

1 Phylogenomic reconstruction of *Cryptosporidium* spp. captured 2 directly from clinical samples reveals extensive genetic diversity

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34 Abstract

35 *Cryptosporidium* is a leading cause of severe diarrhea and mortality in young children and infants in
36 Africa and southern Asia. More than twenty *Cryptosporidium* species infect humans, of which *C. parvum*
37 and *C. hominis* are the major agents causing moderate to severe diarrhea. Relatively few genetic
38 markers are typically applied to genotype and/or diagnose *Cryptosporidium*. Most infections produce
39 limited oocysts making it difficult to perform whole genome sequencing (WGS) directly from stool
40 samples. Hence, there is an immediate need to apply WGS strategies to 1) develop high-resolution
41 genetic markers to genotype these parasites more precisely, 2) to investigate endemic regions and
42 detect the prevalence of different genotypes, and the role of mixed infections in generating genetic
43 diversity, and 3) to investigate zoonotic transmission and evolution. To understand *Cryptosporidium*
44 global population genetic structure, we applied Capture Enrichment Sequencing (CES-Seq) using
45 74,973 RNA-based 120 nucleotide baits that cover ~92% of the genome of *C. parvum*. CES-Seq is
46 sensitive and successfully sequenced *Cryptosporidium* genomic DNA diluted up to 0.005% in human
47 stool DNA. It also resolved mixed strain infections and captured new species of *Cryptosporidium* directly
48 from clinical/field samples to promote genome-wide phylogenomic analyses and prospective GWAS
49 studies.

50 Introduction

51 Diarrheal diseases account for about half a million child fatalities every year worldwide, making
52 them one of the leading causes of morbidity and mortality in children less than five years of age (1,2).
53 Although most diarrhea-related deaths are preventable through adequate sanitation, children with
54 impaired immunity, malnutrition, or immunocompromised adults are at the most risk of life-threatening
55 diarrhea. Strikingly, based on a recent Global Enteric Multicenter Study (GEMS) (1), the intestinal
56 apicomplexan parasite *Cryptosporidium* is the second leading cause of severe diarrhea in infants under
57 two years of age and is the most common persistent diarrhea-causing pathogen in young children in
58 Africa and southern Asia (3). In addition to diarrhea-related death, cryptosporidiosis in children is
59 associated with malnutrition, persistent growth retardation, impaired immune responses, and cognitive
60 deficits (4,5). Due to the absence of vaccines, the control of cryptosporidiosis relies solely on sub-
61 optimal chemotherapy, which is limited in efficacy and supply (6). Thus, there is an urgent need for new
62 genetic technologies to identify potential drug/vaccine targets to combat this parasitic infection.

63 Currently, greater than 20 *Cryptosporidium* species have been identified to infect humans (7,8).
64 Genotyping markers that are applied to resolve the species causing infection include conserved genes
65 such as the 18S small subunit (SSU) rRNA, the 70 kDa heat shock protein (*hsp70*), actin, and the
66 oocyst wall protein 1 (*cowp1*) genes (9). The most commonly used subtyping target is the highly
67 polymorphic 60 kDa glycoprotein gene (*gp60*). The major human cryptosporidiosis-causing pathogens

68 are *C. hominis* and *C. parvum* (10). Despite their high genetic similarity, *C. hominis* maintains a strictly
69 anthroponotic transmission cycle whereas a majority of *C. parvum* genotypes are considered zoonotic
70 pathogens. In addition to these two species, several additional zoonotic species have been detected
71 infecting humans including *C. canis* (usual host, dogs), *C. cuniculus* (rabbits), *C. felis* (cats), and *C.*
72 *meleagridis* (various) (11). The genetic basis for host adaptation, zoonotic and/or anthroponotic
73 transmission is unclear due principally to a paucity of *Cryptosporidium* spp. whole-genome sequence
74 (WGS) data from this parasite's wide host range.

75 Over the last decade, WGS data using next-generation sequencing (NGS) has grown
76 extensively, however it has not been widely applied to *Cryptosporidium* because it is difficult to purify
77 enough parasite material from clinical samples or propagate the parasite *in vitro*. Specifically, few
78 animal models are available to propagate parasites, and *in vitro* propagation is not well developed to
79 produce sufficient pure genomic DNA for WGS (12-14) (15). Additionally, the current oocyst purification
80 efforts to extract high quality gDNA involve sucrose flotation (16), discontinuous sucrose gradient
81 centrifugation (17), cesium chloride (CsCl) gradient centrifugation (18), and immunomagnetic
82 separation (IMS) (19). These steps to enrich parasite oocysts must be performed using fresh
83 clinical/field samples, which must be collected and stored without freezing because the freeze-thaw
84 procedure mechanically ruptures the oocyst wall rendering the techniques above intangible (20). Third,
85 the absence of autofluorescence by *Cryptosporidium* oocysts makes it difficult to FACS-purify single
86 oocysts to perform single-cell sequencing. Finally, current enrichment methods often copurify material
87 with similar buoyant-density or substances adhered to the surface of the oocysts, including host cells
88 (>3Gb) and food particles (>300Mb), that significantly reduce sequence coverage of the
89 *Cryptosporidium* genome (20).

90 Currently, the few WGS studies pursued to investigate *Cryptosporidium* have been conducted
91 by concentrating and purifying oocysts from stool samples (21-24). However, these are typically limited
92 to symptomatic samples that possess relatively high parasite burdens ($\geq 10^3$ oocysts per gram).
93 Strikingly, asymptomatic clinical manifestation among African and South Asian children is very common
94 (25) and parasite oocyst numbers in these patients are very limited. High-resolution WGS followed by
95 comparative genomics between asymptomatic versus symptomatic carriers is critical to address how
96 parasite genetic factors may influence cryptosporidiosis in the context of other variables such as host
97 immunity and genetics, malnutrition and the gut microbiome. Little progress has been made to develop
98 a highly sensitive WGS technology directly from asymptomatic patient stool samples as most of the
99 enrichment methods result in significant oocyst loss (26). Additionally, it is well documented that
100 increased disease severity has been strongly associated with mixed infections among other closely
101 related apicomplexan parasites such as *Plasmodium* (27). Mixed infection in *Cryptosporidium* has been

102 shown to promote genetic exchange, both intra- and inter-specific in the evolution of new subtypes that
103 possess different biological potentials or host preferences (28). Specifically, admixture of these highly
104 genetically diverse strains has led to the spread of drug-resistant parasites and the emergence of new
105 strains with altered host preferences, or the creation of hypervirulent strains (29-33). High-resolution
106 WGS data has proved to be highly sensitive and discriminatory and can detect population heterogeneity
107 among circulating pathogens (*i.e.* mixed infection) by deconvoluting individual genotypes within a
108 sample (34-36). Thus, it is extremely critical to understand the prevalence of mixed infection using WGS
109 data in high endemic areas and its role introducing genetic diversity through recombination among
110 currently circulating *Cryptosporidium* strains (21).

111 Capture enrichment sequencing (CES-Seq) has been applied successfully to concentrate
112 *Wolbachia* DNA from whole insect DNA extracts (37), to enrich *Yersinia pestis* DNA from Black Death
113 victims (38), and to detect and characterize felid pathogens for veterinary diagnosis and discovery (39).
114 Additionally, RNA enrichment methods for high resolution quantitative transcriptional analysis using
115 SureSelect CES-Seq have been used *in vivo* to enrich fungal transcripts up to 1,600-fold (40). Recently
116 many genomes of *Leishmania donovani* were sequenced directly from visceral leishmaniasis patient
117 samples and the isolates sequenced *in situ* possessed lower aneuploidy and fewer genomic differences
118 than culture-derived amastigotes from the same patients (41). Our aim was to develop *Cryptosporidium*
119 WGS by applying CES-Seq directly on DNA extracted from clinical stool samples to advance the field
120 of *Cryptosporidium* population genetics, to identify the extent to which mixed infections occur and to
121 identify the true diversity of *Cryptosporidium* species that infect people throughout the world, whether
122 or not symptomatic disease is reported. To perform WGS directly from stool samples, without the
123 requirement to purify oocysts directly from fresh samples, we developed 74,973 RNA baits (probes) to
124 capture, amplify and sequence the whole genome of *Cryptosporidium* using SureSelect technology
125 (Agilent, CA, USA). We demonstrate that the probes can successfully capture as little as 0.005% target
126 gDNA present in a clinical sample at WGS resolution, and that the method captures, with nearly equal
127 efficiency, the medically important species *C. parvum*, *C. hominis*, and *C. meleagridis*. We also show
128 that the CES-Seq method resolved infections with other more distantly related *Cryptosporidium*
129 species, including *C. ubiquitum* and *C. canis*, can be applied to previously frozen material, and is able
130 to distinguish multiple genomes in artificially mixed clinical samples at WGS resolution.

131

132 Materials and Methods

133 Ethics Statement and Clinical Samples

134 Stool samples collected from individuals from Colombia, Ecuador and Egypt were analyzed for the

135 presence of *Cryptosporidium* DNA. These samples had been examined previously for the presence of
136 other protists, as described previously (42). DNA extracted from stool samples from *Cryptosporidium*
137 positive patients from collaborators in Ghana and the United Kingdom were also investigated. Briefly,
138 the cohort population from Colombia consisted of healthy volunteers (n=79) between 16 to 41 years-
139 old from equatorial Colombia who had confirmed *Giardia* infections by microscopy. Fecal samples from
140 these volunteers were preserved in 100% ethanol. The second cohort populations were from Ecuador
141 (n=12) and Egypt (n=24) and were also comprised of microscopically *Giardia* positive patient samples
142 that we assayed for co-infection with *Cryptosporidium* spp. These samples were collected based on
143 approved protocols by the ethics committee of Universidad INCCA de Colombia (protocol number =
144 237894) with written consent from volunteers and patients, as described previously (42). Ethics
145 approval to obtain clinical samples from Ghana (n=10) was obtained from the Noguchi Memorial
146 Institute for Medical Research Institutional Review Board (Certified Protocol Number: 061/15 - 16).
147 Permission was also obtained from the appropriate authorities from each study site. DNA from clinical
148 samples from the UK (n=10) was provided by Public Health Wales – PHW – Microbiology and Health
149 Protection, UK under a material transfer agreement between NIH and PHW for the analysis of de-
150 identified *Cryptosporidium* DNA, which did not require ethical approval. DNA was extracted from patient
151 samples according to the protocols listed below, anonymized by dis-linking all patient identifiers, and
152 the extracted gDNAs were shipped to NIH for CES-seq.

153 **DNA Isolation**

154 Genomic DNA was extracted from *C. parvum* oocysts (purchased from Bunchgrass Farms) and the UK
155 clinical samples using the Qiagen DNA Stool kit whereas for all other clinical samples, the Dneasy
156 PowerSoil Pro kit (Qiagen, USA) was used, according to the manufacturer's instructions (Supplemental
157 figure 1A). Specifically, 1×10^7 excysted oocysts of *C. parvum* purchased from Bunchgrass Farms were
158 utilized to prepare the gDNA which was used to generate artificial mixtures by diluting *C. parvum* gDNA
159 into gDNA from healthy human stool samples. *C. hominis* DNA from isolate TU502 (NR-2520) and *C.*
160 *meleagridis* DNA from isolate TU1867 (NR-2521) were obtained from BEI Resources (Manassa, VA).
161 The gDNAs were analyzed by 1% agarose gel electrophoresis, stained with ethidium bromide, imaged
162 by Syngene Gel documentation system (GBX-CHEMI-XL1.4. Fisherscientific, USA) and compared with
163 GeneRuler 1kb Plus DNA Ladder (Thermo Scientific, USA). Quantity and quality of gDNA were
164 assessed using the Invitrogen Qubit 4 Fluorometer (Invitrogen, USA), DS-11 Series Spectrophotometer
165 (DeNovix, USA), 4200 TapeStation System (Agilent, USA), and 2100 Bioanalyzer Instrument (Agilent,
166 USA).

167 **Real-time Quantitative PCR (qPCR)**

168 Real-time qPCR was performed using 18S rRNA gene primers (Table S1), a QuantStudio 6 Flex Real-
169 Time PCR system (Applied Biosystems, USA), with a 20 μ l reaction mixture, containing 1 X Power
170 SYBR Green PCR master mix (Applied biosystems, USA), 500nM of each primer, and 10-fold serial
171 dilutions (from 10ng to 0.0001ng) of *C. parvum* gDNA. Sterile water and gDNA from healthy volunteers
172 were used as non-template controls and were run with every assay. All assays were performed in
173 triplicate to ensure reproducibility.

174 **Sanger Sequencing of 18S rRNA Gene for Species Assignment**

175 DNA extracted from clinical samples infected with *Cryptosporidium* were also subjected to 18S rRNA
176 gene Sanger sequencing to assign species after PCR amplification using nested primers: external
177 primers, 18SrRNA(ext)F (5'- CCTGCCAGTAGTCATATGCTT-3') and 18SrRNA(ext)R (5'-
178 GAATGATCCTCCGCAGGT-3'); internal primers, 18SrRNA(int)F (5'-GTTAAACTGCGAATGGCTCA-
179 3') and 18SrRNA(int)R (5'- CGAAACTTCCATTACATGTATTGCT-3'). PCR products were purified with
180 a QIAquick PCR purification kit (QIAGEN, USA) according to the manufacturer's instructions.
181 Sequencing was conducted with two independent templates using BigDye cycle sequencing (Applied
182 Biosystems, USA) by Quintara Bio (Quintarabio, USA). Nucleotide sequences were assembled using
183 FinchTV 1.4 Sequence Alignment Software (<https://digitalworldbiology.com/FinchTV>, Geospiza, Inc)
184 and aligned with previously published sequences in GenBank (<https://www.ncbi.nlm.nih.gov/genbank/>)
185 using ClustalX (43) with default settings. Maximum-likelihood phylogenetic trees were generated with
186 the aligned sequences in Nexus format (44) using Molecular Evolutionary Genetic Analysis (MEGA-X)
187 (45), Tamura-Nei model (46). 1000 replicates were used to generate the bootstrap values.

188 **Whole Genome CES-Seq**

189 Initially, artificial mixtures (*C. parvum* genomic gDNA diluted in gDNA extracted from healthy human
190 stool samples) of 200 ng total gDNA in different ratios were generated for capture-hybridized
191 sequencing using genome-wide biotinylated RNA baits for *Cryptosporidium*. In total, 74,973 RNA baits
192 (probes) were generated based on the available reference genome of *C. parvum* at the time this study
193 was initiated (<https://cryptodb.org/common/downloads/release-34/Cparvumlowall/fasta/data/>).
194 Duplicated baits and baits with high similarity to the human genome were filtered out. Exclusion criteria
195 for bait selection were known regions of mis-assembly, long stretches of low complexity, and gene-
196 poor regions in telomeric and sub-telomeric regions. In total, baits were designed to capture
197 approximately 92% of the *C. parvum* genome, based on the assembly of the Iowa II isolate, released
198 on CryptoDB_v34 (Table 1). Within the genome regions that were included for bait design, every
199 nucleotide position was covered with at least 2 overlapping baits, except in highly variable regions,
200 including telomeric and sub-telomeric regions that were gene-rich, in which 5 baits were designed.

201 Genomic DNA of artificial samples and clinical samples were first sheared by focused ultrasonication
202 (Covaris Inc., USA) into 200 to 300 bp fragments followed by a quality control assessment using the
203 Invitrogen Qubit 4 Fluorometer (Invitrogen, USA), 4200 TapeStation System (Agilent, USA), and 2100
204 Bioanalyzer Instrument (Agilent, USA). Capture enrichment libraries were prepared using SureSelect^{XT}
205 HS Target Enrichment System for Illumina Paired-End Multiplexed Sequencing Library
206 (<https://www.agilent.com/cs/library/usermanuals/public/G9702-90000.pdf>, Agilent Technologies, USA)
207 using genome wide baits with 12 amplification cycles following the manufactures' protocol (Fig. 1A).
208 Amplified capture libraries were purified with streptavidin coated beads and quantified again with
209 Bioanalyzer DNA 1000 chip (Agilent, USA) and D1000 ScreenTape (Agilent, USA). Finally, the genomic
210 libraries were prepared with Nextera XT Kits (Illumina, USA) and DNA sequencing was performed using
211 paired-end reads on a MiSeq, NextSeq or HiSeq system (Illumina, USA).

212 **DNA Sequence Read Processing**

213 Paired-end reads from CES-Seq were trimmed using Trimmomatic v.0.36 (47) to remove sequencing
214 adapters and low quality bases, followed by filtering of reads derived from human contamination.
215 Filtered raw reads were aligned against the *C. parvum* reference genome CryptoDB-54_Cparvumlowa-
216 ATCC_Genome (Table S3) using the Burrows-Wheeler Aligner (BWA mem, v0.7.15) (48). BAM files
217 were processed with SAMtools v1.3.1(49), BEDtools v2.25.0 (50), and Picard tools v2.7.1
218 (<http://broadinstitute.github.io/picard>) to remove duplicated reads. Reads were then locally realigned
219 around potential INDELs with GATK v3.7 tools RealignerTargetCreator and IndelRealigner. Quality
220 control metrics of the whole genome sequenced data were quantified and visualized by MultiQC (51).
221 Next, SNP data was generated from processed bam files using a customized SNP annotation pipeline.
222 Briefly, raw SNPs and INDELs were identified with HaplotypeCaller followed by GenotypeGVCFs tools
223 from the GATK (52). All SNPs having either QD (Quality of Depth) < 2, FS > 60, MQ <40,
224 MQRankSum < -12.5, or ReadPosRankSum < -8.0 were filtered out. VCF files containing the SNPs
225 that passed the filtering step were further filtered to select only those SNPs that were supported by a
226 minimum of 5 reads in at least one of the samples present in each VCF file.

227 **Genetic Diversity and Structural Variation Analyses**

228 Filtered genome-wide SNPs that differ from the reference (see Table S3) were identified using VCFtools
229 (53), and were concatenated and converted into a table file containing predicted genotypes (one
230 column per sample, one row per SNP position) using GATK tool VariantsToTable (v4.2.0.0). The table
231 was then converted into different multiFASTA files containing pseudosequences from concatenated
232 SNPs for a selected subset of samples with the shell script run_tbl2fasta_loop.sh, which converts the
233 table to FASTA format in order to generate a phylogenetic neighbor-net network tree using SplitsTree

234 v.4.17.1 (54) with 1,000 bootstrap replicates. The total number of SNPs and the nucleotide diversity (π)
235 (55) per 10 kb windows were also calculated using VCFtools and were plotted using the circos software
236 (<http://circos.ca/>) (56). To calculate copy number variation (CNV), the coverage at each base pair was
237 first calculated using genomeCoverageBed, BEDTools (50) and then combined into 10 kb sliding
238 windows using custom scripts as described previously (57). Mean and standard deviations were
239 calculated for each strain and bins were generated as “1X” if the value was up to 1 standard deviation
240 from the mean (1SD) and plotted using ggplot2 package (58) as described previously (57). The allele
241 composition in each variant SNPs per strains was plotted using a bottle brush plot as described
242 previously (59). For the SNP density statistic plots, read depth statistics were calculated with SAMtools
243 depth tool version 1.9 with the -aa parameter on. The remaining SNP statistics were estimated using
244 an in-house python script to present the median SNP density per chromosome, read depth and
245 coverage. Read coverage per 10 kb was estimated as the number of bases in a 10kb window having a
246 read depth of 1X or higher. To account for the presence of sequencing gaps, normalized SNP density
247 was defined as the number of SNPs counted in a 10kb window multiplied by 10,000 and divided by the
248 read coverage of the same window. Dots were used to represent normalized SNP density values
249 (number of high confidence SNPs / 10kb), Green lines to depict read depth / 10kb, Gray lines to
250 represent read coverage / 10kb, and red dashed lines to show median SNP density per chromosome.

251 **DEploid to Infer Mixed Infections**

252 To create an artificially mixed sample, 1 ng of each *C. parvum*, *C. hominis*, and *C. meleagridis* gDNAs
253 were added to 197 ng of gDNA of human stool sample (total = 200 ng) followed by whole genome
254 sequencing using CES-Seq methodology as described above. The number of strains and their relative
255 proportion and the haplotypes present in the mixed sample were estimated by deconvolving multiple
256 genome sequences using the software package DEploid (35). We downloaded 12 *C. parvum*, 8 *C.*
257 *hominis*, 2 *C. meleagridis*, 2 *C. ubiquitum*, and 1 *C. tyzzeri* genomes from SRA
258 (<https://www.ncbi.nlm.nih.gov/sra>) (Table S4) followed by variant calling using GATK (52) to construct
259 the reference genome. SNP positions containing heterozygote genotypes or missing data were
260 removed from the PANEL file, which was used for recalibrating the VCF file of artificial mixed infection
261 samples. DEploid was run in IBD (identity by descent) mode and the relative composition of
262 heterozygous and homozygous SNPs present in the artificial mixed sample was calculated in 10kb
263 sliding windows using custom Java scripts (59) to construct histogram plots in circos (56).

264 **Population Genetic Structure and Admixture Analyses**

265 Population genetic structure and admixture analyses were determined using the POPSICLE pipeline
266 (60) with 10 kb window block sizes and using the number of clusters set from K=1 to 15 followed by

267 calculating the optimal number of ancestries using the Dunn index (61) as described previously (60).
268 The contribution of each haplotype was then painted in a circos plot (56) with color assignment based
269 on the number of clusters (ancestries). For haplotype plots, we used an in-house python script to
270 calculate the percentage of different SNPs within a non-overlapping sliding window of 10kb between
271 the strain of interest (*i.e.* Isolate EC1, UKUB17 or NMIMR11) and the reference strains of either *C.*
272 *parvum*, *C. meleagridis*, *C. hominis* or *C. ubiquitum*. Within each of these windows, SNPs falling within
273 sequencing gaps were ignored in the calculation.

274

275 Results

276 **Cryptosporidium Capture Enrichment (CES-Seq) for Whole-Genome Sequencing**

277 To capture and sequence *Cryptosporidium* at whole-genome resolution directly from stool
278 samples, 74,973 CES-Seq RNA baits were developed, tiled end-to-end to cover ~92% of the 8.2 Mb
279 *C. parvum* reference genome CryptoDB-34_Cparvumlowall_Genome. The baits were tiled without
280 gaps and with two overlapping sequences, except within hypervariable regions, particularly telomeric
281 and sub-telomeric regions, where at least 5 overlapping sequence baits were used. Regions that were
282 comprised primarily of low complexity and repetitive regions were removed. Genome coverage of the
283 final bait set is listed in Table 1, each chromosome had a different coverage and this ranged from as
284 low as 87.5% to as high as 94.7% (Table 1). To test the sensitivity and specificity of the CES-Seq
285 method, different concentrations of *C. parvum* gDNA (50ng to 0.0001ng) were spiked into human stool
286 gDNA to generate a 200ng artificial gDNA mixture (Fig.1). DNA extraction using the DNeasy PowerSoil
287 Pro kit was determined empirically to produce high yield, high molecular weight DNA preps to optimize
288 shearing of gDNA into 200-300 bp fragments using the Covaris platform (Suppl. Fig. 1). As a guide to
289 identify viable clinical samples suitable for the CES-Seq method, a qPCR assay at the 18S rRNA gene
290 was developed and applied against the artificial mixed gDNA samples to determine the threshold C_T
291 associated with high-quality whole-genome sequencing. *Cryptosporidium parvum* gDNA was serially
292 diluted tenfold in human stool gDNA from 10 ng/ μ l to 0.0001 ng/ μ l, followed by qPCR (Suppl. Fig 1B
293 and C).

294 We first determined the lower limit for optimal genome wide analyses using the CES-Seq method
295 against the artificial mixtures of *C. parvum* DNA, as well as capture enrichment efficiency, the
296 percentage of reads that aligned, and the depth of coverage after trimming and quality filtering paired-
297 end reads that aligned against the *C. parvum* Iowa II reference genome. The percentage of aligned
298 reads was greater than 98% and depth of coverage >200X for all artificial mixtures that contained 5ng
299 or greater of *C. parvum* gDNA, presumably due to saturation of the baits at high concentrations of target

300 gDNA (Table 2). Notably, 0.01ng of *C. parvum* gDNA (0.005%), which corresponded to a C_T of 24.36,
301 was optimal for genome wide analysis with at least 38X coverage of the reference genome, at the depth
302 of sequencing used (Table 2). Further, linearity assays based on titrating DNA showed an ~1,500-fold
303 enrichment from a starting concentration of 0.1 ng of *C. parvum* gDNA spiked into 200 ng of total human
304 stool gDNA. Hence, C_T values equal to or lower than 24 (0.01 ng/ μ l *C. parvum* gDNA) were determined
305 empirically as optimal for achieving greater than 30X genome coverage for SNP calling and genotyping
306 at WGS resolution (Fig.1B, C) (Table 2). However, *C. parvum*-specific reads were also recovered at
307 0.001ng of input DNA (C_T of 28), and represented ~28 Mb of *C. parvum* DNA sequenced that was
308 distributed genome-wide. Although it was considered lower-confidence because the majority of
309 sequences obtained were at only 2-3X coverage, it nevertheless represented a significant improvement
310 over MLST-based genotyping strategies that recover sequences for only 5-10kb of the genome.

311 **Suitability of *C. parvum* Probes to Capture Other *Cryptosporidium* Species Genomes**

312 To test the efficiency of the *C. parvum* baits to capture other medically important *Cryptosporidium*
313 species at genome-wide resolution, we generated artificial gDNA samples by adding 1 ng of either *C.*
314 *parvum*, *C. hominis* or *C. meleagridis* gDNA into 199 ng of human stool gDNA. We next submitted
315 these different *Cryptosporidium* species gDNA samples to the capture hybridization protocol to produce
316 libraries that were sequenced using the Illumina platform (Fig. 2A). All paired-end reads were filtered
317 and aligned against the *C. parvum* reference genome for variant calling using Genome Analysis Toolkit
318 (GATK) (52). 37,367 total single nucleotide polymorphisms (SNPs) were identified between *C. parvum*
319 and *C. hominis*, and 345,437 between *C. parvum* and *C. meleagridis*. The SNPs were distributed evenly
320 throughout the genome (Fig. 2B) and were used subsequently for phylogenomic analysis (Fig. 2C). For
321 the phylogenomic comparison, we downloaded two previously published WGSs for each species from
322 SRA database (*C. hominis*: ERR311205, ERR311209; *C. parvum*: ERR1738341, ERR1035621; and
323 *C. meleagridis*: SRR1179185, SRR793561). Additionally, *C. cuniculus* (SRR6813608) and *C. tyzzeri*
324 (SRR5683558) WGSs were analyzed to generate an unrooted phylogenetic tree. CES samples
325 clustered together with publicly available WGSs, *C. parvum*, *C. hominis* or *C. meleagridis* gDNA,
326 establishing the feasibility of applying the *C. parvum*-derived baits against more distantly related
327 *Cryptosporidium* parasites to genotype samples at WGS resolution, such as *C. meleagridis* (Fig. 2C).

328 To represent the genome-wide distribution of amplified reads by the CES-Seq method, we
329 generated Bottlebrush plots (59) to display the allele count distribution at each SNP position of *C.*
330 *parvum*, *C. hominis*, and *C. meleagridis*. The plots showed that the reads had a uniform distribution
331 along all eight chromosomes of these three species (Fig. 2D), closely resembling the SNP distribution
332 plot (Fig. 2B) and confirming the high degree of specificity of CES technology. Next, to evaluate whether
333 the CES-Seq method was able to resolve genome-wide structural variation, copy number variation

334 (CNV) was calculated by estimating the combined coverage/base pair in a 10 kb sliding window. CNV
335 estimation using raw reads from CES samples showed significant changes across the genome within
336 *C. parvum*, *C. hominis* and *C. meleagridis*, particularly in subtelomeric regions (Fig. 2E).

337 It is often the case, that in endemic regions, individuals with cryptosporidiosis are infected with
338 multiple strains of *Cryptosporidium* that possess varying levels of relatedness (62). To determine the
339 capability of the CES-Seq method to resolve a mixed-species infection, and to detect the relative
340 proportion of the different species present in a sample, we next generated a mock community by adding
341 1 ng each of *C. parvum*, *C. hominis*, and *C. meleagridis* gDNA to 197 ng of human stool gDNA to
342 perform WGS-CES protocol (Fig. 3A). In order to deconvolute the mixed infection, aligned raw reads
343 were analyzed using DEploid, which uses haplotype structure with a reference PANEL of cloned
344 isolates of defined haplotypes to reference map against in order to calculate allele frequencies (35,36).
345 In this case, the reference PANEL used was comprised of 12 *C. parvum*, 8 *C. hominis*, 2 *C. meleagridis*,
346 2 *C. ubiquitum*, and 1 *C. tyzzeri* genomes (Table S4). DEploid detected the presence of all three
347 species in the mock mixed sample with a ratio of 43% *C. parvum*, 50% *C. hominis*, and 7% *C. meleagridis*
348 (Fig. 3B). To investigate whether the lower relative proportion of raw reads captured from
349 *C. meleagridis* (at 7%) was related to the higher degree of genetic polymorphism between *C. meleagridis*
350 and *C. parvum*, we assessed the genome-wide distribution of heterozygous versus
351 homozygous SNPs in blocks of 5 kb across each chromosome. The hetero-homozygosity sliding
352 window plot (Fig. 3C) corresponded closely with the DEploid estimation, because a significant
353 proportion of SNPs (resolved as blocks of yellow within the circos plot) within the *C. meleagridis*
354 genome plot were filtered out prior to conducting the DEploid analysis. However, the analysis pipeline
355 did establish that *C. meleagridis* specific SNPs were detected with sufficient frequency, with genome-
356 wide distribution, to validate this approach as a viable method to resolve mixed species infections in a
357 biological sample.

358 **359 Utility of CES-Seq to Assemble *Cryptosporidium* at WGS Resolution Directly from Human Stool
Samples**

360 To investigate the capacity of CES-Seq methodology to capture genome-wide *Cryptosporidium*
361 spp. SNP variation directly from previously frozen or ethanol-preserved clinical stool samples without
362 any prior oocyst purification or enrichment among isolates, we tested human stool samples from
363 Colombia (N=79), Ecuador (N=12), and Egypt (N=24) that had been determined previously to be
364 *Giardia*-positive and were obtained from healthy adults with no gastro-intestinal clinical symptoms (42)
365 (Fig. 4A). To determine which of the stool samples were qPCR positive for *Cryptosporidium*, we applied
366 modified GEMS primers (63,64) to screen for the presence of the following three enteroparasites:
367 *Cryptosporidium*, *Giardia* and *Entamoeba histolytica* (Table S1). All samples except one (ID-2) were

368 positive for *Giardia*, as expected. The presence of *Giardia* gDNA served as an important control to
369 investigate the specificity of the CES-Seq method to show that it specifically pulls out *Cryptosporidium*
370 DNA in the context of a mixed infection with other enteroparasites unrelated to *Cryptosporidium*. Four
371 samples were positive for the presence of *E. histolytica*, 3 from Colombia, and 1 from Egypt, with a
372 prevalence rate of 3.5% (4/115) (Table S2). Four samples gave positive C_T values for *Cryptosporidium*,
373 1 from Colombia, 2 from Ecuador, and 1 from Egypt, with a prevalence rate of 3.5% (4/115) (Table S2).
374 The Colombia sample was only weakly positive for *Cryptosporidium* (COL3, C_T =39.35) and *Giardia*
375 (C_T =36.15), but was strongly positive for *E. histolytica* (COL3, C_T =18.19). Due to the high C_T for
376 *Cryptosporidium*, we excluded COL3 but selected EC1, EC4 and FEgypt samples for CES-Seq.
377 Additionally, we screened 20 *Cryptosporidium*-positive samples, 10 from Ghana and 10 from the UK,
378 that had been previously tested by qPCR (Table S2). We selected three human fecal samples from
379 each country, Dg045 (C2), Dg083(C8), NMIMR11, UKP196, UKH101 and UKUB17 for CES-Seq based
380 on their low C_T values, the *Cryptosporidium* species expected to be present, and gDNA quality
381 (assessed by TapeStation).

382 To identify the *Cryptosporidium* species present in the samples, we performed nested PCR
383 within the 18S rRNA gene on the 9 samples and Sanger sequenced the amplicons. However, only
384 NMIMR11, UKP196, UKH101, EC1, EC4 and UKUB17 had good sequences (QV20+ \geq 20 for majority
385 of bases sequenced, with average signal to noise ratios for each base called >100). Partial 18S
386 sequencing from these samples were aligned against 18S rRNA sequences from different species of
387 *Cryptosporidium* obtained from NCBI to construct a maximum-likelihood phylogenetic tree (Fig. 4B).
388 The phylogenetic tree demonstrated that EC1 is closely related to but distinct from *C. meleagridis*
389 (AF112574), EC4 is related to but distinct from *C. canis* (AF112576), NMIMR11 and UKP196 clade
390 with *C. parvum* (AF108864), UKH101 to *C. hominis* (DQ286403.1) whereas UKUB17 is closely related
391 to but distinct from *C. ubiquitum* (HM209366.1).

392 We next performed CES-Seq to assemble the genomes of the 9 *Cryptosporidium*-positive
393 human clinical samples collected from different regions of the world (Fig. 4A). After capture
394 hybridization, we produced libraries for each sample that were sequenced using the Illumina platform.
395 All paired-end reads were filtered and aligned against the *C. parvum* Iowa-ATCC v54 reference genome
396 to determine read coverage and read distribution. Two samples EC4 (*C. canis*) and FEgypt had only
397 2141 (0.014X coverage) and 103,590 (0.98X coverage) reads map to the reference genome,
398 respectively (Table 3). Whereas samples C2 and C8 each had greater than 500,000 reads map (2.12X,
399 2.14X coverage, respectively), however, a majority of the reads from the sequencing run did not map,
400 indicating a lower yield of *Cryptosporidium* DNA from the capture hybridization reaction (Table 3). In
401 contrast, samples NMIMR11, UKUB17, UKH101, UKP196 and EC1 each possessed higher coverage,

402 ranging from 4.7x to 531X (Table 3) with the number of reads mapping specifically to the *C. parvum*
403 reference genome ranging from a low of 2.4% (UKUB17) to a high of 99% (NMIMR11). Importantly, for
404 samples EC1, EC4, and FEgypt, which were coinfecte^d with *Giardia* ($C_T = 28.3, 28.6, 27.7$,
405 respectively), essentially no reads mapped to *Giardia*, confirming the specificity of the probes and the
406 fidelity of the CES-Seq method to specifically pull out *Cryptosporidium* gDNA (Table 3). While the
407 coverage and percentage of mapped reads generally tracked with the qPCR C_T (lower C_T corresponded
408 to higher coverage and percentage of mapped to unmapped reads), this was not always the case, see
409 for example sample EC4 that had a C_T of 29.47, but only 0.014X coverage versus sample FEgypt with
410 a C_T of 35.46 but 0.979X coverage (Table 3); or sample EC1 with a C_T of 27.15 and read coverage of
411 85.56x versus sample C2 with a similar C_T of 26.46 but only 2.12X coverage. How sample
412 storage/preservation (i.e., freezing, ethanol fixation) or the species of *Cryptosporidium* present
413 influences the success of the CES-Seq methodology is currently being evaluated, but may represent a
414 significant variable that needs to be factored when performing this technique.

415 After reference mapping the reads, we used the Genome Analysis Toolkit (GATK) to call SNP
416 variants across the genomes of the seven clinical samples that had coverage equal to or greater than
417 2X. SNP variants were filtered to select for only high confidence SNPs defined as those SNPs that
418 were supported by at least 5 reads in at least one of the 7 samples in the VCF (see Methods). We next
419 constructed hierarchy-based phylogenetic trees using only the high confidence SNPs identified after
420 reference mapping the CES-Seq biological samples above against SNPs identified in the reference
421 sequences that were comprised of 12 whole-genome sequences from different human-infective species
422 that have been published previously (SRA database), specifically *C. parvum* (ERR1035621,
423 ERR1738341 and ERR1760143), *C. hominis* (ERR311205, ERR311209 and ERR363534), *C.*
424 *meleagridis* (SRR1179185 and SRR793561), *C. ubiquitum* (SRR7895345 and SRR7895268), *C.*
425 *cuniculus* (SRR6813608) and *C. tyzzeri* (SRR5683558). The raw reads from these reference samples
426 were likewise mapped against the *C. parvum* Iowa-ATCC v54 genome to call SNP variants. However,
427 due to a wide range in the distribution and read coverage between the CES-Seq samples and reference
428 samples, too few high confidence SNPs were identified when samples were analyzed altogether.
429 Hence, it was more informative to produce individual trees for each CES-Seq sample. To do this, we
430 developed individual SNP files after reference mapping each CES-Seq sample to increase the total
431 number of high confidence SNPs (without any gaps) available for phylogenetic classification of each
432 biological sample against a reference set of previously published genomes (Table S4). For EC1
433 329,989 high quality SNPs were identified to construct a phylogenetic tree using the reference
434 genomes. Correspondingly, 432,372 SNPs were identified for UKP196, 58114 SNPs for UKUB17,
435 329989 SNPs for EC1, 24,716 SNPs for C2, and 14,582 SNPs for C8 network trees (Fig. 4C, Suppl.

436 Fig. 2). This approach better resolved the phylogenetic position of each captured genome dataset within
437 the context of different *Cryptosporidium* species to visually depict their evolutionary history and
438 determine the extent to which recombination played a role in their origin. Notably, the hierarchy trees
439 were largely congruent with the 18S rRNA maximum-likelihood tree in Fig. 4B, specifically, that UKP196
440 and NMIMR11 clustered closely with *C. parvum*, UKH101 with *C. hominis*, whereas EC1 was distinct
441 from *C. meleagridis*, and UKUB17 was distinct from *C. ubiquitum* at WGS resolution and was
442 recombinant (Fig. 4C). Of note, the *C. hominis* sample ERR363534, with Biosample ID ERS226604,
443 clustered with *C. parvum* (Fig. 4C, Suppl. Fig. 2). The most parsimonious explanation for this result is
444 that it was named erroneously in the Wellcome Pilot Study performed to sequence diverse *Giardia* and
445 *Cryptosporidium* isolates.

446 **Genetic Population Structure of Human Clinical Samples**

447 To produce a robust model for the evolutionary history and population genetic structure
448 fingerprint for each clinical sample that underwent CES-Seq, it is imperative to know the relative read
449 distribution, coverage, and whether any bias was introduced during the amplification step after capture
450 hybridization. This is to ensure that all SNP variants called are of high confidence to support the genetic
451 models and to inform on the possibility to perform GWAS studies using sequence datasets from this
452 technique. To visualize this, we developed a number of custom analysis pipelines to generate SNP
453 density statistic plots for the datasets assembled. For each of the CES-Seq datasets from the clinical
454 samples investigated, we generated read depth statistics using SAMtools and plotted these values in
455 in order to visualize 1) the median SNP density per chromosome, 2) the read depth per 10kb window,
456 3) the read coverage percentage per 10kb window, and 4) the normalized SNP density per 10kb window
457 (Fig. 5). To confirm the utility of these pipeline scripts to visualize SNP distribution and depth across all
458 8 chromosomes in 10kb windows, we utilized the artificial *C. parvum*_1ng Bunchgrass sample mixture
459 in human stool gDNA as a control for the analysis pipeline. All CES-Seq Illumina reads were first
460 mapped to the *C. parvum* Iowa-ATCC v54 genome. Read depth per 10kb was at or above 50X and this
461 was depicted using a green line (Fig. 5) except in a very few focal regions on Chromosomes 2, 3, 4, 5,
462 6, and 8 (comprised of less than 100kb of sequence total) where read depth dropped to 10-30X, which
463 corresponded to regions mis-assembled between the Iowa II v34 (the genome used to pick RNA baits)
464 vs. Iowa-ATCC v54 (the genome used for reference mapping CES-Seq datasets), or undergoing copy
465 number variation (Fig. 2D, 2E). Differences in the Bunchgrass vs. Iowa-ATCC genome may also result
466 in a failure of some CES-Seq probes to capture sequences within these variant regions. Identification
467 of genome blocks that contain high confidence SNPs is tantamount to facilitate prospective GWAS
468 studies. To identify the genome-wide distribution and percentage of SNPs that are high confidence for
469 each CES-Seq dataset, we calculated read coverage (depicted with a grey line) per 10kb to identify the

percentage and distribution of SNPs that are considered high confidence based on read depth within each 10kb block (Fig. 5). Then, for each 10kb window, a normalized SNP density was calculated and plotted to identify the number of SNPs in each block that are both high confidence, and different from the reference genome, which were plotted using a blue dot. Finally, the red line for each chromosome represents the median SNP density compared to reference for each CES-Seq dataset, to estimate the extent of diversity from the reference genome, per chromosome. The integration of these SNP density statistic plots generates a confidence statistic for each dataset to inform on the degree of admixture, potential for mixed infection, and divergence from reference genomes for each CES-Seq dataset of *Cryptosporidium*. This methodology was then applied to each CES-Seq dataset for each of the 9 clinical samples investigated. *Cryptosporidium parvum* sample NMIMR11 from Ghana was highly similar to the Bunchgrass isolate, but clear differences in the SNP statistic per 10kb (blue dots) were readily visualized (Fig. 5). Additional CES-Seq datasets, including UKP196, UKH101, UKUB17, and EC1 each possessed sufficient read depth and read coverage (%) to identify high-confidence SNPs that were distributed genome-wide. Importantly for each 10kb window, there were sufficient variant SNPs detected that were distributed genome-wide to confer confidence that phylogenomic analyses to determine the evolutionary history, degree of admixture, and the ability to distinguish single strain from mixed infections is possible using the CES-Seq method. Further, for samples where genome-wide coverage was low - between 0.5-5X (samples FEgypt, C2, C8) - high confidence SNPs were still identified genome-wide, with sufficient density to infer phylogenetic ancestry. Only the EC4 CES-Seq dataset was insufficiently resolved to facilitate onward phylogenomic characterization (Fig. 5).

To determine the ancestry and relatedness within the population genetic structure of *Cryptosporidium* for each CES-Seq genome level dataset, we first needed to resolve whether the high confidence SNPs identified predicted a mixed-species infection. We used DEploid to identify the number of haplotypes and their relative frequency present in each CES-Seq dataset. Each VCF file was mapped against a reference PANEL of 12 *C. parvum*, 8 *C. hominis*, 2 *C. meleagridis*, 2 *C. ubiquitum*, and 1 *C. tyzzeri* genomes (Table S4) to determine weight supported allele frequencies, plotted as histograms, to resolve the number of haplotypes present, their relative frequency, and hence, whether each CES-Seq dataset was from a single isolate. As proof-of-principle, we re-ran the DEploid analysis pipeline using the raw read sequences from the mixed species mock community generated by spiking an equivalent amount of *C. parvum*, *C. hominis*, and *C. meleagridis* into human fecal DNA, capture hybridized, sequenced and mapped against the updated reference genome CryptoDB-54_CparvumIowa-ATCC_Genome. As in Figure 3B, 3 haplotypes were resolved with a ratio of 58% *C. parvum*, 32% *C. hominis*, and 12% *C. meleagridis* (Fig. 6A, top left panel). We then assessed each CES-Seq dataset for NMIMR11, UKP196, UKH101, UKUB17 and EC1, samples that possessed

504 sufficient read coverage and depth to call high confidence SNPs. DEploid predicted that all 5 were
505 single isolate infections (Fig. 6A). Although NMIMR11 did possess a low frequency of minor alleles,
506 they were too few, and likely represent genetic drift or could possibly represent a mixed strain infection
507 between two highly similar *C. parvum* isolates that are not resolved by this analysis, because the vast
508 majority of SNPs mapped to a single *C. parvum* haplotype (Fig. 6A).

509 We next estimated the number of supported ancestries (K) that could be resolved among the
510 five high confidence CES-Seq genome-level datasets for NMIMR11, UKP196, UKH101, UKUB17 and
511 EC1, as well as the lower coverage datasets for C2 and C8 compared against representative genomes
512 of 4 *C. parvum*, 3 *C. hominis*, 2 *C. meleagridis*, 2 *C. ubiquitum*, and 1 *C. tyzzeri* downloaded from the
513 SRA. As controls, we also included the CES-Seq datasets generated using 1ng of *C. hominis* (TU502
514 isolate), 1ng of *C. parvum* (Bunchgrass isolate) and 1ng of *C. meleagridis* (TU1867 isolate). To do this,
515 we calculated the Dunn index (61), which supported K=7 ancestral populations for the 22 genomes
516 analyzed against each other. To visualize the shared ancestry across the different isolates and species
517 of *Cryptosporidium* analyzed, we used the POPSICLE software to cluster the genomes into seven
518 different colors and distribute them within the inner ring of the circos plot (Fig. 6B). POPSICLE also
519 calculated the number of clades present for every 10 kb window across each input genome, and each
520 clade was assigned a different color hue that was painted across the genome to resolve ancestry and
521 the degree to which recombination had impacted each isolate. This was displayed in the outer ring of
522 the circos plot (Fig. 6B). The middle ring of the circos plot estimated the percentage of each ancestry,
523 represented by a different color hue, present in each genome-level dataset. Among the genomes
524 downloaded from the SRA database, POPSICLE assigned the predicted species type recorded in the
525 SRA, except for ERR363534 that was listed as *C. hominis*, but our POPSICLE and network tree
526 analysis rather showed that it shared ancestry with *C. parvum*. Also, the highly similar *C. cuniculus*
527 genome (SRR6813608) clustered together with *C. hominis* at K=7 and was only resolved from *C.*
528 *hominis* at K=8. As expected, all three CES-Seq datasets from the artificial mixture samples
529 successfully clustered together with their respective species types (*C. parvum*, *C. hominis* or *C.*
530 *meleagridis*). Importantly, the C2 and C8 partial genomes resolved unambiguously with *C. parvum*.
531 These specimens failed to genotype using sequencing primers targeting the sensitive 18S rRNA locus,
532 but the CES-Seq partial genome datasets facilitated species assignment, highlighting the sensitivity of
533 the CES-Seq method.

534 POPSICLE identified mosaic ancestries for two human isolates EC1 and UKUB17, that had been
535 previously genotyped at the 18S rRNA locus as *C. meleagridis* and *C. ubiquitum*, respectively.
536 However, genome-wide, EC1 resolved as a genetic mosaic, that shared 80.75% of its genome with *C.*
537 *meleagridis* (darkgreen) and 18.70% of its genome with a new ancestry (orange) that had introgressed

538 throughout the genome in large haplotype blocks that resembled genetic recombination (middle ring,
539 circos plot). Likewise, UKUB17 shared 37.1% of its genome with *C. ubiquitum* (blue), however, 18.5%
540 of the genome resolved with *C. parvum* (red), 3.5% with *C. hominis* (brown), and 34.2% with a distinct,
541 new ancestry (cyan). In addition, 6.7% of the assembled genome had no reads that mapped to the *C.*
542 *parvum* reference (white). These regions were distributed in discrete haploblocks throughout the
543 genome and were largely restricted to sub-telomeric and telomeric sites (Fig. 6B). DEploid analysis
544 unambiguously assigned these two assemblies as single isolates, suggesting that the two isolates
545 recovered from two individuals likely represent inter-specific recombinants and highlight the utility of
546 the CES-Seq method to inform on the genetic diversity within zoonotic species of *Cryptosporidium*.

547 To further resolve the ancestry and distribution of the predicted admixture blocks identified in the
548 POPSICLE analysis for EC1 and UKUB17, we developed a new software pipeline that calculated the
549 percentage of different SNPs within non-overlapping sliding windows of 10kb between each CES-Seq
550 dataset compared against closely related species of *Cryptosporidium*. These values were then plotted
551 across each chromosome to resolve ancestry and transitions consistent with recombination breakpoints
552 (Fig. 7 and Suppl. Fig. 3 to 5). For EC1 and NMIMR11 (control) we used *C. parvum*, *C. meleagridis*
553 and *C. hominis* genome references; for UKUB17 we used *C. parvum*, *C. hominis* and *C. ubiquitum*
554 genome references because this sample was predicted to be *C. ubiquitum* by typing at the 18S rRNA
555 locus and possessed a proportion of *C. ubiquitum* ancestry by POPSICLE (Fig. 6B). Pair-wise SNP
556 plots for NMIMR11 resolved this sample as *C. parvum* genome-wide (magenta line for *C. parvum*
557 mapped at essentially zero percent of different SNPs per 10 kb window) with no evidence of
558 recombination. In contrast, EC1 was polymorphic, possessed clear ancestral blocks related to *C.*
559 *meleagridis* (blue line mapped close to zero percent of different SNPs per 10kb window), however it
560 also possessed large haploblocks that were divergent and unrelated to *C. parvum* (magenta line) or *C.*
561 *hominis* (green line). In the genome-wide plots, there was also evidence of 7 haploblocks that were in
562 fact *C. hominis* or *C. parvum*, suggesting that inter-specific recombination had impacted the population
563 genetics of this human infective isolate. The pair-wise SNP haplotype plots also established that
564 UKUB17 is novel, it was neither *C. parvum*, *C. hominis* nor *C. ubiquitum*, it was extensively recombined,
565 with distinct haploblocks indistinguishable from *C. parvum*, *C. hominis*, or *C. ubiquitum* introgressed
566 throughout the genome. It also possessed large haploblocks that were new, that did not BLAST with
567 any known species of *Cryptosporidium* thus far analyzed. The blocks were as similar to *C. ubiquitum*
568 as *C. hominis* is to *C. parvum*, which mapped to NMIMR11 genome-wide in the 10-20 percent of
569 different SNPs per 10 kb, which may suggest that the differences are significant enough to reflect
570 introgression of a new species-type thus far not resolved in the population genetics of *Cryptosporidium*
571 (Fig. 7 and Suppl. Fig. 3 to 5). The DEploid analysis unequivocally predicted UKUB17 was the result of

572 a single clone infection. The pair-wise SNP haplotype plots likewise showed that it was not synonymous
573 with a mixed infection, as it produced unambiguous recombination break-points. Rather, the data and
574 analysis pipelines herein supported UKUB17 to be an inter-specific genetic mosaic of mixed ancestry
575 within a single haplotype.

576

577 **Discussion**

578 In this study, we applied Capture Enrichment Sequencing (CES-Seq) to generate genome-level
579 datasets of *Cryptosporidium* associated predominantly with asymptomatic carriage directly from human
580 stool specimens. The majority of samples had been previously frozen or preserved in ethanol. We
581 demonstrated that ~75,000 RNA baits that cover ~92% of the *Cryptosporidium* genome were sufficient
582 to capture and enrich *Cryptosporidium* DNA ~1500 fold directly from stool samples with C_T below 28,
583 and for samples with C_T between 28 and 31, the ability to produce partial genomes for phylogenetic
584 analysis and genotyping far exceeded current multi-locus typing methods for the genus. We showed
585 that the CES-Seq genome-wide sequencing method is both highly sensitive and specific and
586 can successfully resolve species of *Cryptosporidium* at whole genome resolution from parasite gDNA
587 present in as little as 0.005% in human stool DNA. Further, the baits generated against *C. parvum*
588 captured and produced genome-level datasets for more distantly related species without the
589 requirement or need to purify oocysts, including *C. meleagridis*, *C. canis*, *C. ubiquitum* as well as
590 interspecific recombinants present in clinical/field samples. We also showed the applicability of applying
591 various phylogenomic pipelines, including DEploid, SplitsTree, and POPSICLE software suites to infer
592 genetic diversity, degree of mixed infection, and the population genetic structure
593 of *Cryptosporidium* species infecting predominantly asymptomatic patients. Our approach
594 demonstrates the utility and cost-effectiveness of “*in situ*” CES-Seq to generate genome level datasets
595 of *Cryptosporidium* from patient samples to effectively genotype the *Cryptosporidium* present in
596 endemic settings at whole genome resolution, and how to use phylogenomic analyses to infer the
597 potential for their zoonotic transmission, drug resistance, and capacity to cause new disease or alter
598 host range. Ultimately this technology, which enables high-throughput sequencing of samples without
599 the requirement to isolate oocysts, should empower scientists in lower-to-middle income countries
600 (LMICs) where transmission is endemic to better understand the burden of *Cryptosporidium* in their
601 countries and track how infection changes over time.

602 With approximately 22 species identified to infect people within the genus *Cryptosporidium*,
603 many of which are zoonotic and possess broad host ranges based on recent molecular epidemiological
604 studies (8), it is essential to determine the applicability of utilizing RNA baits designed against one

605 reference genome (in this case *C. parvum*) to effectively capture and enrich for other species of
606 *Cryptosporidium* that infect humans. Data herein established that these baits, which represent the first
607 set of baits developed for the *Cryptosporidium* field, possessed a broad specificity, and were capable
608 of pulling out, with high efficiency and at whole genome resolution, zoonotic species such as *C.*
609 *meleagridis*, *C. canis*, and *C. ubiquitum*, as well as interspecific recombinants. However, the RNA baits
610 in this paper were not designed to capture gene-poor sub-telomeric and telomeric regions, because
611 they were designed prior to the release of the most up-to-date telomere-to-telomere (T2T) assembly
612 that identified additional genes not present in the Iowa II v34 *C. parvum* assembly (65). Importantly,
613 the CES-Seq datasets generated using baits designed against the first Iowa II v34 assembly did map
614 with high fidelity to the new reference *C. parvum* T2T Iowa-ATCC v54 *C. parvum* assembly and our
615 ability to perform robust population genomic analyses was not impacted. With the current push within
616 the *Cryptosporidium* community to generate *de novo* genomes for a majority of the zoonotic species,
617 future studies to increase the sensitivity of this approach should focus on generating additional RNA
618 baits that span genomic regions that are either highly divergent in the other 21 species that infect
619 humans or not present in *C. parvum*. One such effort that shows great promise recently used ~ 10
620 genomes from different *Cryptosporidium* species to design a new bait set to more broadly detect diverse
621 *Cryptosporidium* species and genotypes both *in silico* and experimentally for specificity and sensitivity
622 (66).

623 Recent molecular epidemiological and genomic surveillance studies have established that the
624 genus *Cryptosporidium* exhibits a complex evolutionary and transmission dynamics, including
625 recombination events that contribute to adaptive genetic exchange between different species
626 (interspecific) and within the same species (intraspecific) of *Cryptosporidium*. Such recombination
627 events observed among zoonotic species of *Cryptosporidium* have highlighted the role of genetic
628 hybridization in the emergence and enhanced transmission of specific genotypes that possess an
629 increased capacity to cause disease or alter their host range (29-33). CES-Seq genome-level datasets
630 sequenced directly from patient samples should greatly facilitate studies aimed at understanding the
631 true genetic diversity of *Cryptosporidium* species and the role of recombination in the emergence of
632 new genotypes associated with outbreaks, or that are circulating in endemic regions. Indeed, our NGS
633 workflow identified extant recombination within the single UKUB17 specimen that typed as *C. ubiquitum*
634 at the 18S rRNA gene (Fig. 4) but was shown to be a complex mosaic possessing large haploblocks
635 with ancestry belonging to *C. parvum*, *C. hominis*, *C. ubiquitum*, and a novel sequence type (Fig. 7).
636 This capacity to resolve genomes at high-resolution directly from human stool samples should inform
637 future work that tests the hypothesis that the evolution of anthropozoonotic transmission in humans is the
638 result of acquiring pathogenicity determinants from within the pan-*Cryptosporidium* genome to make

639 human infection possible (30,31,33), analogous to the inheritance of secreted pathogenicity islands
640 that determine host adaptation and disease potential in the related apicomplexan parasite *Toxoplasma*
641 *gondii* (67,68). Importantly, the CES-Seq method and the associated population genomic pipelines
642 developed herein also makes it relatively straightforward to determine the degree to which co-
643 circulating strains (mixed infections) occur among susceptible hosts, including humans. This is
644 necessary information, for two reasons: 1) the degree to which mixed infection occurs provides an
645 estimate for the role genetic exchange plays in the evolution of genotypes that possess new host
646 preferences or the capacity for zoonotic versus anthroponotic transmission (21); and 2), the ability to
647 resolve whether the infection is by a single genotype is relevant for genotype to phenotype studies to
648 support GWAS analyses. Hence, estimation of the rate and relatedness of such mixed infections is a
649 critical factor to correctly construct the population genetic structure of *Cryptosporidium*, particularly in
650 endemic regions, where the probability of mixed infection is relatively high (69,70).

651 In our study, the CES-Seq methodology coupled with the computational pipelines we applied
652 and/or developed to analyze the assemblies produced establishes a new paradigm for resolving the
653 genome sequences present in both archived and prospectively collected biological samples. Our
654 workflow facilitates the identification of high confidence SNPs that are required for robust phylogenetic
655 resolution of derived genome datasets in the context of defined *Cryptosporidium* species to determine
656 the true population genetic structure of circulating strains, in reference to their zoonotic transmission.
657 Our computational analysis platform is also capable of resolving mixed-strain infections, determining
658 the genome structure of single strain infections, and applying new visualization tools to identify high-
659 confidence SNPs in order to envisage future GWAS studies to map specific genes that contribute to
660 phenotypic traits. Understanding the genetic basis of adaptation and admixture in *Cryptosporidium*
661 populations is crucial for elucidating the epidemiology and pathogenesis of cryptosporidiosis, informing
662 public health interventions, and guiding the development of effective control strategies.

663 In summary, the CES-Seq genome-level sequencing method is a novel, highly sensitive and
664 specific tool to conduct phylogenomic studies on *Cryptosporidium* species circulating in stool samples
665 to understand the population genetic structure of *Cryptosporidium* pan-genomes throughout the world.
666 These studies are necessary to understand not only the zoonotic potential of a strain, but also the
667 evolution and emergence of novel subtypes that impact *Cryptosporidium* transmission, host adaptation,
668 and disease potential both locally (in the context of an outbreak) and globally (in the context of a
669 selective sweep).

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692

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861

862 Author contributions

863 Study was designed by AK, EVCAF and MEG. Experiments were conducted by AK, EVCAF, and HV.
864 Samples were provided by MCP, SB, IA, GR and RMC. Analysis was conducted by AK, HL, EVCAF,

865 and HV, and the manuscript was written by AK, EVCAF and MEG.

866

867 **Competing interests**

868 The authors declared no competing financial interests.

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870

871 **Table and Figure and Legends**

872

873 **Fig. 1.** Validation of SureSelect CES of protozoan diarrheal agent *Cryptosporidium*. A) Workflow of
874 SureSelect TES. Whole genome of *Cryptosporidium* was captured by hybridization with genome wide
875 uniformly distributed biotinylated probes (Table 1). The strand extension and the capture steps allow
876 the recovery of the amplified library for NGS using Illumina sequencing. B) Graphical display of average
877 sequencing depth and coverage percentage with $\geq 10X$ for SureSelect capture enrichment followed by
878 whole genome sequencing using Illumina platform. Different dilutions of *C. parvum* gDNA in 200 ng of
879 total gDNA are plotted with different color circles as listed in the legend. C) Fraction of the reference
880 genome recovered after target enrichment using genome wide RNA baits and whole genome
881 sequencing using Illumina platform. Different dilution of *C. parvum* gDNA are depicted with different
882 color lines as listed in the legend.

883

884 **Fig. 2.** Capture enrichment sequencing of distantly related species of *Cryptosporidium*. A) Diagram to
885 depict SureSelect sequencing of *C. parvum*, *C. hominis*, and *C. meleagridis*. 1 ng of parasite gDNA
886 was introduced into 199 ng of gDNA from healthy human stool sample (Total = 200 ng gDNA). B) Circos
887 plot showed the distribution of pairwise SNPs (red) and Pi (blue) across the genome per 10kb sliding
888 windows. 1 and 2, pairwise SNP and Pi plots between *C. parvum* and *C. hominis* respectively. 3 and
889 4, pairwise SNP and Pi plots between *C. parvum* and *C. meleagridis* respectively. C) Neighbour-net
890 analysis of *Cryptosporidium* spp. based on genome wide SNPs after amplifying the whole genome
891 using SureSelect baits and Illumina sequencing followed by reference mapping the genome
892 sequences. *C. parvum*, *C. hominis*, and *C. meleagridis* were whole genome sequenced using CES
893 methodology whereas ERR311205, ERR311209, ERR1738341, ERR1035621, SRR1179185, 793561,
894 *C. cuniculus* (SRR7895182), and *C. tyzzeri* (SRR5683558) were downloaded from NCBI SRA
895 (<https://www.ncbi.nlm.nih.gov/sra>) site. Scale bar represents the number of SNPs per site. D)
896 Bottlebrush representation of genome wide distribution of amplified reads using CES. X-axis = size of
897 chromosome, y-axis = inferred read depth, normalized across the entire genome. Red = *C. parvum*,
898 Purple = *C. hominis*, green = *C. meleagridis*. E) Genome-wide CNV was determined in 10kb tiling
899 windows. The 10kb blocks with no CNV are plotted as black circles (1X). Green, blue, and red dots
900 indicate 1, 2, and 3 SDs from the mean (1X), respectively. Y-axis indicates the CNV

901

902 **Fig. 3.** Detection of mixed infection using capture enrichment sequencing. A) Diagram to generate an
903 artificial mixed sample. 1ng of *C. parvum*, *C. hominis*, and *C. meleagridis* gDNAs (in total 3ng of

904 *Cryptosporidium* gDNA) were mixed with 197ng of fecal stool sample (total = 200ng) gDNA to create a
905 mixed infection of sample with the presence of three *Cryptosporidium* species B) Estimation of
906 multiplicity of infection by genome wide abundance of heterozygous genotypes using software Deploid
907 (35). CP = *C. parvum*, CH = *C. hominis*, CM = *C. meleagridis*, Mixed = mock mixed infection sample
908 generated as described previously (Fig 3. A). X-axis indicates the percentage of abundance of species
909 in each sample. CP, CH, and CM data were generated based on Fig. 2A. C) Circos representation of
910 genome wide heterozygosity and homozygosity plots of 3 *Cryptosporidium* species and the mixed
911 sample. Red color = >90% of heterozygous SNPs, blue = > 90% of homozygous SNPs, yellow = 50%
912 heterozygous, 50% homozygous SNPs. Each track represents each sample. Chromosome numbers
913 are shown on the outer ring. CP = *C. parvum*, CH = *C. hominis*, CM = *C. meleagridis*, Mixed = mock
914 mixed sample

915

916 **Fig. 4.** Detection of *Cryptosporidium* in asymptomatic and symptomatic patient's stool samples . A)
917 Geographic locations of clinical samples used in this study show in arrow heads. Sample numbers are
918 in the parentheses. B) Phylogenetic analysis of 18S rRNA. An unrooted maximum-likelihood tree was
919 generated using Geneious- PhyML tree with 1,000 bootstrap replicates. Representative sequences of
920 different species of *Cryptosporidium* were downloaded from NCBI (<https://www.ncbi.nlm.nih.gov/>) and
921 compare with the sequenced samples (green). Bootstrap values between 75 to 100% are represented
922 by the red nodes, values between 50 to 75% are depicted by blue nodes, and values between 25 to
923 50% are denoted by yellow nodes. Scale = number of SNPs per site. C) Phylogenetic Network Trees.
924 Trees were generated with SplitsTree (v4.17.1), total number of SNPs below each tree.

925

926 **Fig. 5.** SNPs distribution of *Cryptosporidium* species directly from patient's stool samples (C2, C8,
927 NMIMR11, UKP196, UKH101, UKUB17, EC1, EC4 and FEgypt) and the *Cryptosporidium parvum*
928 spiked control (*C. parvum* 1ng). SNP density plots were generated using an in-house python script.
929 Blue dots represent normalized SNP density values by 10kb. Green line, read depth / 10kb; gray line,
930 read coverage / 10kb; red dashed line, median SNP density per chromosome (not considering
931 sequencing gaps).

932

933 **Fig. 6.** SureSelect-WGS population genetic structure of *Cryptosporidium* species. A) DEploid software
934 was used to determine if the clinical sample was from a single strain, or mixed strain infection. Genome-
935 level datasets were mapped against a PANEL of 25 different reference *Cryptosporidium* species (Table
936 S4) to determine the relative proportion of different haplotypes present in mixed-strain clinical samples

937 and plotted as bar plots. The relative proportion of each genotype was pseudocolored to reflect the
938 underlying genome present. The following pseudocolor code based on Fig. 6B was used: red for *C.*
939 *parvum*; green for *C. meleagridis*; brown for *C. hominis*; blue for *C. ubiquitum*; orange for EC1-like, and
940 cyan for UKUB17-like. The histograms in gray represent the weight supported allele frequencies
941 (WSAF) for each clinical sample. Single histograms with WSAF values between 0.9-1.0 indicate a
942 single strain infection, and the bar plot is pseudocolored to reflect the species genotype resolved. B)
943 Circos plot representation of admixture analysis and chromosome painting of *Cryptosporidium*
944 conducted by POPSICLE (6) based on an ancestral population size of K=7. Outermost concentric circle
945 represents a chromosome-level admixture profile for each strain, painted in 10 kb sliding windows and
946 pseudocolored to represent the species type resolved. The middle concentric circle plots the
947 percentage of shared ancestry. The innermost concentric circle indicates the cluster assignment by
948 color hue for a population size of K=7. Each color represents one ancestral population. Representative
949 sequences were downloaded from SRA (<https://www.ncbi.nlm.nih.gov/sra>). Whole genome sequences
950 of the clinical samples (yellow dots) were obtained after CES-Seq.

951

952 **Fig. 7.** Pair-wise SNP haplotype plots used to resolve the degree of identity of each CES-Seq examined
953 clinical sample against reference genomes from related *Cryptosporidium* species. The percentage of
954 SNPs different between the strain of interest (i.e., Isolate UKUB17, EC1, or NMIMR11) and the
955 reference strains of either *C. parvum* (magenta line), *C. meleagridis* (blue line), *C. hominis* (green line)
956 or *C. ubiquitum* (blue line) were calculated and plotted in non-overlapping sliding windows of 10kb along
957 each chromosome (x-axis, chromosome position in Mb). Colored lines at the baseline of “0” percent of
958 different SNPs per 10kb indicated that the clinical sample had 100% identity to the reference genome
959 of that color type. A) UKUB17 sample, blue - *C. ubiquitum*, magenta – *C. parvum* and green – *C.*
960 *hominis*. B) EC1 sample, blue - *C. meleagridis*, magenta – *C. parvum* and green – *C. hominis*.

961

962 **Supplemental Fig. 1.** Sensitivity and Specificity Analysis of CES-Seq Method. A) gDNA extraction
963 comparison using two different kits shows high yield using DNeasy PowerSoil Pro kit. B and C) 18S
964 rRNA qPCR assay using Serial dilutions of *Cryptosporidium parvum* gDNA in human stool gDNA.

965

966 **Supplemental Fig. 2.** Phylogenetic Network Trees of C2 and C8 samples. Trees were generated using
967 SplitsTree (v4.17.1), total number of SNPs are indicated below each tree.

968

969 **Supplemental Fig. 3.** Pair-wise SNP haplotype plot (10kb sliding window) for all chromosomes for
970 sample EC1.

971

972 **Supplemental Fig. 4.** Pair-wise SNP haplotype plot (10kb sliding window) for all chromosomes for
973 sample NMIMR11.

974

975 **Supplemental Fig. 5.** Pair-wise SNP haplotype plot (10kb sliding window) for all chromosomes for
976 sample UKUB17.

977

978 **Table 1.** Design of *Cryptosporidium* specific biotinylated baits for SureSelect target enrichment
979 sequencing.

980

981 **Table 2.** Whole genome sequencing of libraries after amplification using SureSelect Target
982 enrichment sequencing.

983

984 **Table 3.** Sample information.

985

986 **Table S1.** 18S Ribosomal RNA primers used in this study.

987

988 **Table S2.** C_T values of 18S Ribosomal RNA qPCR assay using species specific primers for
989 *Cryptosporidium*, *Giardia*, and *E. histolytica*.

990

991 **Table S3.** Chromosome ID for both genome references used.

992

993 **Table S4.** PANEL of 25 different reference *Cryptosporidium* species. Each VCF file was mapped
994 against a reference PANEL of 12 *C. parvum* (ERR1035619, ERR1035621, ERR1738340,
995 ERR1738345, ERR1738350, ERR1738337, ERR2366918, ERR961651, SRR6117460,
996 SRR6147472, SRR6147945 and SRR614796), 8 *C. hominis* (ERR1305020, ERR1305025,

997 ERR2240056, ERR2240065, ERR2240074, ERR2366927, ERR2366935 and SRR1015721), 2 *C.*
998 *meleagridis* (SRR1179185 and SRR793561), 1 *C. tyzzeri* (SRR5683558) and 2 *C. ubiquitum*
999 genomes (SRR7895268 and SRR7895345)

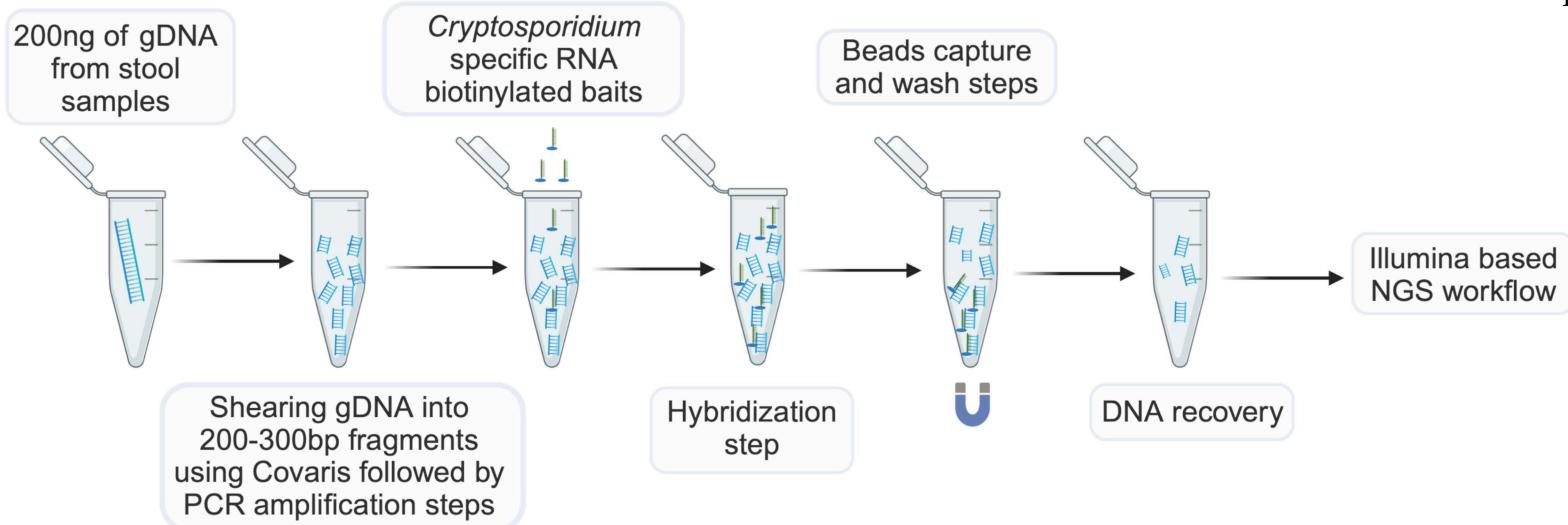
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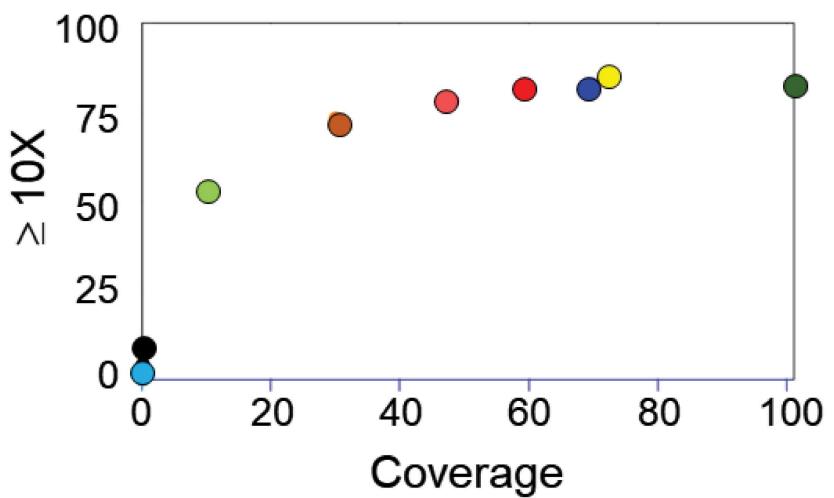
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A

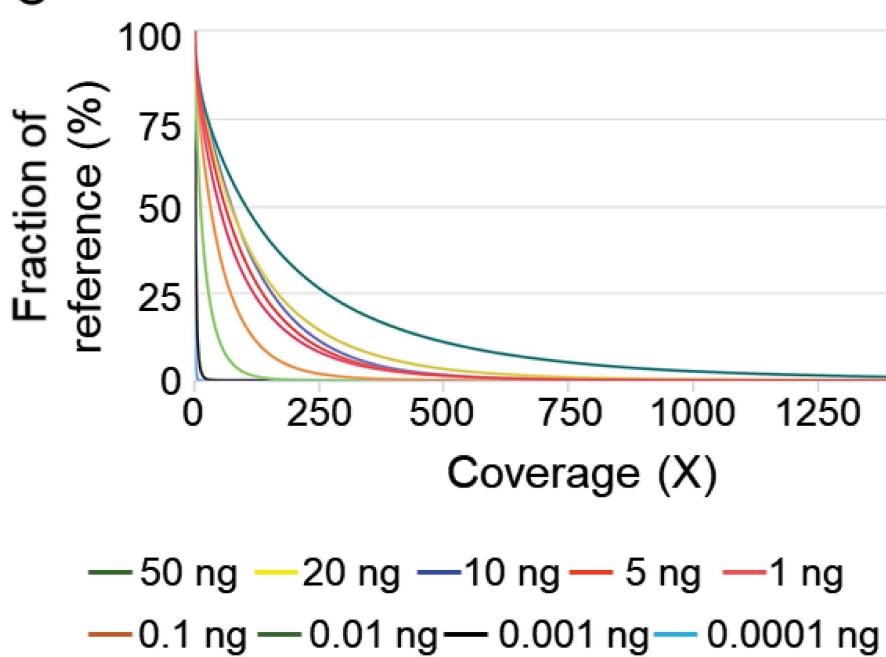


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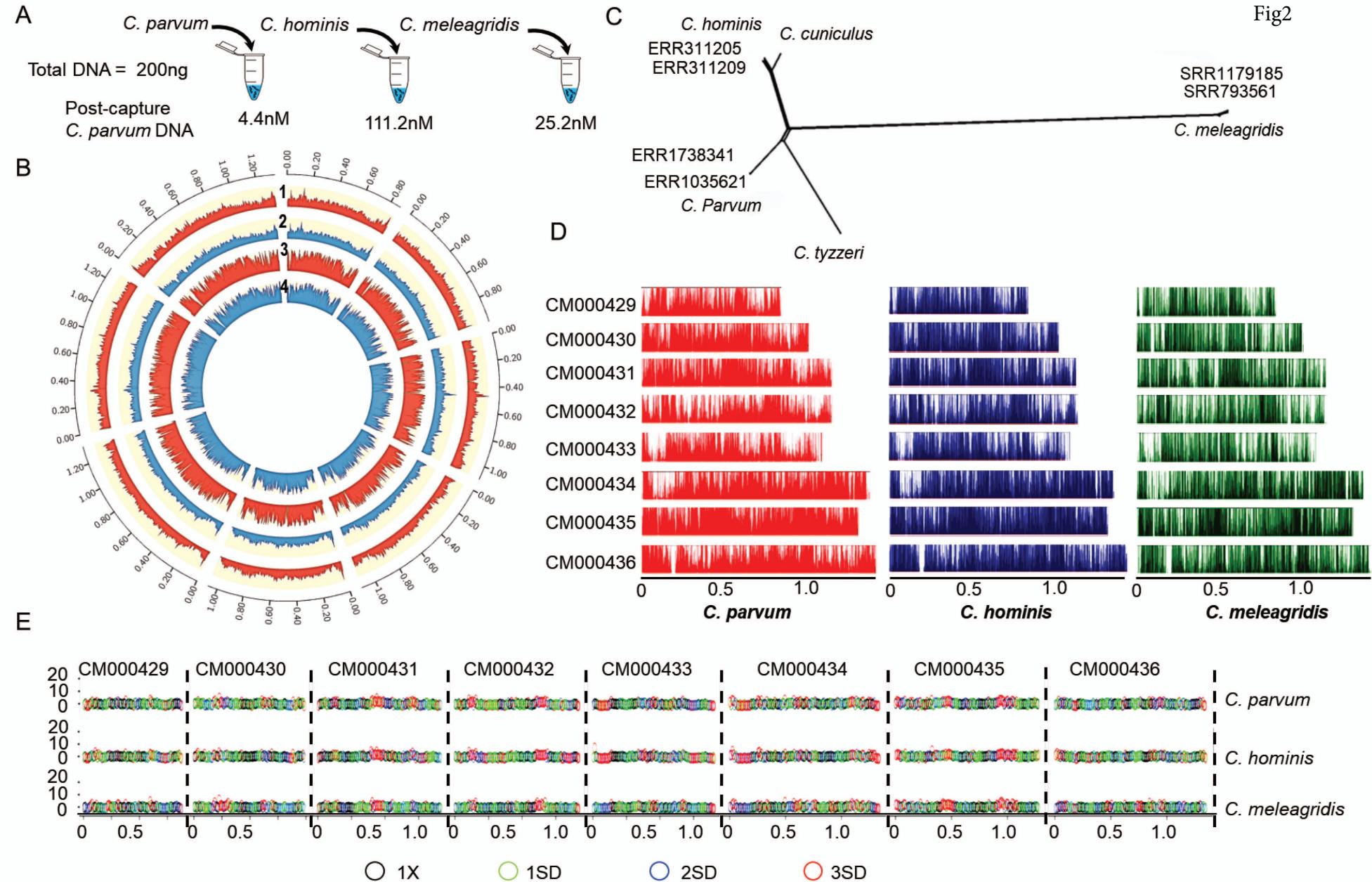
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C

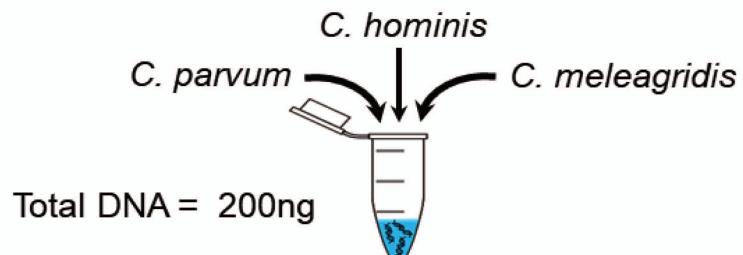


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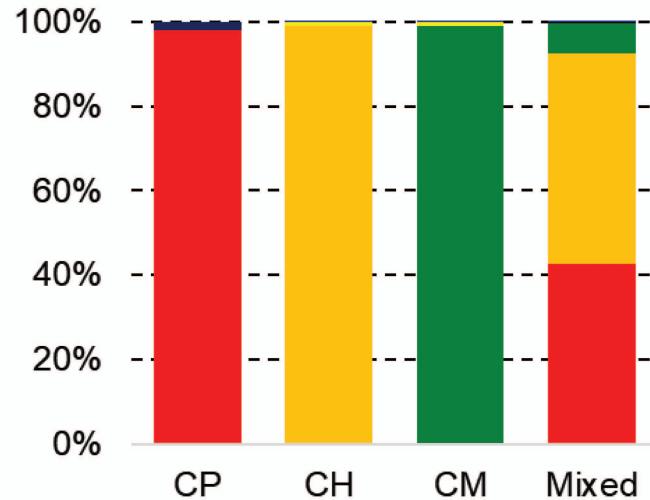
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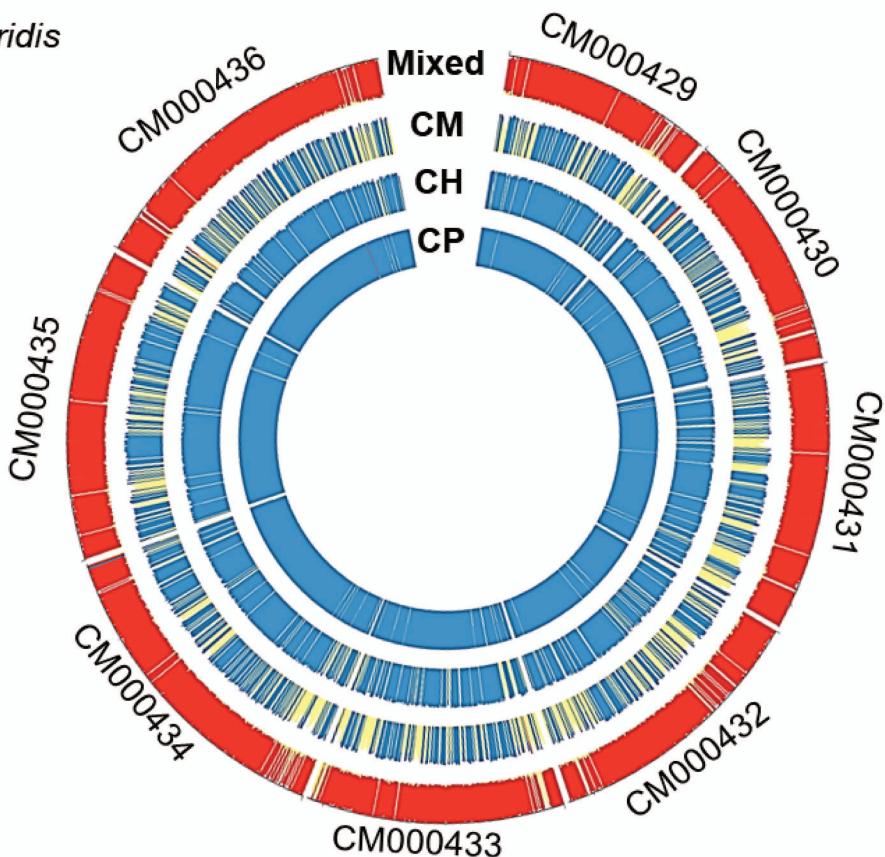
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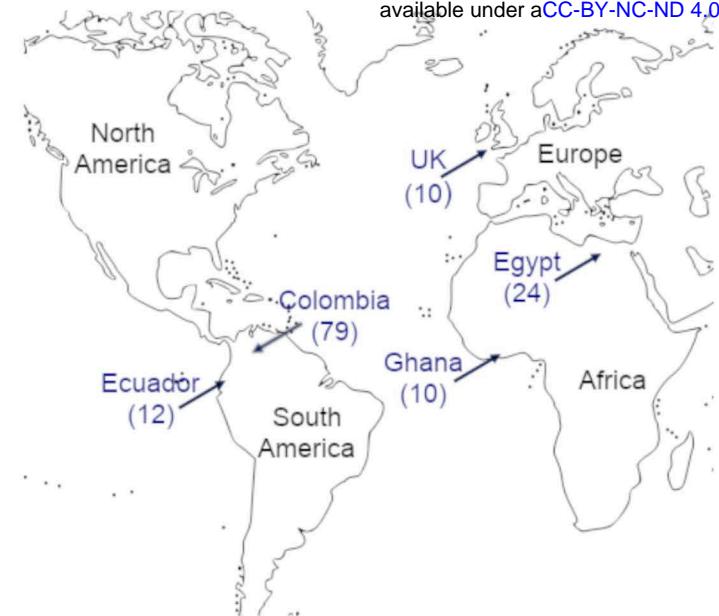


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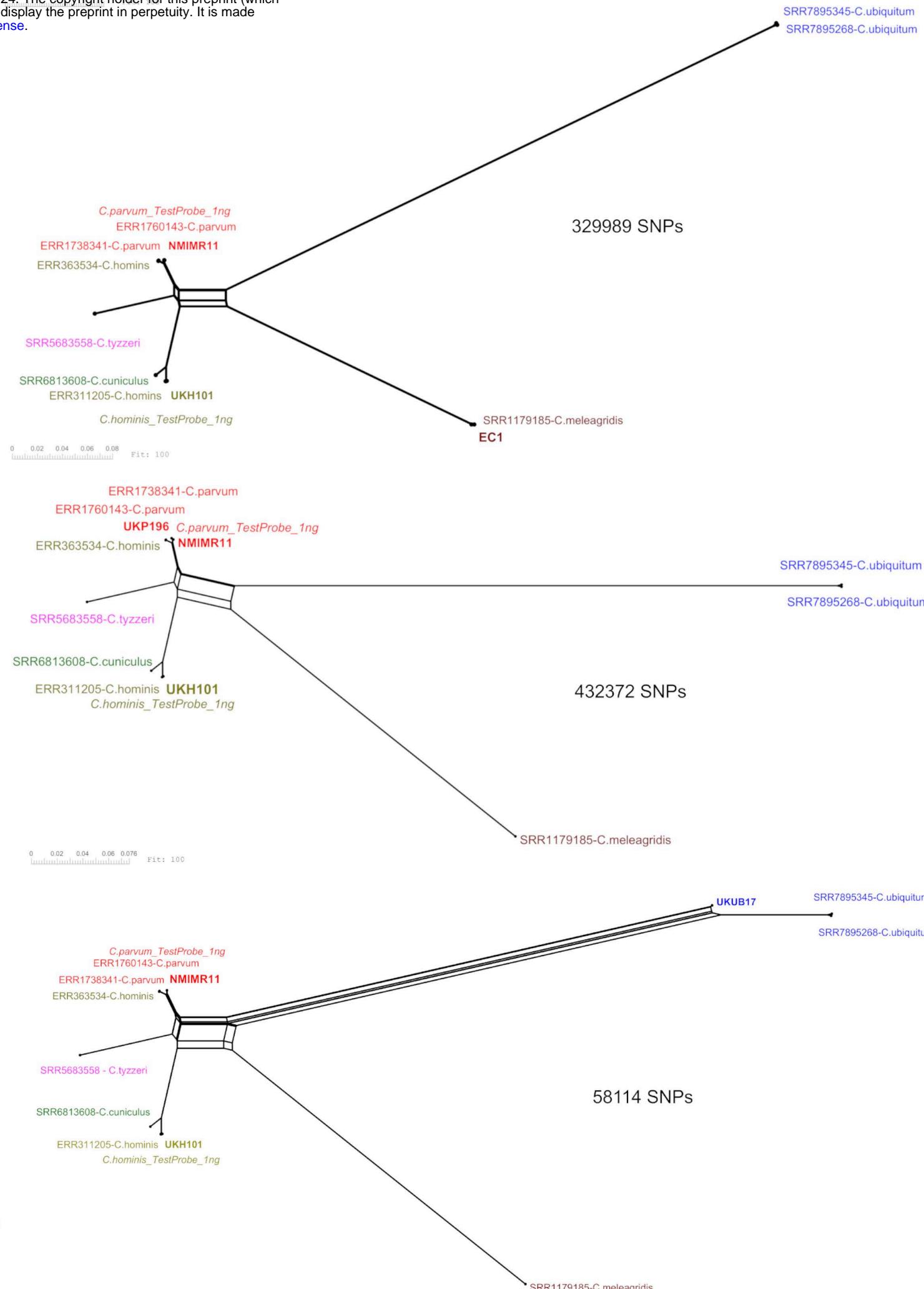


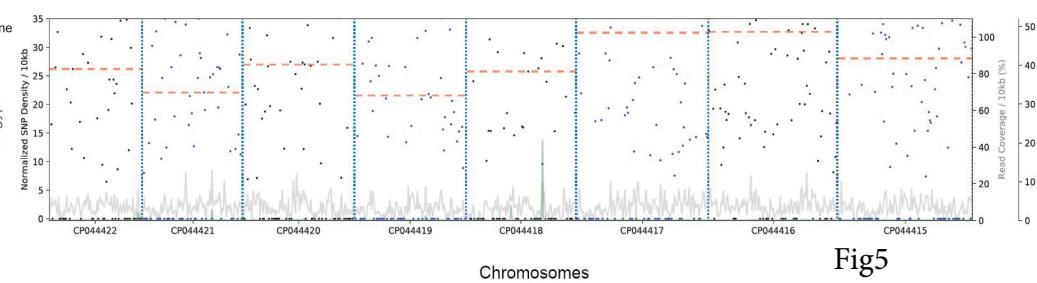
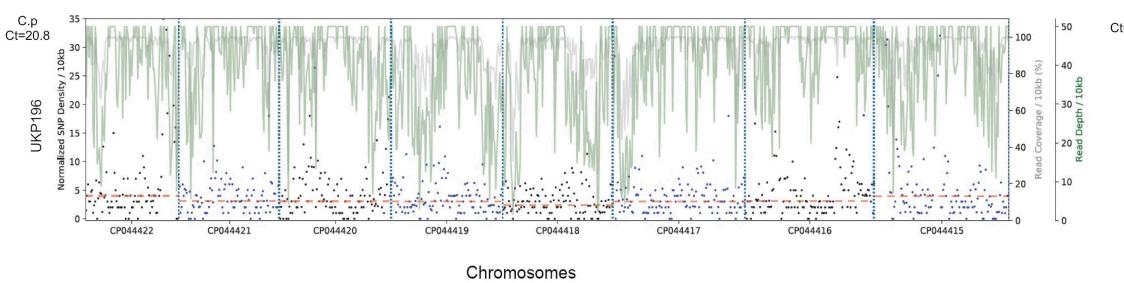
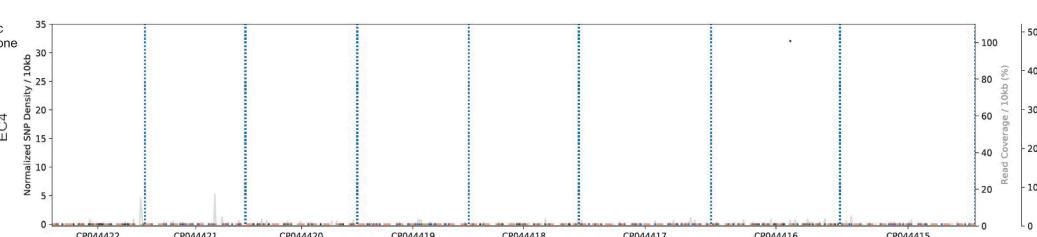
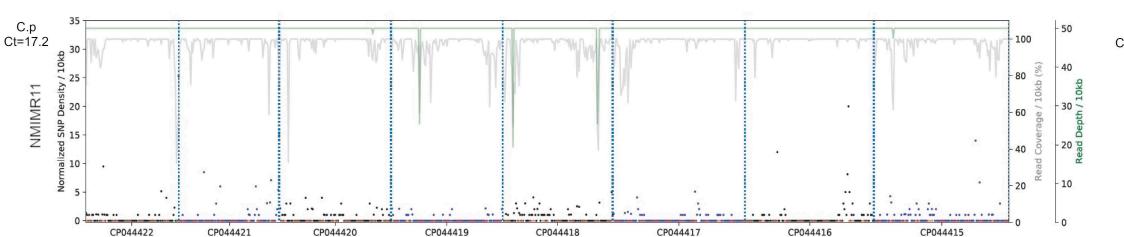
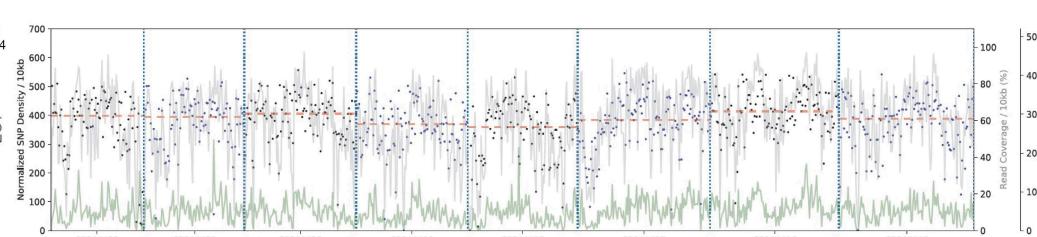
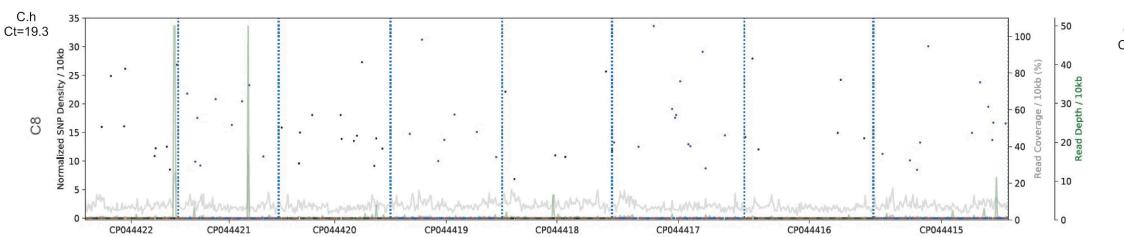
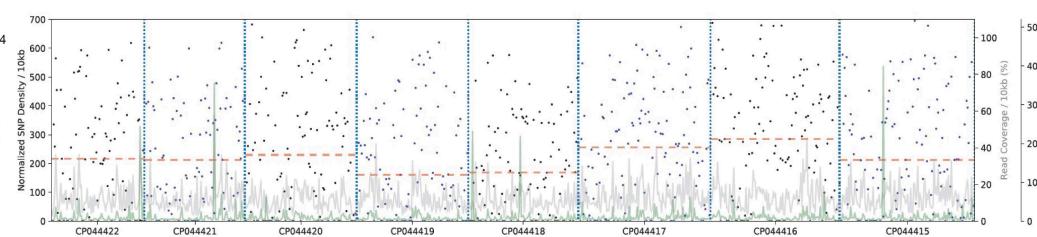
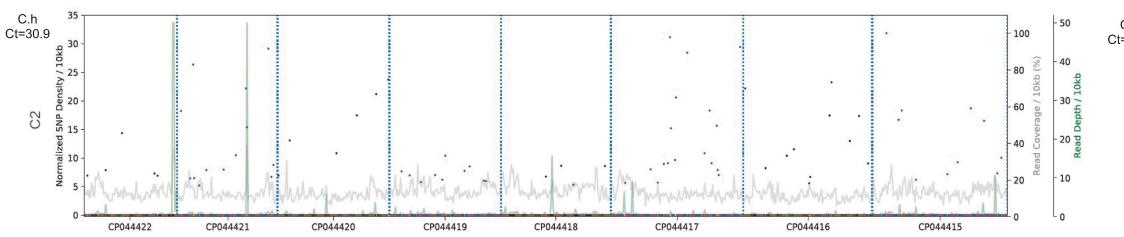
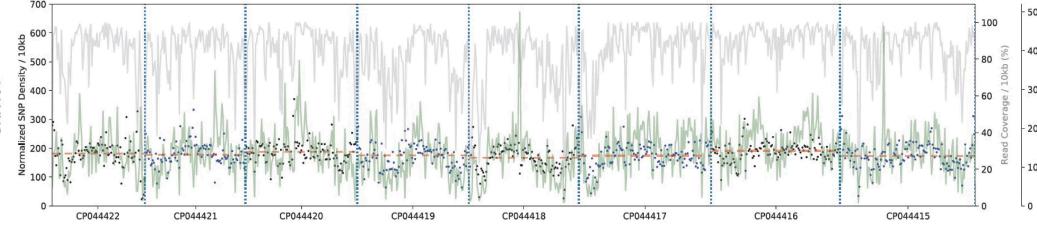
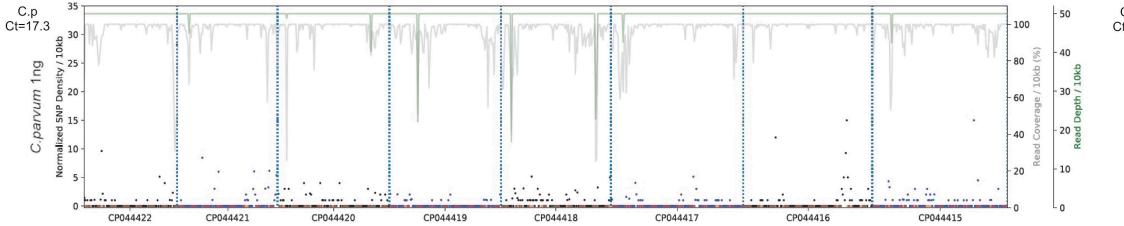
A

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C



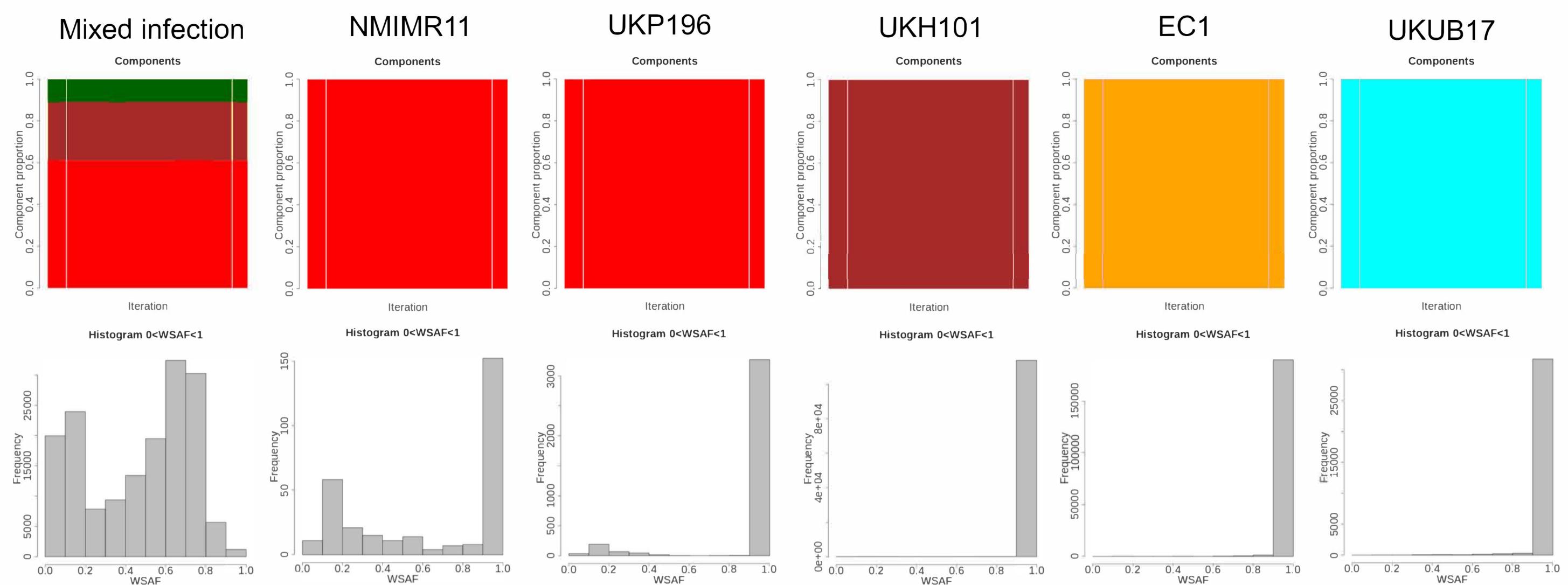


Chromosomes

Chromosomes

Fig5

A



B

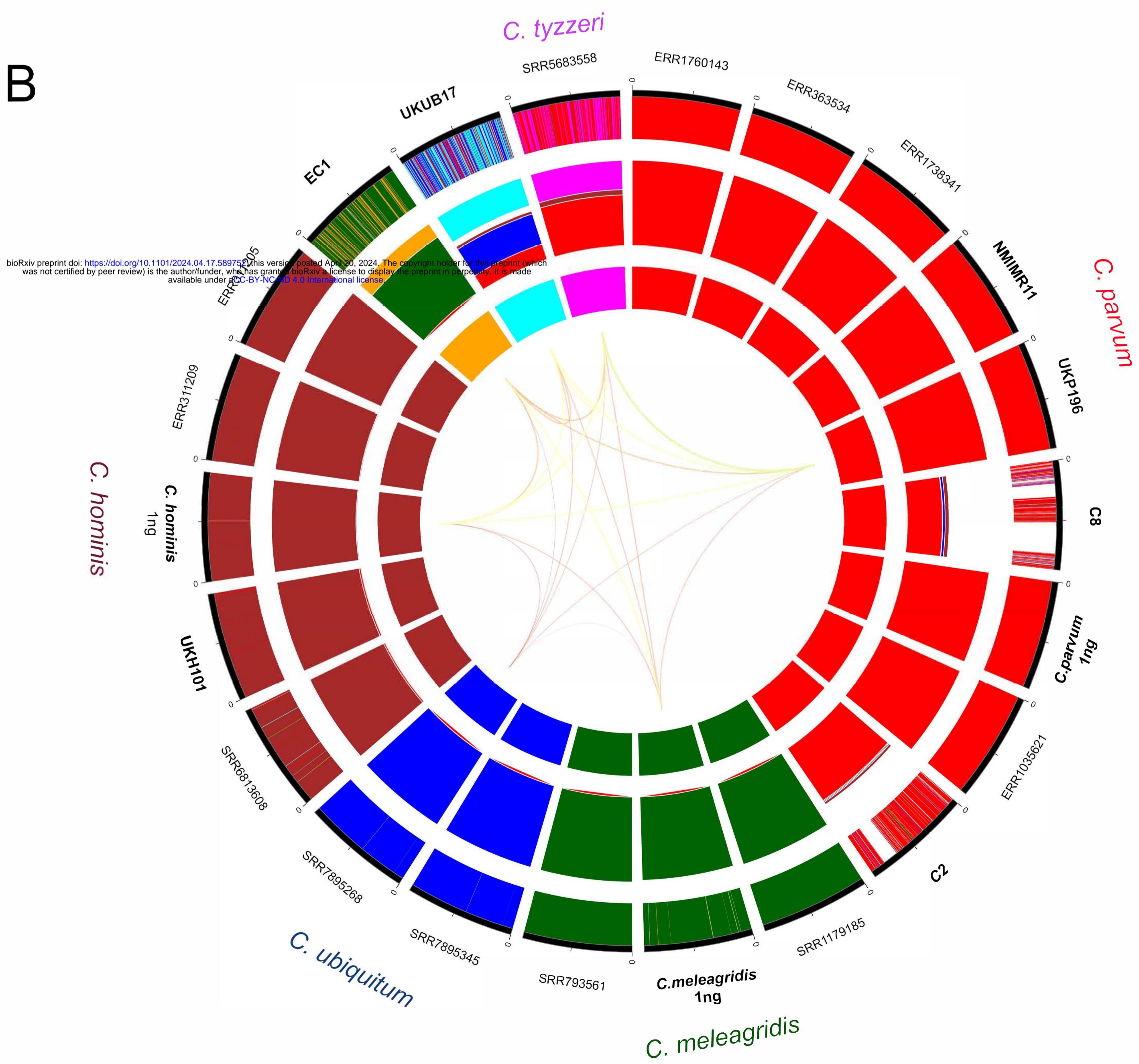
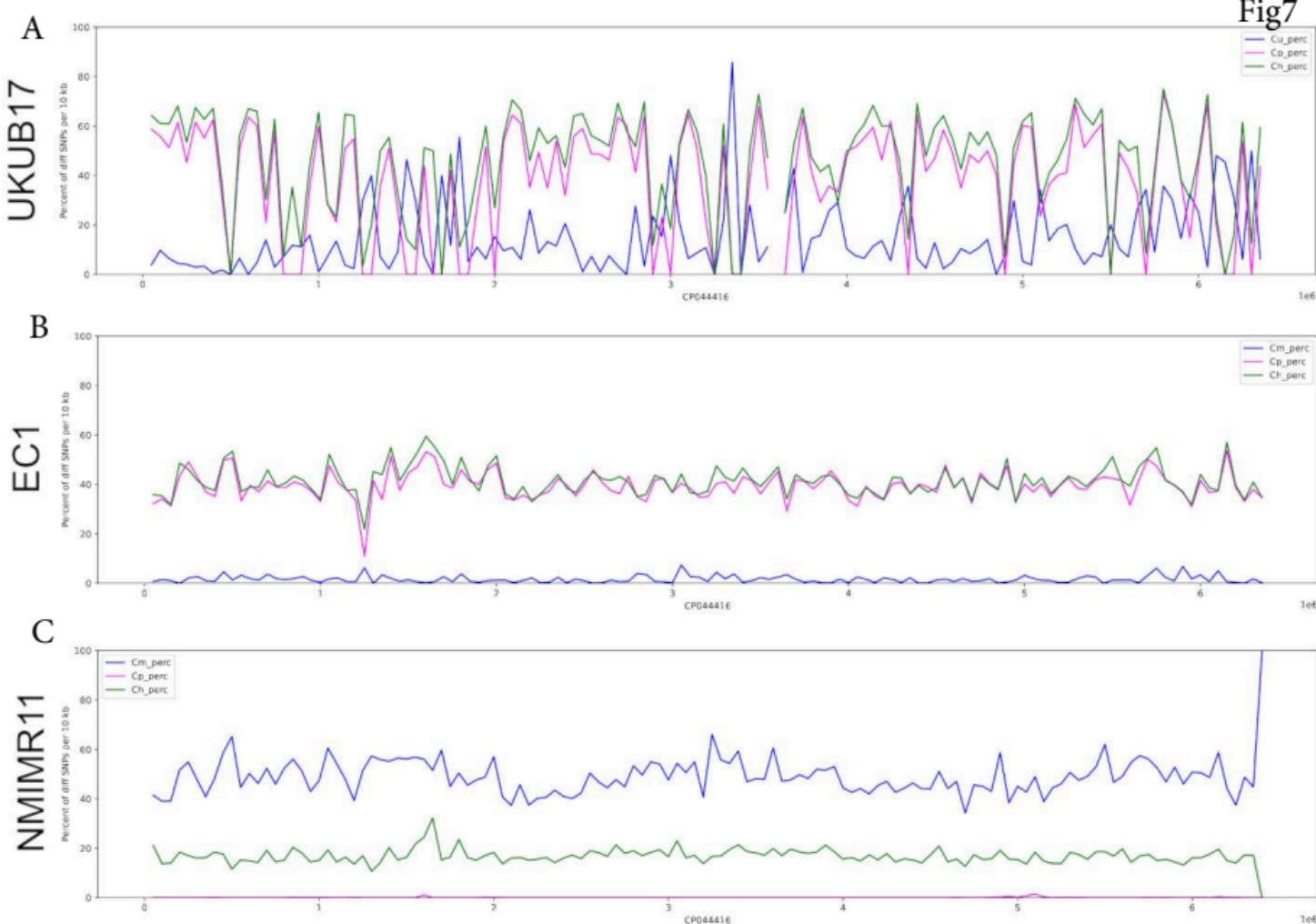
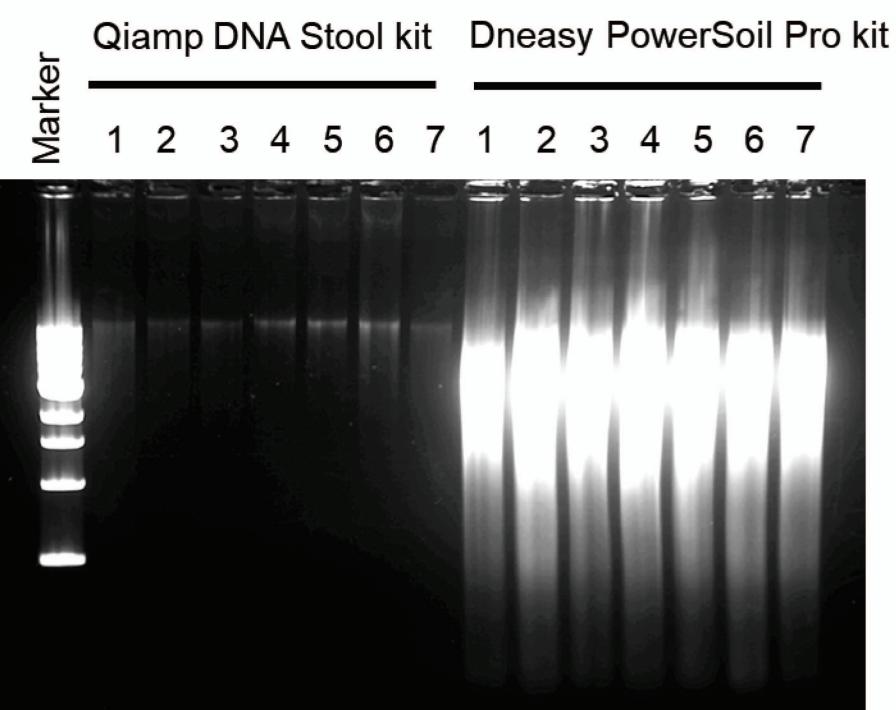


Fig7

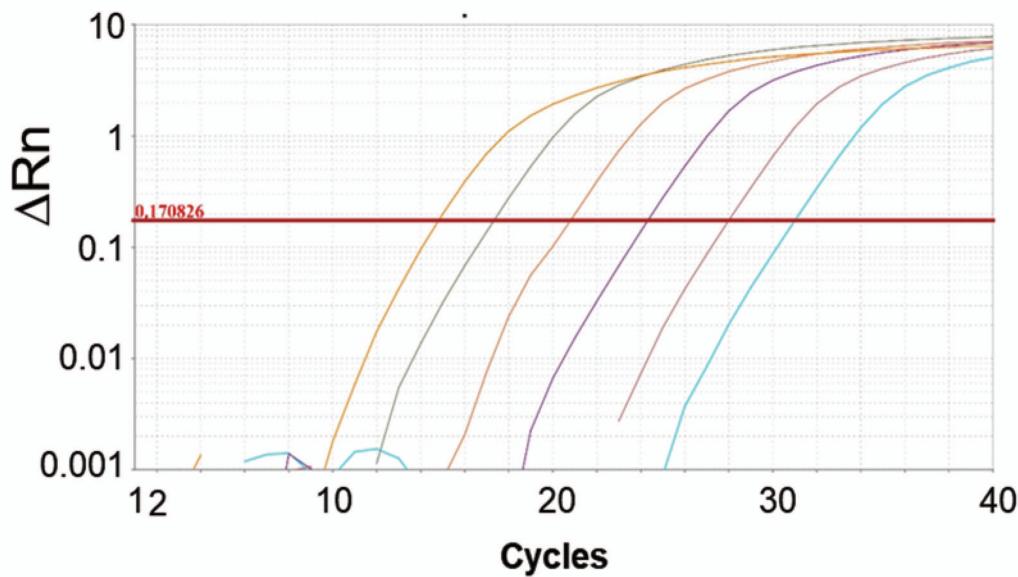


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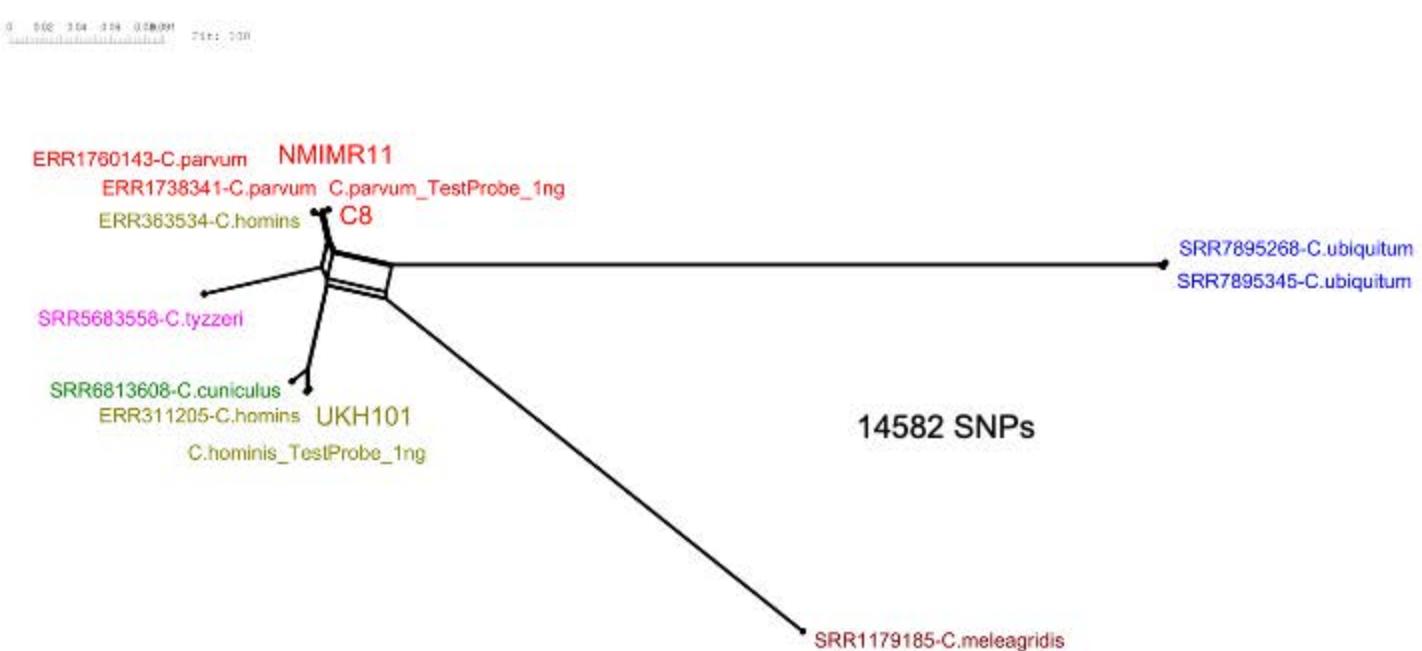
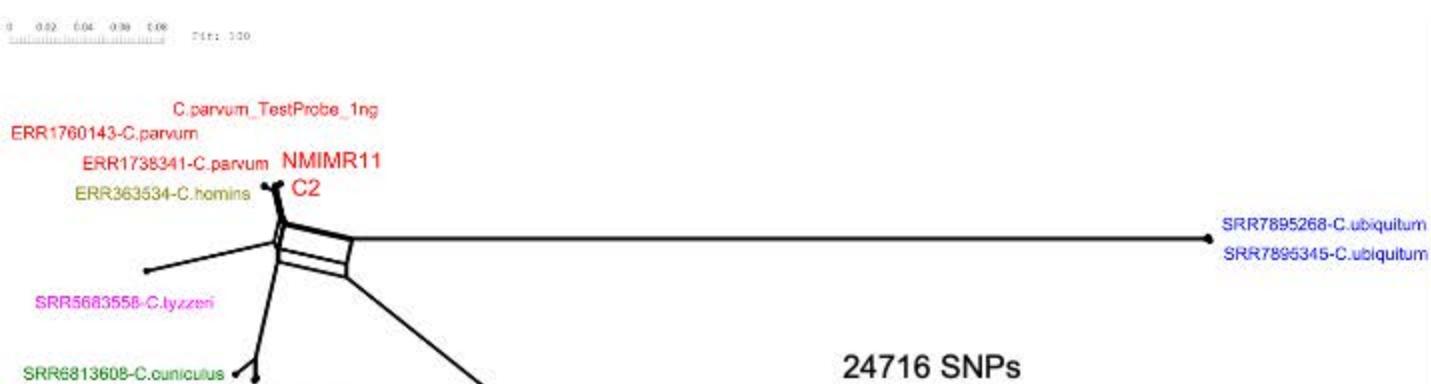
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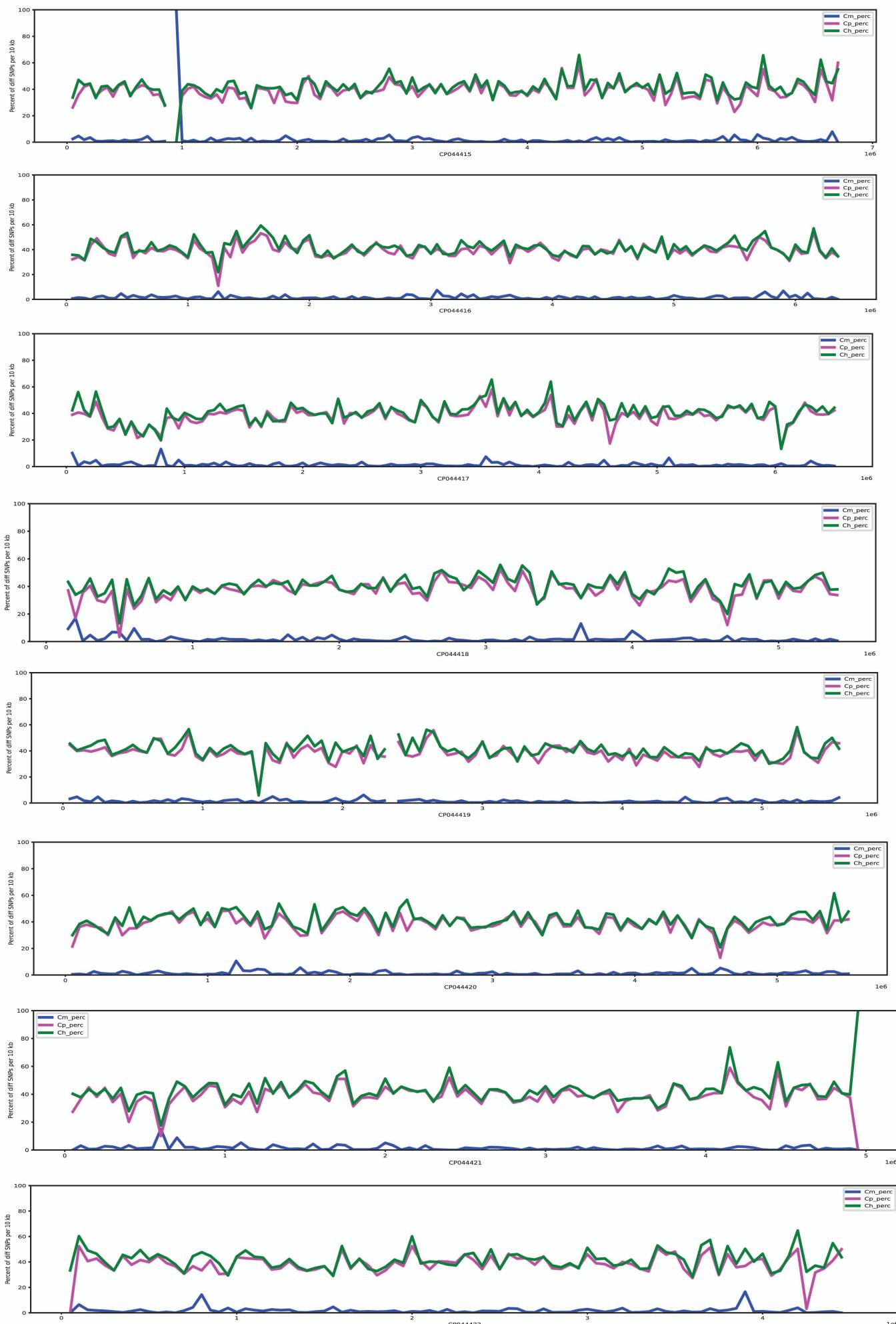


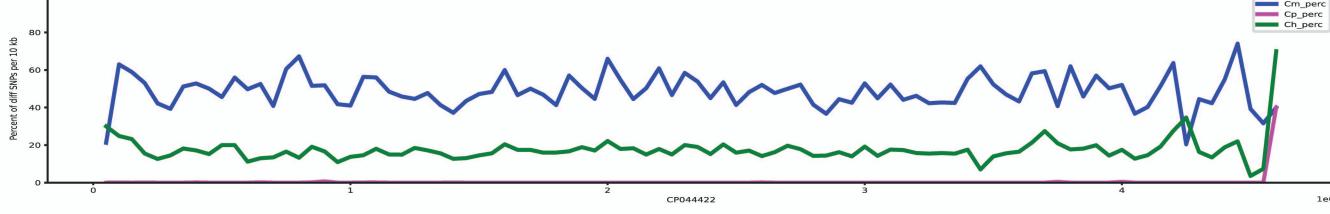
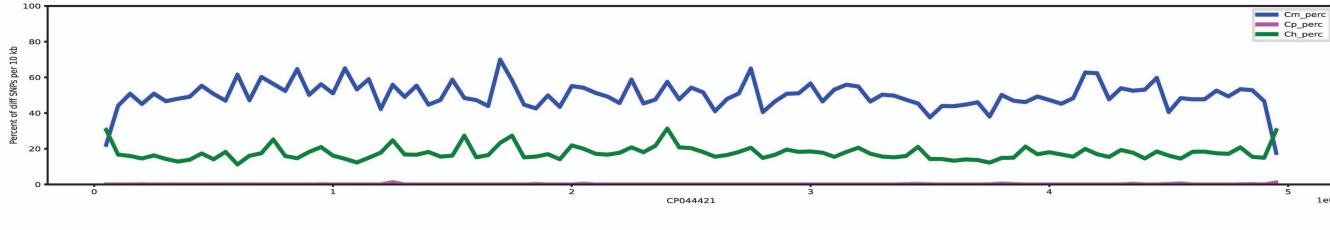
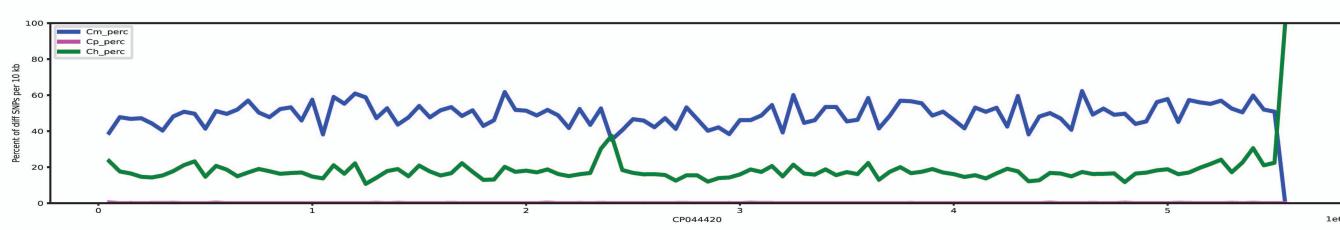
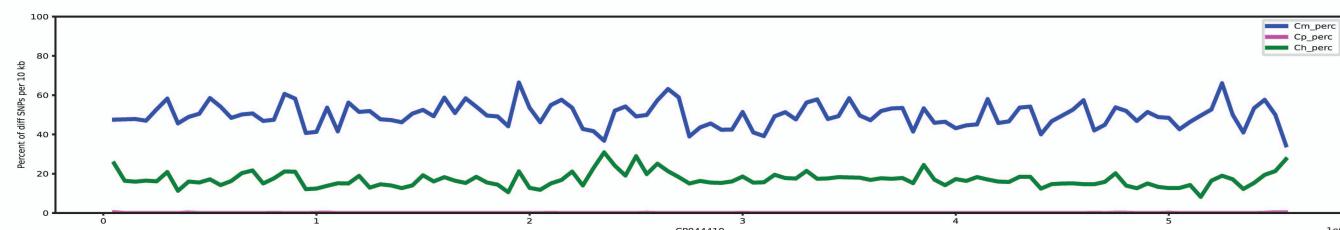
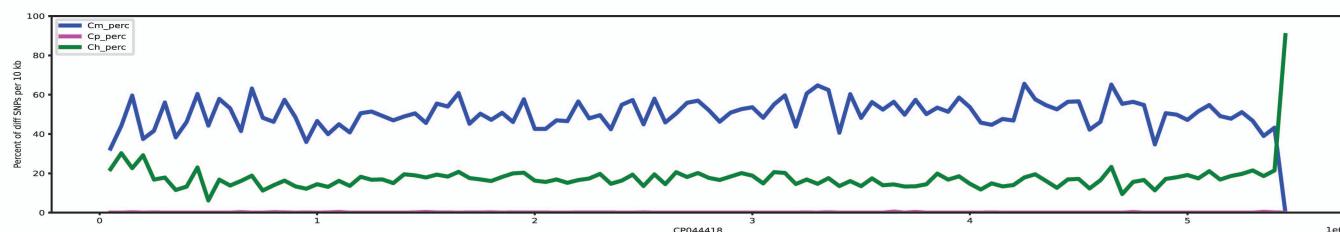
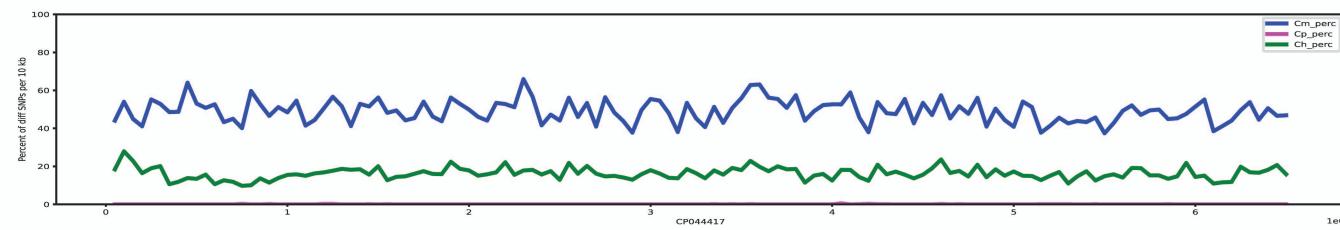
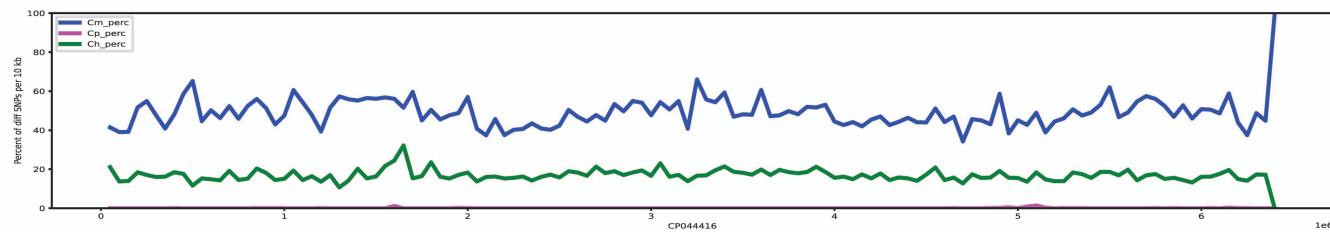
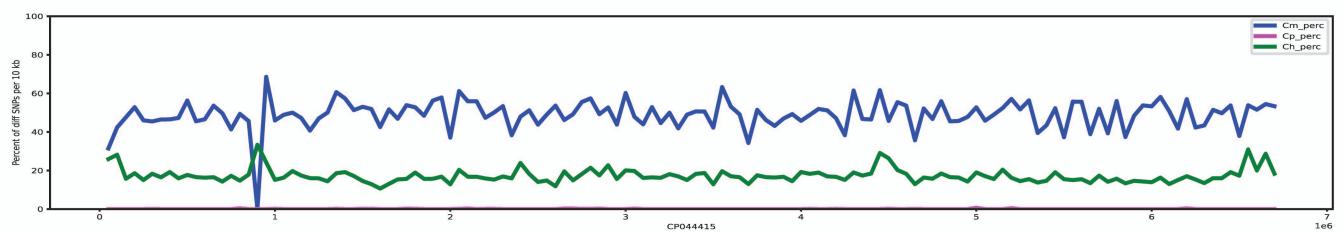
C

DNA dilution
10 ng/ul
1 ng/ul
0.1 ng/ul
0.01 ng/ul
0.001 ng/ul
0.0001 ng/ul
positive control
negative control

CT values
14.788
17.352
20.798
24.360
27.989
30.994
15.704
Undetermined







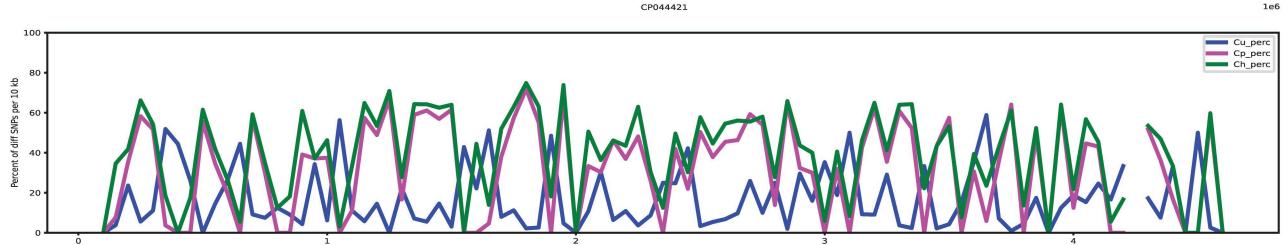
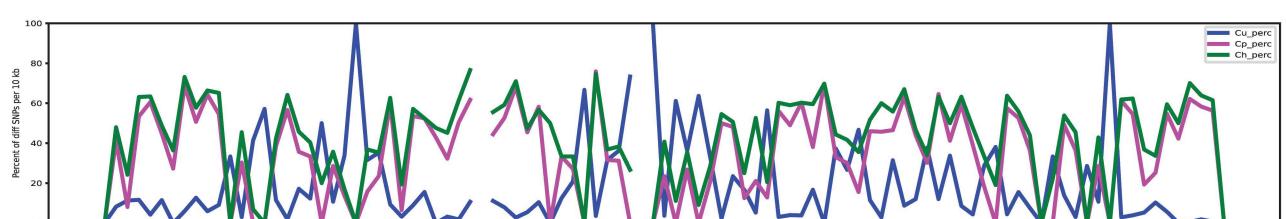
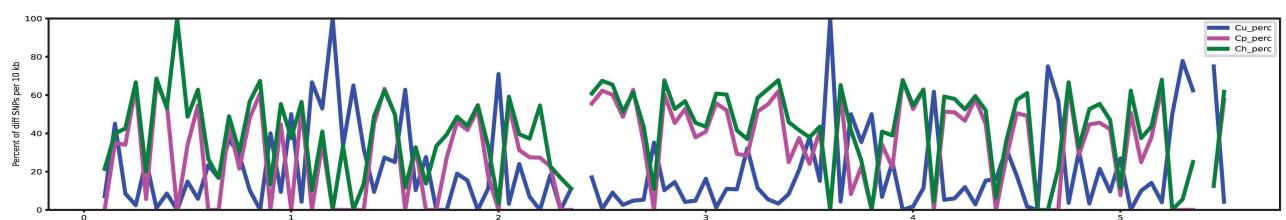
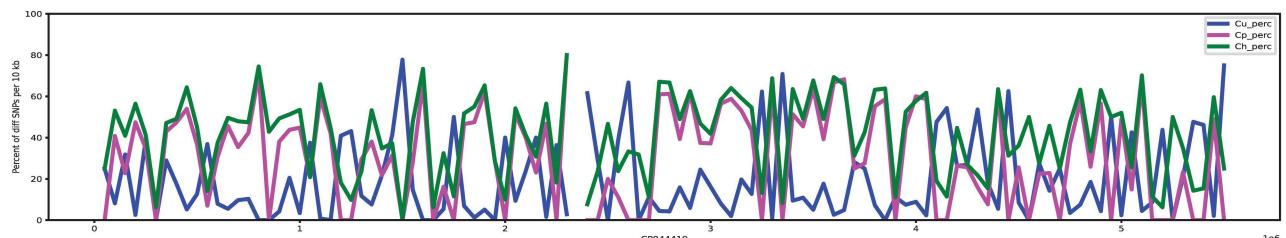
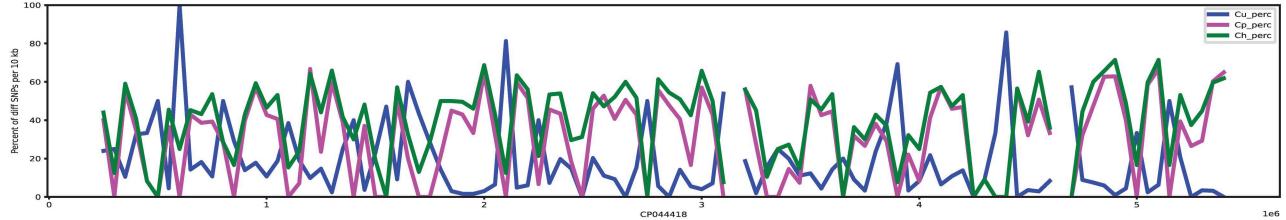
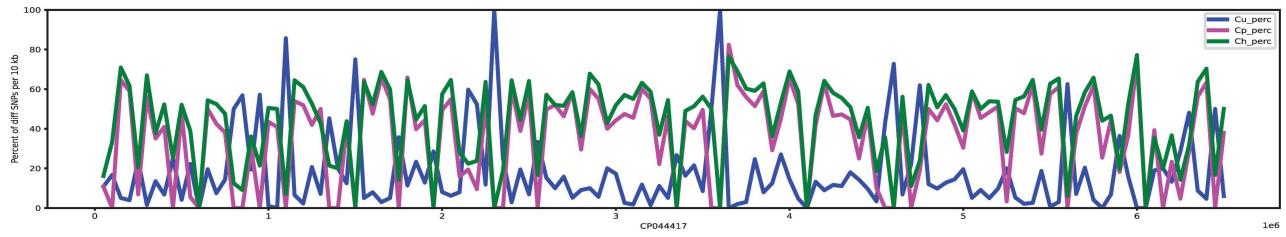
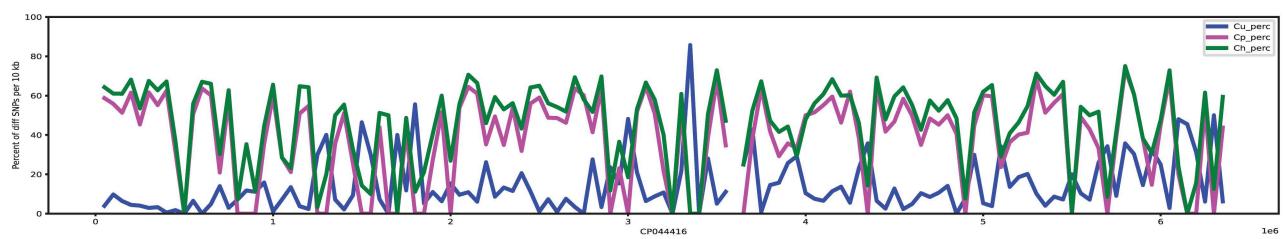
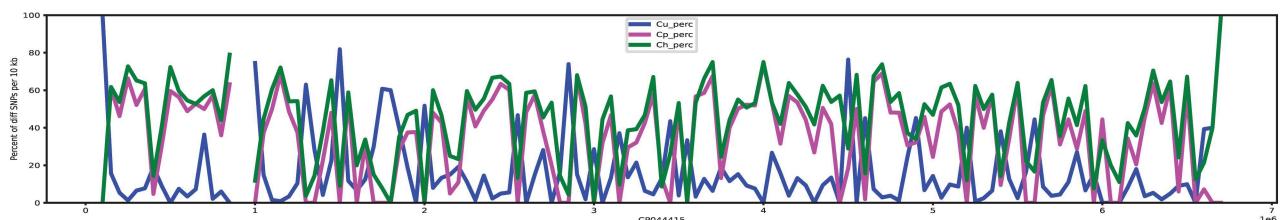


Table 1. Design of Cryptosporidium specific biotinylated baits for SureSelect target enrichment sequencing.

Chromosome ^a	Target Size ^b	Covered	Fraction Covered
CM000429	875659	808459	0.9232578
CM000430	985969	902040	0.91487663
CM000431	1099352	1019400	0.92727352
CM000432	1104417	977040	0.88466585
CM000433	1080900	945720	0.87493755
CM000434	1332857	1204320	0.9035628
CM000435	1278458	1210560	0.94689071
CM000436	1344712	1206000	0.89684631

^a Genome sequence of *C. parvum* was utilized to develop biotinylated baits throughout the genome (https://cryptodb.org/common/downloads/release-34/Cparvumlowall/fasta/data/CryptoDB-34_Cparvumlowall_Genome.fasta)

^b Total base pair (bp) size of each chromosome of *Cryptosporidium parvum*.

^c Total base pair (bp) size of the covered genome using biotinylated baits for SureSelect TES.

Table 2. Whole genome sequencing of libraries after amplification using SureSelect TES.

Amount of <i>C. parvum</i> gDNA ^a	Total number of sequenced reads ^b	Total number of mapped reads ^c	% of aligned read	Coverage (X)	Insert size	qPCR Ct
5.0ng	9224038	9018315	97.8	213	167	n.d.
1.0ng	8774746	8097021	92.3	172	166	17.35
0.1ng	6308564	4794797	76	108	156	20.80
0.01ng	6836298	2100385	30.7	38	153	24.36
0.001ng	5564782	201378	3.6	2	138	27.99
0.0001ng	2256618	27146	1.2	~0.0	20	30.99

^a Amount of *C. parvum* gDNA spiked into 200ng of total gDNA from host stool samples

^b total number of sequenced reads after Illumina sequencing using NextSeq500/550 Mid Output Kit v2.5 (300 cycles)

^c Total number of mapped reads after ref-mapping Illumina short read sequences against reference *C. parvum* reference sequence https://cryptodb.org/common/downloads/release-34/Cparvumlowall/fasta/data/CryptoDB-34_Cparvumlowall_Genome.fasta

n.d = not done

Table 3. Sample information.

ID	Sample names	Country	Origin	Material	Number of reads	Mapped reads	Unmapped reads	Coverage	Ct for <i>Cryptosporidium</i>	18S Type	Sequencer	Popsicle Type	Ct for <i>Giardia</i>	FastQ Screen Mapped to <i>G. lamblia</i> WB
Dg045	C2	Ghana	Human	Stool	88,344,601	622,429 / 0.7%	87,722,172 / 99.3%	2,1251	26.46	<i>C. hominis</i>	MiSeq+NextSeq	<i>C. parvum</i>		
Dg083	C8	Ghana	Human	Stool	88,437,644	539,311 / 0.61%	87,898,333 / 99.39%	2,1369	28.87	<i>C. hominis</i>	MiSeq+NextSeq	<i>C. parvum</i>		
ENRH_0052	NMIMR11	Ghana	Human	Stool	51,262,478	50,546,533 / 98.6%	715,945 / 1.4%	531,4183	17.19	<i>C. parvum</i>	MiSeq	<i>C. parvum</i>		
UKUB17	UKUB17	UK	Human	Stool	27,999,948	668,726 / 2.39%	27,331,222 / 97.61%	4,7236	29.37	<i>C. ubiquitum</i>	HiSeq	Different ancestral		
UKH101	UKH101	UK	Human	Stool	187,685,611	12,512,057 / 6.67%	175,153,554 / 93.33%	134,382	22.6	<i>C. hominis</i>	MiSeq+NextSeq	<i>C. hominis</i>		
UKP196	UKP196	UK	Human	Stool	45,329,037	12,234,615 / 26.99%	33,094,422 / 73.01%	181,1639	25.85	<i>C. parvum</i>	HiSeq	<i>C. parvum</i>		
EC1	EC1	Ecuador	Human	Stool	20,872,547	8,339,764 / 39.96%	12,532,783 / 60.04%	85,5575	27.15	<i>C. meleagridis</i>	MiSeq	Different ancestral	28.3	0
EC4	EC4	Ecuador	Human	Stool	136,152	2,141 / 1.57%	134,011 / 98.43%	0.0139	29.47	<i>C. canis</i>	MiSeq	Not used	28.6	6
51 Egypt OC	FEgypt	Egypt	Human	Stool	3,196,662	103,590 / 3.24%	3,093,072 / 96.76%	0.9797	35.46	none	MiSeq	Not used	27.71	1

Table S1. 18S Ribosomal RNA primers used in this study.

Organism	Target	Sequences	Reference
Cryptosporidium	18S	F, GGGTTGTATTTATTAGATAAAGAACCA	1(modified)
		R, AGGCCAATACCTACCGTCT	
Giardia	18S	F, GACGGCTCAGGACAACGGTT	2
		R, TTGCCAGCGGTGTCCG	
E. histolytica	18S	F, ATTGTCGTGGCATCTAACTCA	2
		R, GCGGACGGCTCATTATAACA	

F, forward; R, reverse

1. Stroup SE, Roy S, McHele J, Maro V, Ntabaguzi S, Siddique A, Kang G, Guerrant RL, Kirkpatrick BD, Fayer R, Herbein J, Ward H, Haque R, Houpt ER. 2006. Real-time PCR detection and speciation of Cryptosporidium infection using Scorpion probes. *J. Med. Microbiol.* 55:1217–1222.
2. Verweij JJ, Blange RA, Templeton K, Schinkel J, Brien EA, van Rooyen MA, van Lieshout L, Polderman AM. 2004. Simultaneous detection of *Entamoeba histolytica*, *Giardia lamblia*, and *Cryptosporidium parvum* in fecal samples by using multiplex real-time PCR. *J. Clin. Microbiol.* 42:1220–1223.

Table S2. CT values of 18S Ribosomal RNA qPCR assay using species specific primers for *Cryptosporidium*, *Giardia*, and *E. histolytica*.

Country	Samples Name	<i>Cryptosporidium</i> (C1 s40)	<i>Giardia</i> (C1 s40)	<i>E. histolytica</i> (C1 s40)	<i>Cryptosporidium</i> primers	<i>Cryptosporidium</i> species
	COL40	-	37.516	-	185	
	COL39	-	38.546	-	185	
	COL38	-	34.554	-	185	
	COL37	-	38.420	-	185	
	COL36	-	38.891	-	185	
	COL33	-	36.265	-	185	
	COL34	-	34.521	-	185	
	COL32	-	34.625	34.174	185	
	COL29	-	34.301	-	185	
	COL28	-	34.342	-	185	
	COL27	-	35.187	-	185	
	COL26	-	35.357	-	185	
	COL25	-	33.678	-	185	
	COL24	-	34.376	-	185	
	COL23	-	36.891	-	185	
	COL22	-	35.144	-	185	
	COL20	-	33.922	34.426	185	
	COL19	-	21.551	-	185	
	COL5	-	33.433	-	185	
	COL4	-	35.870	-	185	
	COL3	39.351	36.151	18.187	185	
	COL2	-	34.338	-	185	
	COL1	-	35.702	-	185	
	ID-14	-	35.007	-	185	
	ID-13	-	37.194	-	185	
	ID-12	-	36.793	-	185	
	ID-11	-	34.176	-	185	
	ID-10	-	36.155	-	185	
	ID-9	-	38.024	-	185	
	ID-8	-	36.437	-	185	
	ID-7	-	35.829	-	185	
	ID-6	-	36.418	-	185	
	ID-5	-	38.888	-	185	
	ID-4	-	34.310	-	185	
	ID-3	-	35.725	-	185	
	ID-2	-	-	-	185	
	ID-1	-	33.459	-	185	
	Ana-14	-	33.107	-	185	
	Ana-13	-	36.859	-	185	
	Ana-12	-	34.181	-	185	
	Ana-11	-	31.982	-	185	
	Ana-10	-	29.781	-	185	
	Ana-9	-	34.645	-	185	
	Ana-8	-	33.807	-	185	
	Ana-7	-	30.946	-	185	
Colombia	Ane-6	-	33.964	-	185	
	Ane-5	-	34.406	-	185	
	Ane-4	-	34.979	-	185	
	Ane-3	-	35.010	-	185	
	Ane-2	-	35.420	-	185	
	Ane-1	-	35.687	-	185	
	Sha-14	-	33.377	-	185	
	Sha-13	-	32.283	-	185	
	Sha-12	-	35.634	-	185	
	Sha-11	-	37.209	-	185	
	Sha-10	-	34.775	-	185	
	Sha-9	-	33.690	-	185	
	Sha-8	-	33.753	-	185	
	Sha-7	-	36.113	-	185	
	Sha-6	-	35.433	-	185	
	Sha-5	-	35.833	-	185	
	Sha-4	-	35.718	-	185	
	Sha-3	-	33.644	-	185	
	Sha-2	-	34.626	-	185	
	Sha-1	-	33.150	-	185	
	Fge-14	-	34.460	-	185	
	Fge-13	-	35.350	-	185	
	Fge-12	-	34.300	-	185	
	Fge-11	-	29.415	-	185	
	Fge-10	-	34.196	-	185	
	Fge-9	-	36.501	-	185	
	Fge-8	-	35.612	-	185	
	Fge-7	-	36.464	-	185	
	Fge-6	-	33.464	-	185	
	Fge-5	-	38.932	-	185	
	Fge-4	-	38.594	-	185	
	Fge-3	-	34.169	-	185	
	Fge-2	-	32.174	-	185	
	Fge-1	-	32.304	-	185	
	Ez0	-	31.187	-	185	
	Ez0	-	27.446	-	185	
	Ez17	-	25.795	-	185	
	Ez16	-	35.536	-	185	
	Ez15	-	26.812	-	185	
	Ez14	-	29.318	-	185	
	Ez13	-	28.986	-	185	
	Ez12	-	30.172	-	185	
	Ez9	-	30.448	-	185	
	Ez8	-	24.247	-	185	
	Ez7	-	26.005	-	185	
	Ez30	-	31.187	-	185	
	Ez20	-	27.446	-	185	
	Ez17	-	25.795	-	185	
	Ez16	-	35.536	-	185	
	Ez15	-	26.812	-	185	
	Ez14	-	29.318	-	185	
	Ez13	-	28.986	-	185	
	Ez12	-	30.172	-	185	
	Ez9	-	30.448	-	185	
	Ez8	-	24.247	-	185	
	Ez7	-	26.005	-	185	
	Ez1	29.468	28.397	-	185	<i>C. cuniculi</i>
	Ez1	27.145	28.336	-	185	<i>C. meleagridis</i>
Egypt	35.461	27.732	-	185	not resolved	
	50 Egypt OC	-	26.256	-	185	
	49 Egypt OC	-	28.884	-	185	
	49 Egypt OC	-	26.478	27.455	185	
	47 Egypt OC	-	28.150	-	185	
	46 Egypt OC	-	25.785	-	185	
	45 Egypt OC	-	29.302	-	185	
	44 Egypt OC	-	27.220	-	185	
	43 Egypt OC	-	27.531	-	185	
	39 Egypt OC	-	23.661	-	185	
	34 Egypt OC	-	18.606	-	185	
	34 Egypt OC	-	21.277	-	185	
	30 Egypt OC	-	23.293	-	185	
	31 Egypt OC	-	22.828	-	185	
	28 Egypt OC	-	21.244	-	185	
	29 Egypt OC	-	20.902	-	185	
	24 Egypt OC	-	25.759	-	185	
	23 Egypt OC	-	25.694	-	185	
	21 Egypt OC	-	21.032	-	185	
	21 Egypt OC	-	26.494	-	185	
	19 Egypt OC	-	26.862	-	185	
	18 Egypt OC	-	17.005	-	185	
	8 Egypt OC	-	26.653	-	185	
	3 Egypt OC	-	23.227	-	185	
Ghana	Dg045	28.48			185	<i>C. hominis</i>
	Dg046	26.45			185	<i>C. hominis</i>
	Dg048	30			185	<i>C. hominis</i>
	Dg052	23.86			185	<i>C. hominis</i>
	Dg053	23.23			185	<i>C. hominis</i>
	Dg076	27.27			185	<i>C. hominis</i>
	Dg081	26.8			185	<i>C. hominis</i>
	Dg082	28.87			185	<i>C. hominis</i>
	NMMR17	n.d.			185	<i>C. hominis</i>
	17.19				185	<i>C. parvum</i>
UK	UKP193	27.42			185	<i>C. hominis</i>
	UKP195	23.88			185	<i>C. parvum</i>
	UKP196	25.85			185	<i>C. parvum</i>
	UKP197	26.43			185	<i>C. parvum</i>
	UKC119	33.42			185	<i>C. cuniculi</i>
	UKH101	22.6			185	<i>C. hominis</i>
	UKH102	26.89			185	<i>C. hominis</i>
	UKH104	23.21			185	<i>C. hominis</i>
	UKH105	27.97			185	<i>C. hominis</i>
	UKU17	29.37			185	<i>C. ubiquitous</i>

"-" negative by qPCR
 "n.d." not done

Table S3. Chromosome ID from both genome references used.

Reference used	Chromossomes ID	Chromosome Number	Chromossomes ID	Reference used
<i>C. parvum</i> Iowall (CryptoDB Version 34)	CM000429	1	CP044422.1	<i>C. parvum</i> Iowa-ATCC (CryptoDB Version 54)
	CM000430	2	CP044421.1	
	CM000431	3	CP044420.1	
	CM000432	4	CP044419.1	
	CM000433	5	CP044418.1	
	CM000434	6	CP044417.1	
	CM000435	7	CP044416.1	
	CM000436	8	CP044415.1	

Table S4. PANEL of 25 different reference *Cryptosporidium* species for DEploid analysis.

ERR1035619	<i>Cryptosporidium parvum</i>
ERR1035621	<i>Cryptosporidium parvum</i>
ERR1738340	<i>Cryptosporidium parvum</i>
ERR1738345	<i>Cryptosporidium parvum</i>
ERR1738350	<i>Cryptosporidium parvum</i>
ERR1738337	<i>Cryptosporidium parvum</i>
ERR2366918	<i>Cryptosporidium parvum</i>
ERR961651	<i>Cryptosporidium parvum</i>
SRR6117460	<i>Cryptosporidium parvum</i>
SRR6147472	<i>Cryptosporidium parvum</i>
SRR6147945	<i>Cryptosporidium parvum</i>
SRR6147964	<i>Cryptosporidium parvum</i>
ERR1305020	<i>Cryptosporidium hominis</i>
ERR1305025	<i>Cryptosporidium hominis</i>
ERR2240056	<i>Cryptosporidium hominis</i>
ERR2240065	<i>Cryptosporidium hominis</i>
ERR2240074	<i>Cryptosporidium hominis</i>
ERR2366927	<i>Cryptosporidium hominis</i>
ERR2366935	<i>Cryptosporidium hominis</i>
SRR1015721	<i>Cryptosporidium hominis</i>
SRR1179185	<i>Cryptosporidium meleagridis</i>
SRR793561	<i>Cryptosporidium meleagridis</i>
SRR7895268	<i>Cryptosporidium ubiquitum</i>
SRR7895345	<i>Cryptosporidium ubiquitum</i>
SRR5683558	<i>Cryptosporidium tyzzeri</i>