abstract

23 Conserved Noncoding Elements (CNEs) are elements exhibiting extreme noncoding 24 conservation in Metazoan genomes. They cluster around developmental genes and act as long-25 range enhancers, yet nothing that we know about their function explains the observed 26 conservation levels. Clusters of CNEs coincide with topologically associating domains (TADs), 27 indicating ancient origins and stability of TAD locations. This has suggested further hypotheses 28 about the still elusive origin of CNEs, and has provided a comparative genomics-based method 29 of estimating the position of TADs around developmentally regulated genes in genomes where 30 chromatin conformation capture data is missing. To enable researchers in gene regulation and 31 chromatin biology to start deciphering this phenomenon, we developed CNEr, a R/Bioconductor 32 toolkit for large-scale identification of CNEs and for studying their genomic properties. We apply 33 CNEr to two novel genome comparisons - fruit fly vs tsetse fly, and two sea urchin genomes -34 and report novel insights gained from their analysis. We also show how to reveal interesting 35 characteristics of CNEs by coupling CNEr with existing Bioconductor packages. CNEr is 36 available at Bioconductor (https://bioconductor.org/packages/CNEr/) and maintained at github 37 (https://github.com/ge11232002/CNEr).

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44 Introduction

45 Conserved Noncoding Elements (CNEs) are a pervasive class of extremely conserved elements 46 that cluster around genes with roles in development and differentiation in Metazoa [1,2]. While 47 many have been shown to act as long-range developmental enhancers [3,4], the source of their 48 extreme conservation remains unexplained [5,6]. The need to maintain arrays of CNEs in cis to 49 the genes they regulate has led to their spatial arrangement into clusters termed Genomic 50 Regulatory Blocks (GRBs) [7,8]. The role of those clusters in genome organisation is suggested 51 by recent findings demonstrating that ancient metazoan clusters of extreme noncoding 52 conservation coincide with topologically associating domains (TADs) [9].

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Numerous recent studies highlight and seek to elucidate the importance of functional noncoding regions, most recently by employing the CRISPR-Cas9 based techniques to locate and dissect elements that affect gene expression and phenotype/disease - associated processes [10–12]. Prioritizing target loci of interest for interrogating the function of their regulatory context will be one of the major focuses of functional genomic studies, as has been shown in the case of the *POU5F1* locus [13], and of *NF1*, *NF2* and *CUL3* genes [14]. CNEs and the regulatory landscapes defined by their clusters serve as excellent candidates for such studies [3,15,16].

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A handful of CNE resources exist, mainly databases, which contain already pre-computed clusters of CNEs. These databases are static and mostly not updated. A summary of these resources is available in the review by Polychronopoulos et al. [6]. To our knowledge, there are only two tools available for the identification of conserved elements: PHAST [17] and CNEFinder [18]. The former relies on multiple sequence alignments and requires extensive computation time to derive "conserved" and "non-conserved" states from a two-state phylogenetic hidden Markov model (phylo-HMM), a space-time probabilistic model that

69 considers both the nucleotide substitution of each base in the genome sequence through 70 evolution and the transition from one base to the next. The latter produces CNEs based on a k-71 mer technique for computing maximal exact matches thus finding CNEs without the requirement 72 of whole-genome alignments or indices. Neither of them comes with a comprehensive, easy-to-73 follow suite of tools tailored to the integrated exploration of CNEs from end-to-end: from 74 identification to guality control and visualisation. Our package couples those processes 75 together, enabling the user to harness the support and wealth of packages available through the 76 common the Bioconductor infrastructure. Our package is specifically designed for efficient 77 identification of CNEs using user-specified thresholds, and it functions equally well across 78 vertebrates, invertebrates or plants. To study the evolutionary dynamics of these elements and 79 their relationship to the genes around which they cluster, it is essential to be able to both 80 produce and explore genome-wide sets of CNEs for a large number of species comparisons in 81 a dedicated workflow, each with multiple length and conservation thresholds.

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83 The CNEr package aims to detect CNEs and visualise them along the genome under a unified 84 framework. For performance reasons, the implementation of CNE detection and corresponding 85 I/O functions are primarily written as C extensions to R. We have used CNEr to produce sets of 86 CNEs by scanning pairwise whole-genome net alignments with multiple reference species, each 87 with two different window sizes and a range of minimum identity thresholds, available at 88 http://ancora.genereg.net/downloads. In this work, we demonstrate the application of CNEr to 89 the investigation of noncoding conservation between fruit fly Drosophila and tsetse fly Glossina -90 the two species at the evolutionary separation not previously investigated in insects [7] - and 91 between two species of sea urchins. This has enabled us to observe some properties of GRB 92 target genes shared across Metazoa. In a previous study, we showed that more distant 93 comparisons in Diptera (between Drosophila and mosquitoes) failed to identify CNEs [7]. On the 94 other hand, the conservation level across different species of the Drosophila genus is

95 comparable to that across placental mammals. With Drosophila and Glossina, we wanted to 96 explore the evolutionary divergence comparable to human vs. fish in another lineage and 97 establish whether it is the same functional class of genes that is accompanied by CNEs 98 featuring such a deep level of conservation. In the case of sea urchins, we wanted to investigate 99 a lineage at an intermediate distance to vertebrates - closer than insects, more distant than the 100 early branching chordates - in order to establish the continuum of GRBs across Metazoa. We 101 present a series of downstream analysis of the newly identified CNEs, identifying their 102 characteristic sequence features in invertebrates and functional classes of genes whose loci 103 they span.

104 Design and Implementation

105 Overview of CNEr workflow

106 CNEr provides the functionality of large-scale identification and advanced visualisation of CNEs 107 based on our previous strategies of detecting CNEs [7,8,19] as shown in Fig 1. CNEr scans the 108 whole genome pairwise net alignment, which can be downloaded from UCSC or generated by 109 the CNEr pipeline, for conserved elements. Various guality controls of the alignments are 110 provided. The composition of aligned bases in the alignment can be used for tuning parameters 111 during pairwise alignment (S1 Fig). More closely related species are expected to give higher 112 rates of matched bases. The syntenic dotplot of the alignments (S2 Fig) quickly shows the 113 syntenic regions between two assemblies.

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115 Fig 1: CNEr workflow.

(A) A typical pipeline of identification and visualisation of CNEs. (B) Illustration of scanning analignment for CNEs. The scanning window moves along the alignment for conserved regions.

118 The exons and repeats regions are skipped during the scanning by default.

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Considering the different extents of evolutionary divergence and sequence similarity between assemblies, we typically use the identity thresholds of 70% to 100% identity over a scanning window of 30 bp or 50 bp. Known annotations of exons and repeats are compiled from sources such as UCSC [20] and Ensembl [21] for common genomes, and elements overlapping with these regions are typically skipped during the scanning. Genome annotation pipeline, such as MAKER [22], can be used to create annotations for new genome assemblies.

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127 Net alignments only keep the best match for each region in the reference genome. This is not 128 acceptable when one of the aligned genomes underwent one or more whole genome 129 duplications, leading to legitimate deviations from 1:1 orthology for many CNEs. To eliminate 130 the bias of the choice of reference genome in the alignment and to capture duplicated CNEs 131 during whole genome duplication (WGD), we scan two sets of net alignments by using each of 132 the two compared genomes as reference in turn. This strategy performs well when comparing 133 species with different numbers of WGD rounds, such as tetrapod vertebrates against teleost fish 134 [23], or common carp [24] against other teleost fish. In such cases, some of the identified CNE 135 pairs from two rounds of screening do overlap on both assemblies, and hence are merged into 136 one CNE pair. As the last step, we align the CNEs back to the two respective genomes using 137 BLAT and discard the ones with high number of hits. The remaining elements are considered to 138 be a reliable set of CNEs.

139

CNEr provides a quick overview of the genomic distribution of CNEs along chromosomes. In S3
Fig, each CNE between human and mouse is plotted relative to each human chromosome (x-

142 axis). We plot the cumulative number of CNEs over chromosomal positions. A CNE cluster is 143 represented as a sharp increase of height in y-axis with small change in x-axis. For visualisation 144 of CNEs in any genome browser, CNEr can export the CNE coordinates in BED file format and 145 CNE density (measured by the percentage of area covered by CNEs within a smoothing 146 window) in **bedGraph** and **bigWig** formats. Since running the whole pipeline of CNE detection 147 can be time-consuming, we also implemented a storage and guery system with SQLite as 148 backend. Based on the visualisation capability of the Gviz package [25], CNEr can produce 149 publication-quality horizon plots of CNE density along with other genomic annotations (see 150 Methods and Data). Examples of the horizon plots are given in the following sections.

151 CNEr package implementation

152 *CNEr* is a Bioconductor package developed in R statistical environment, distributed under the 153 GPL-2 licence for *CNEr* code, and UCSC Kent's licence for Jim Kent's C source code it builds 154 on [20]. Although *CNEr* supports compilation for both 32-bit and 64-bit systems across multiple 155 platforms, it has limited functionality on the Windows platform due to the lack of the external 156 sequence alignment software *BLAT* [26], which is required in the pipeline.

157 Overview of whole genome pairwise alignment pipeline

UCSC Genome Informatics (http://hgdownload.soe.ucsc.edu/downloads.html) provides the pairwise alignments between many popular species. However, there is a frequent need to produce pairwise alignments for novel genome assemblies for new species, or using specific assembly versions when they are not available from UCSC. This pipeline mostly requires external sequence aligners and UCSC Kent's utilities [20], and provides well-tested parameters for species with a varying degree of evolutionary divergence. In brief, first a sequence alignment software, *LASTZ* [27] or LAST [28], is used to find the similar regions between two repeat165 masked genomes. Then, if two neighbouring alignments are close enough in the genome, they 166 are joined into one fragment. During the alignment, every genomic fragment can match several 167 others, and the longest one is kept. Finally, blocks of alignments are grouped into stretches of 168 synteny and form the so called "net" alignments in Axt format [29]. *CNEr* comes with a vignette 169 to demonstrate the whole pipeline. The produced Axt alignment can be manipulated in R as the 170 *Axt* class, which is extended from *GRangePairs* class defined in *CNEr* (see S2 Text).

171 Overview of the Axt scanning algorithm

The Axt alignment scanning algorithm constitutes the central part of this package for the identification of conserved noncoding elements. Due to the massive manipulation of characters, we implemented this algorithm purely in C for performance reasons; it is available to the R environment through R's C interface. The minimal input is the Axt alignment and the ranges to filter out, i.e., the coding and/or repeat masked regions.

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178 The Axt screening algorithm proceeds as in S1 Algorithm. First, the Axt alignment is converted 179 into a linked Axt data structure as implemented in Jim Kent's UCSC source code [20]. The 180 filtering ranges are encoded into a hash table, where keys are the chromosome/sequences 181 names and values are pointers to the linked lists of coordinates ranges. We then iterate over the 182 linked Axt alignments. For each alignment, we use a running window to scan the alignment with 183 a step size of 1 bp. Each base is searched against the filtering hash table and matched bases 184 are skipped. All segments above the identity threshold are kept. The overlapping segments are 185 merged into larger pieces. This procedure produces a set of CNEs conserved between the two 186 aligned genome assemblies.

187 CNEr visualisation capability

188 Instead of using the standard density plot for CNE density (as implemented in e.g. the Ancora 189 browser), we introduce the horizon plot with the aim to increase the dynamic range of CNE 190 density visualisation. The horizon plot provides a way of visualising CNE density over several 191 orders of magnitude, and eliminates the need for multiple standard density tracks at different 192 thresholds along the genomic coordinates. Instead, a relatively low conservation threshold is 193 used, and multiple overlaid sections of the horizon plots will reveal peaks with different 194 conservation density (see Fig 3A and Fig 3B in horizon plot, Fig 3C and Fig 3D in Ancora 195 browser). We expand the functionality of "horizonplot" in *latticeExtra* package and integrate it 196 into Gviz [25], which is the plot engine used in CNEr.

197 **Results**

198 CNEr use case I: Drosophila-Glossina CNEs

199 Here we demonstrate the application of *CNEr* to the analysis of Tsetse Fly (*Glossina morsitans*) 200 CNEs and their putative target genes. *Glossina* is the sole vector of African trypanosomiasis 201 ("sleeping sickness"), and it mediates transmission of the disease during feeding on blood. It 202 has been shown previously [7] that, while there are tens of thousands of CNEs detected across 203 different Drosophila species, there are almost no highly conserved elements found between 204 Drosophila and malaria mosquito Anopheles gambiae or other mosquitos. Glossina and 205 Drosophila are much closer to each other than either of them is to mosquitos, having a common 206 ancestor that has diverged around 60.3 Mya (S4 Fig). With the recently available assembly and 207 gene annotation of Glossina [30] (see S1 Text), we were able to identify clusters of CNEs 208 between these two species. The clusters correspond to a subset of clusters defined by the

209 CNEs derived from comparisons of different *Drosophila* species. A further investigation of gene 210 functions, which are retained or missing in *Glossina*, was carried out by comparison with the 211 *Drosophila* clusters.

212

213 A summary of CNEs detected between Glossina and Drosophila is given in Table 1. As 214 expected, many fewer CNEs are detected from the comparison between Glossina and 215 Drosophila than between any two Drosophila species, since Glossina is an outgroup to the 216 Drosophila/Sophophora family. A closer examination of the CNE density plot in Ancora browser 217 [31] revealed many missing clusters of CNEs relative to CNE density across Drosophila 218 species, especially at a more stringent threshold. We wanted to find out if the missing and 219 retained CNE clusters differ with respect to the functional categories of the genes they span. In 220 the following analysis, the CNEs that are conserved for more than 70% over 30 bp are 221 considered.

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Table 1: The number of CNEs between D. melanogaster and several other species,

Minimum identity	vs D.	vs D.	vs D.	vs D.	vs G.
	ananassae	pseudoobscura	mojavensis	virilis	morsitans
70% over 30 bp	NC	NC	176366	204970	9691
80% over 30 bp	NC	313570	127293	146793	3924
90% over 30 bp	NC	212951	81436	92288	1922
96% over 30 bp	177759	128843	47408	52134	813

including G. morsitans

100% over 30 bp	112073	76715	26972	29445	414
70% over 50 bp	266385	248357	104476	120628	3185
80% over 50 bp	223975	177266	66063	75204	1796
90% over 50 bp	142899	96994	33455	37098	732
96% over 50 bp	79631	49380	16387	17831	244
98% over 50 bp	55460	33463	10741	11548	150
100% over 50 bp	29218	17201	5250	5585	66

NC, not counted due to the threshold being too low for close species.

223

224 The most deeply conserved vertebrate CNEs are usually associated with genes involved in 225 transcriptional regulation or development (trans-dev) functions [19]. Due to high divergence 226 between Drosophila and Glossina, the regions with detectable CNE arrays tend to be of low 227 CNE turnover, i.e. the process of sequence divergence and loss of ancestral CNEs is slow. If 228 the same functional subset of genes is surrounded by low-turnover CNE clusters as in 229 vertebrates, the encompassed genes will more likely be essential key developmental genes [5]. 230 Indeed, Drosophila genes associated with (i.e. nearest to) Glossina vs. Drosophila CNEs are 231 also associated with trans-dev terms (Fig 2A). Development, including organ, system and tissue 232 development, appears at the majority of the top Gene Ontology (GO) terms. The other highly 233 significant GO terms include biological regulation, regulation of cellular process and cell 234 differentiation. CNE clusters can span regions of tens to hundreds of kilobases around the 235 actual target gene, which is on average shorter than the equivalent spans in vertebrate 236 genomes. This is in agreement with our observation that CNE clusters and the GRBs they

237 define (and, by extension, the underlying TADs) expand and shrink roughly in proportion to 238 genome size [9]. The H15 and mid locus (Fig 3A) is one of the biggest CNE clusters retained 239 between *Glossina* and *Drosophila*. The *H15* and *mid* genes encode the T-box family proteins 240 involved in heart development [32]. Although the CNE density between Drosophila and Glossina 241 is much lower than that across the Drosophila genus, it clearly marks the CNE cluster 242 boundaries of this locus, containing 67 CNEs at the 70% identity over 30 bp threshold. For the 243 40 largest retained CNE clusters, we provide a comprehensive list of CNE cluster coordinates, 244 the target genes, the protein domains and the number of associated CNEs (S1 Table). As we 245 can see, the majority of the target genes have Homeobox, Forkhead or C2H2 Zn finger 246 domains, just like the genes spanned by the most conserved CNE clusters in vertebrates.

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Fig 2: Over-represented GO Biological Process terms ranked by GeneRatio.

The gene ratio is defined as the number of genes associated with the term in our selected genes divided by the number number of selected genes. The p-values are adjusted using "BH" (Benjamini-Hochberg) correction. The visualisation is done by *clusterProfiler [33]*. (A) GO enrichment for genes nearest to *Drosophila* and *Glossina* CNEs. (B) GO enrichment for genes in the missing CNEs clusters compared between *Drosophila* and *Glossina*.

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Fig 3: Horizon plot of CNE density around key developmental genes along *D. melanogaster* as reference.

(A) *H15* and *mid* genes are spanned by arrays of CNEs. Despite the much lower CNE density from *D. melanogaster* and Glossina, a CNE cluster boundary shows up that is consistent with CNEs from other drosophila species. (B) The CNE cluster around *ct* gene is missing in the comparison of *D. melanogaster* and Glossina since no CNEs are detected. This implies that this region undergoes a higher CNE turnover rate. (C, D) The same loci as in (A, B) are shown on the Ancora browser in order to compare the normal CNE density plot with the horizon plot. Notations: ensGene, Ensembl gene track; *Glossina* 21/30, G. morsitans 70% identity over 30
bp; droAna2 49/50, *D. ananassae* 98% identity over 50 bp; dp3 48/50, *D. pseudoobscura* 96%
identity over 50 bp; droMoj2 48/50, *D. mojavensis* 96 % identity over 50 bp; droVir2 48/50, *D. virilis* 96% identity over 50 bp.

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268 Some other regions have strong clusters of CNEs conserved among Drosophila species, 269 however, the CNE cluster between Drosophila and Glossina is absent. The ct locus (Fig 3B), 270 encoding the cut transcription factor, is one of the best representative cases. Ct plays roles in 271 the later stages of development, controlling axon guidance and branching in the development of 272 the nervous system, as well as in the specification of several organ structures such as 273 Malpighian tubules [34]. In order to locate the CNE clusters missing from Drosophila vs. 274 Glossina comparison, we use the CNE clusters detected in D. melanogaster vs. D. ananassae 275 comparison as reference and compare them with the aforementioned retained CNE clusters. 276 The genes within those missing CNE clusters are highly enriched for axon guidance and 277 neuronal development (Fig 2B). We then examine the CNE turnover rate (the speed of replacing 278 old CNEs) of the 216 human genes that are associated with the axon guidance term 279 (GO:0007411), with both human and Drosophila as reference. The turnover rate is calculated as 280 the reduction of the number of CNEs between two sets of CNEs. For human reference, we 281 choose the CNE set of human vs. mouse and human vs. zebrafish, while *D. melanogaster* vs. 282 D. ananassae and D. melanogaster vs. Glossina are chosen for Drosophila reference. As 283 shown in Fig 4, the axon guidance genes have significantly higher turnover rate than the other 284 genes (p < 1e-5, Kolmogorov-Smirnov one-sided test) in both human and Drosophila lineages.

285

Fig 4: Cumulative distribution function of the changes of CNE number.

For a 40kb window around each orthologous gene pair between human and *drosophila*, we calculate the reduction of the number of CNEs for human (# of CNEs from human-mouse

comparison minus # of CNEs from human-zebrafish comparison) and drosophila (# of CNEs
from *D. melanogaster* vs. *D. ananassae* comparison minus # of CNEs from *D. melanogaster* vs. *Glossina*) as reference. The axon guidance genes significantly show a higher degree of CNE
number reduction, compared with the other genes (p < 1e-5, Kolmogorov-Smirnov one-sided
test).

294 *CNEr* use case II: sea urchin CNEs

295 In this section we apply CNEr to the comparison of highly fragmented genome assemblies of 296 two sea urchin species Strongylocentrotus purpuratus and Lytechinus variegatus (see S1 Text). 297 The purpose of this analysis is twofold. First, we want to demonstrate how well CNEr is able to 298 call CNEs and their clusters in the case of highly fragmented draft genomes: the ability to 299 perform this analysis on draft genome assemblies would show that our approach can be applied 300 to a large number of available genomes, most of which haven't been assembled past the draft 301 stage and are likely to remain in that state. Second, we wanted to ask if a third lineage, 302 evolutionarily closer to vertebrates than insects but still lacking any shared CNEs with 303 vertebrates, would exhibit the same patterns of noncoding conservation. This could provide a 304 hint towards CNEs' universal presence in Metazoa, in addition to providing an informative 305 additional dataset for comparative studies of genomic regulatory blocks.

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307 S. purpuratus is a popular model organism in cell and developmental biology. S. purpuratus and 308 L. variegatus have a divergence time of 50 Mya [35] and historically moderate rates of 309 sequence divergence, which makes them ideal for comparative genomics studies of regulatory 310 elements. We identified 18,025 CNEs with threshold of 100% identity over 50 bp window. 311 Despite the highly fragmented assemblies, we could clearly detect 808 prominent CNE clusters. 312

313 An especially interesting observation is the largest cluster we detected, at the *Meis* gene locus 314 (Fig 5). The CNE density clearly marks the boundaries of the CNE cluster. In Metazoa, Meis, 315 one of the most well-known homeobox genes, is involved in normal development and cell 316 differentiation. Tetrapod vertebrates have three Meis orthologs as a result of two rounds of 317 whole genome duplication. The CNE cluster around *Meis2* (one of three Meis paralogs arisen 318 by two WGD rounds at the root of vertebrates) is the largest such cluster in vertebrates [19]. 319 Remarkably, the cluster of CNEs around Drosophila's Meis ortholog, hth (homothorax), is also 320 the largest CNE cluster in the *D. melanogaster* genome [7]. It is currently unknown why the 321 largest clusters of deeply conserved CNEs are found around the same gene in three different 322 metazoan lineages, even though none of the CNEs from one lineage has any sequence 323 similarities to CNEs in the other two. The most plausible explanation is that the ancestral Meis 324 (*hth*) locus was already the largest such locus in the ancestral genome, and that CNE turnover 325 has led to three separate current lineage-specific sets of CNEs.

326

327 Fig 5: Horizon plot of CNE density at *Meis* loci on sea urchin *Strongylocentrotus* 328 *purpuratus.*

The density plots of CNEs detected at similarity threshold 96% (48/50), 98% (49/50) and 100% (50/50) over 50 bp sliding window, in *Lytechinus variegatus* comparison, are shown in three horizon plot tracks. The boundaries of CNE clusters from various thresholds are mutually consistent.

334 CNEr reveals interesting sequence features characteristic of

335 ultraconservation

336 It has been shown that vertebrate nonexonic CNEs are enriched in the TAATTA hexanucleotide 337 motif, which is an extended recognition site for the homeodomain DNA-binding module [36]. 338 With *CNEr*, we can easily verify the existence of TAATTA motif in CNEs of invertebrate species. 339 In S5 Fig A, we consider CNEs identified by CNEr that are conserved between D. melanogaster 340 and D. virilis over 98% for more than 50 nucleotides and plot them by increasing width using 341 heatmaps package (https://bioconductor.org/packages/heatmaps/). The first two heatmaps 342 confirm that CNEs are enriched in AT inside but exhibit a marked depletion of AT at their 343 borders, consistent with what is known about their biology in vertebrates [37]. Furthermore, the 344 TAATTA motif is enriched in insect CNEs. The motif seems to be extended further by flanking 345 A/T nucleotides. When replacing A/T (W) with G/C (S), the heatmap pattern disappears. We ask 346 whether this is a general property of CNEs in Metazoa and, using CNEr, proceed to the 347 identification of CNEs that are conserved between (a) C. elegans and C. briggsae at 100% for 348 more than 30 nucleotides (worm CNEs, see S5 Fig B), (b) L. variegatus and S. purpuratus at 349 100% for more than 50 nucleotides (sea urchin CNEs, see S5 Fig C). We observe that the same 350 pattern does not hold in those cases, i.e. it appears like enrichment of CNEs in TAATTA is not a 351 universal phenomenon but applies only to insect and vertebrate elements. Our pipeline is a 352 powerful tool for studying the guestion of how and when this TAATTA-enrichments originated. 353 as well as a multitude of related questions.

355 Downstream overlap analysis of CNEs reveals that they are

³⁵⁶ highly enriched in stem-cell regulatory elements

357 Finally, we demonstrate the utility of CNEr for general hypothesis generation about CNEs by 358 identifying elements highly conserved (>98% identity over 50 bp) between human and chicken 359 and performing global and pairwise overlap analyses against various genomic features using 360 the R/Bioconductor packages LOLA [38] and regioneR [39], respectively. LOLA allows for 361 enrichment analysis of genomic intervals using a core reference database assembled from 362 various resources of genomic data, while regioneR permits cross-validation of the findings 363 through pairwise overlap analyses. As evident from inspection of S2 Table and Fig 6, both 364 packages converge to the conclusion that the identified CNEs between human and chicken are 365 significantly enriched in Sox2 and Oct-4 (POU5F1) binding sites. Sox2 and Oct-4, in concert 366 with Nanog, are believed to play key roles in maintaining pluripotency. This finding comes in 367 accordance with previous reports suggesting that several CNEs are enriched in classical 368 octamer motifs recognized by developmental homeobox transcription factors [40]. Nonetheless, 369 this is the first time that such an association of the most deeply conserved CNEs with key 370 pluripotency elements is reported in the literature, and we anticipate that more associations of 371 this kind will be revealed in the future by coupling *CNEr* with other R/Bioconductor packages.

372

Fig 6: Pairwise overlap analysis of CNEs demonstrates association with Sox2, POU5F1 and Nanog binding regions.

In all three cases, permutation tests with 1000 permutations of CNEs are shown. In grey the number of overlaps of the randomized regions with the test regions of interest (in this case, *Sox2*, *POU5F1* and *Nanog*) are depicted. Those overlaps of the randomized regions cluster around the black bar that represents the mean. In green, the number of overlaps of the actual

379 regions (CNEs in this case) with the test regions is shown and is proved to be much larger than
380 expected in all cases. The red line denotes the significance limit.

381 Availability and Future Directions

The *CNEr* package with self-contained UCSC Kent's utility source code is available at Bioconductor release branch <u>http://bioconductor.org/packages/CNEr/</u>. Active development and bug reports is hosted on github <u>https://github.com/ge11232002/CNEr/</u>. Currently the *BLAT* is the only supported aligner for identifying repeats. Other high performance short read aligners that run in parallel, such as Bowtie1/2 and BWA, are desired for large set of CNEs. Furthermore, integration of GRB identification approach and GRB target gene prediction is planned for future development.

389 **Declarations**

390 Competing interests

391 The authors declare that they have no competing interests.

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396 Authors' contributions

- 397 GT and BL conceived the project. GT implemented the software, analyzed the data. DP tested
- 398 the software, analyzed data. GT, DP, BL wrote the manuscript.

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- 402 Supporting information
- ⁴⁰³ S1 Text. Glossina and sea urchin data.
- 404 S2 Text. Working with Paired Genomic Ranges
- ⁴⁰⁵ S1 Fig. The heatmap shows the percentage of matched bases in
- the Axt alignments.
- 407 This can be useful for examining the quality of Axt alignments, especially from the whole
- 408 genome pairwise alignment pipeline in CNEr package. The left panel has higher rates of
- 409 matches than right panel since the divergence of human and mouse is much smaller than that
- 410 between human and zebrafish.

S2 Fig. The syntenic plot of alignment blocks between chr1, chr2 of human and chr1, chr2 of mouse.

- 413 This plot is mostly used for tuning the parameters during whole genome pairwise alignment to
- 414 get better alignments. It can also show ancient duplications for the alignment of a sequence
- 415 against itself.

416 S3 Fig. The distribution of CNEs along the 6 biggest

417 chromosomes in human genome.

- 418 Each CNE is plotted as a dot with the position in chromosome as x-axis. A sharp increase in y-
- 419 axis represents a CNE cluster.

420 S4 Fig. The species tree of Drosophila, Glossina and mosquitos.

The phylogenetic tree is constructed based on the data on last common ancestors from TimeTree (Hedges et al., 2006). The genome of the malaria mosquito *A. gambiae* is highly divergent from *Drosophila* family and unsuitable for comparative genomics study, while *G. morsitans* is much closer.

425 S5 Fig . Sequence heatmaps of CNEs in different lineages.

- 426 (A) D. melanogaster and D. virilis (B) C. elegans and C. briggsae (C) L. variegatus and S.
- 427 *purpuratus.* The CNEs are ranked by decreasing CNE width.

428 S1 Table. A list of the most prominent CNE clusters detected

429 between Drosophila and Glossina.

430 S2 Table. Overlap analysis of CNEs.

- 431 Global overlap analysis of CNEs against multiple genomic features using LOLA reveals that
- they overlap with Nanog, Sox2 and POU5F1 binding regions. This table is the output of
- 433 runLOLA algorithm sorted by FDR. All top hits include important factors associated with
- 434 pluripotency. userSet: all CNEs conserved between human and chicken over 98% identity for
- 435 more than 50 bp. collection: all elements in CODEX database. universe: all active DNase I
- 436 hypersensitive sites.

437 S1 Algorithm. The algorithm of scanning Axt alignment.

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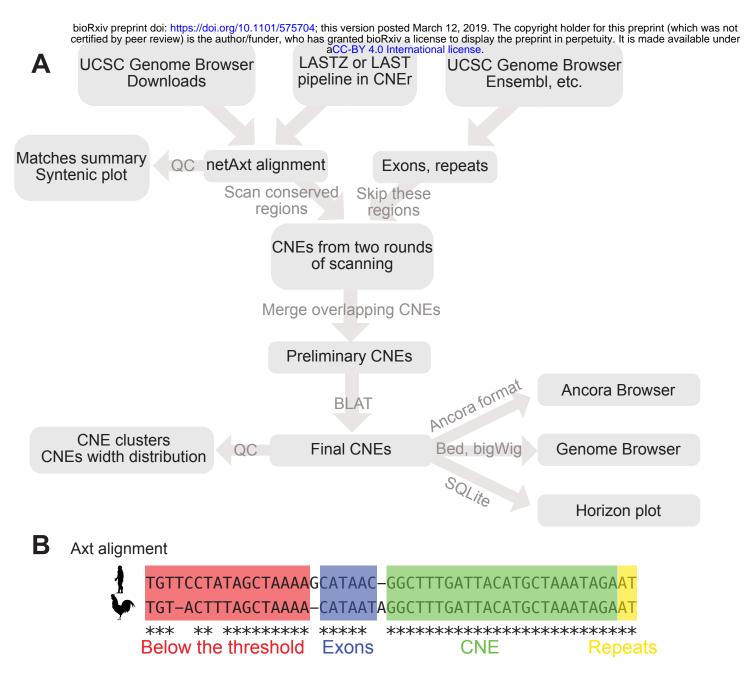
439 **References**:

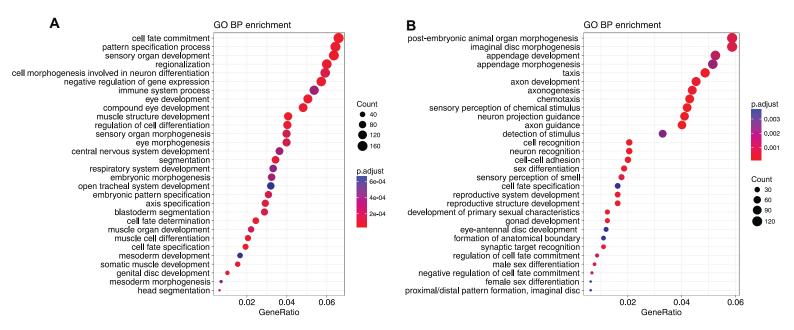
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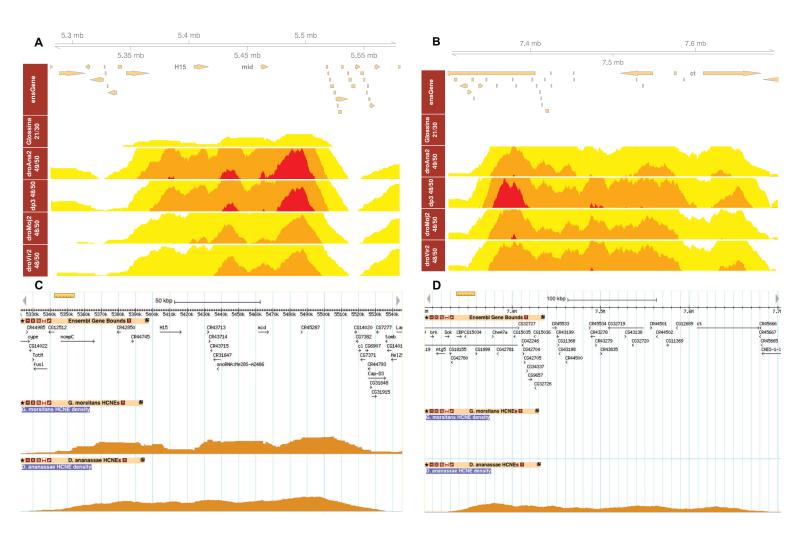
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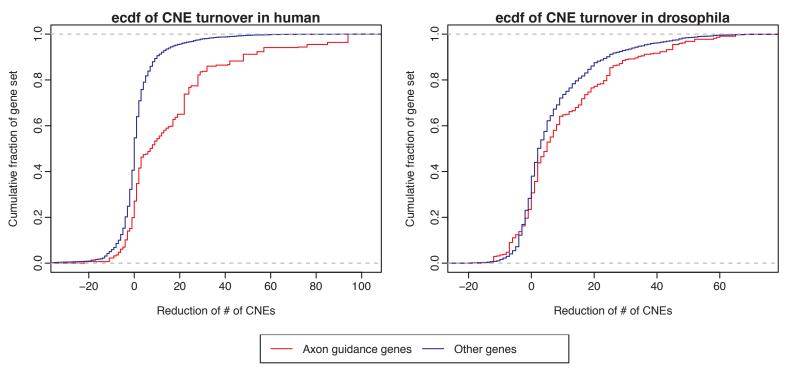


Figure 4

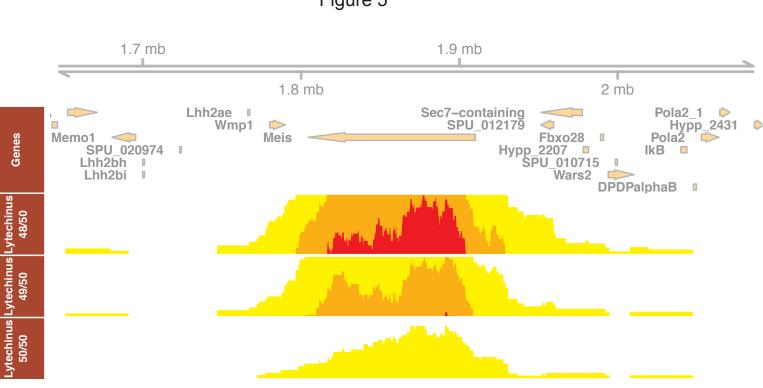


Figure 5

Genes

