1 CaptureCompendium: a comprehensive toolkit for 3C analysis.

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- 6 ⁺ Corresponding author: jim.hughes@imm.ox.ac.uk 7 8 1. MRC WIMM Centre for Computational Biology, MRC Weatherall Institute of Molecular 9 Medicine, University of Oxford, Oxford, UK 10 2. MRC Molecular Haematology Unit, MRC Weatherall Institute of Molecular Medicine, 11 University of Oxford, Oxford, UK 12 3. Big Data Institute, Li Ka Shing Centre for Health Information and Discovery, University 13 of Oxford, Oxford, UK 14 4. Wellcome Centre for Human Genetics, Nuffield Department of Medicine, University of 15 Oxford, Oxford, UK 16 17 18 Keywords: 3C, chromosome conformation capture, Capture-C, software.

19 ABSTRACT

20 DNA folding within nuclei is a highly ordered process, with implications for gene 21 regulation and development. An array of chromosome conformation capture (3C) 22 methods have been developed to investigate how DNA is packaged within nuclei and to 23 interrogate specific interactions. While these methods use different approaches to 24 examine target loci (many-versus-all) or the entire genome (all-versus-all), they all rely on the core principle of endonuclease digestion and proximity-based ligation to re-arrange 25 26 genomic order to reflect the three-dimensional nuclear conformation. This sequence 27 reorganization creates novel chimeric DNA fragments which require specialist 28 bioinformatic tools to analyze and visualize. Despite this need for specialist bioinformatic 29 skills, the core biological importance of genome folding has seen widespread 30 methodological uptake. To service the needs of experimentalists using the many-versus-31 all Capture-C family of methods we have developed CaptureCompendium; a toolkit of 32 software to simplify the design, analysis and presentation of 3C experiments.

33 Introduction

34 The arrangement of the genome into the nucleus is highly regulated to facilitate correct 35 interactions between promoters and their requisite enhancers. Chromosome conformation 36 capture (3C) has emerged as a highly malleable method for studying both local and global trends 37 in genome folding¹. All 3C methods rely on the underlying principle that following DNA digestion, 38 generally with a restriction endonuclease (RE), fragment ends in proximity will ligate together at 39 high frequency. This digestion and proximity driven ligation functions to re-organize the linear 40 genome sequence to reflect its 3D structure². This re-organized sequence is then examined 41 primarily by targeted high-resolution many-versus-all approaches, e.g. NG Capture-C³, 4C-seq⁴, 42 UMI-4C⁵, i4C-seq⁶, and Tri-C⁷; or lower resolution approaches across the genome, e.g. HiC⁸ 43 and Capture-HiC⁹.

44

45 The importance of genome folding to transcriptional regulation has brought about the application 46 of 3C methods to investigate a range of questions, including basic biological problems and the 47 identification of disease relevant gene-to-polymorphism interactions^{10,11}. Owing to the enhanced 48 signal provided by targeted sequencing, oligonucleotide capture methods, based on Capture-49 C^{3,7,9,12-14}, are emerging as a preferred method to look at loci of interest. A standard Capture-C 50 experiment consists of three main bioinformatic components: target selection and 51 oligonucleotide probe design; sequence mapping with read classification; and statistical 52 comparison with data presentation. For each of these processes an array of software tools is 53 available, including HiCapTools¹⁵, GOPHER¹⁶, Capsequm¹², capC-MAP¹⁷, FourCSeq¹⁸, 54 peakC¹⁹, CHiCAGO²⁰, peaky²¹, ChiCMaxima²², HiC-Pro²³, and 4See²⁴; yet, to date no 55 comprehensive pipeline for the complete design, mapping and interpretation of 3C experiments has been made available. This lack of an easily implemented, automated workflow for Capture-56 57 C based experiments forms a barrier to the general use of such methods, particularly in non-58 specialist groups. To address this need we have developed the CaptureCompendium toolkit 59 (Fig. 1) which combines oligonucleotide design (Capsegum2), sequence mapping and data 60 extraction (CCseqBasic), and statistical analyses with data presentation and distribution 61 (CaptureCompare, Tri-C, CaptureSee). This workflow streamlines the bioinformatic process 62 required to generate high quality 3C data using Capture-C experiments.

63 **RESULTS**

64 *Capsequm2: A user-friendly interface for probe design.*

65 Targeted enrichment with biotinylated DNA oligonucleotides forms the basic tenet of Capture-C 66 methods³. Probe design is essential to a successful experiment; probes must be situated near 67 to a RE cut site, be proximal to the object of interest, and not be located on a highly repetitive 68 element. These parameters allow for enrichment of informative on-target chimeric sequencing 69 fragments, while simultaneously avoiding off-target sequencing. Currently, we know of three 70 tools are available for probe design: HiCapTools - a command line-based method¹⁵, GOPHER 71 - an installable JAVA package¹⁶, and Capsequm - an online interface¹². As a webtool, 72 Capsegum provides the most user-friendly option for initial probe design but previously required 73 integration with MIG²⁵, a registration-based service, for probe filtering.

74

75 We redeveloped Capsegum (http://apps.molbiol.ox.ac.uk/CaptureC/cgi-bin/CapSegum.cgi) with 76 a web-based filtering portal (AltSort) to generate Capsequm2 (Fig. 2). Capsequm2 allows 77 researchers to select from a range of 11 genomes covering seven species and the five 78 commonly used RE, as well as allowing a choice in probe length for design to targets of interest. 79 Following fragment identification and BLAT processing, probes can be filtered using AltSort. In 80 AltSort probes filtering are automatically categorised as good or bad using recommended 81 settings for duplication number, BLAT density score (measuring likely off-target capture) (Kent 82 2002), G/C content, CpG content and, repeat content (Table 1). Probes can be visually 83 inspected with integrated UCSC browser tracks, and the default parameters adjusted for all or 84 specific regions of interest. Upon selection of desired probes, users can then download the 85 oligonucleotide sequences, ready for ordering. This download also provides correctly formatted 86 viewpoint input files required for subsequent analysis with the downstream tools CCseqBasic 87 and CaptureCompare.

88

89 Table 1. Default AltSort parameters for probe design.

Filter	"Good"	"Bad"
Duplicates	≤2*	≥3
Density	≤40	>40
% G/C content	≤60	≥60
Repeat	False	True

90 *Often interactions at duplicated genes, e.g. *Hba-1*, *Hba-2*, can still be understood.

91

92 CCseqBasic: Data extraction and quality control.

93 Analysis of sequencing data with the CCseqBasic command line tool (Fig. 3a) is launched using

94 the probe design parameter file provided by AltSort. Enriched chimeric fragments from the

95 Capture-C family of methods, NG Capture-C, LI-Capture-C, Tag Capture-C, Tri-C, are generally 96 sequenced using 150-bp paired-end reads^{3,7,13,26}; this gives 300-bp of total sequence, therefore 97 the majority of read-pairs overlap and the chimeric fragment can be reconstructed through 98 flashing²⁷. RE sites in chimeric fragments generated from 3C digestion and ligation are then 99 identified in both flashed and non-flashed reads and digested-in silico. These digested reads 100 are then mapped and categorised as capture, occurring within a targeted RE fragment, or 101 reporter, occurring in any other fragment. Because Capture-C methods use sonication or 102 tagmentation to generate random ends for reads, they can then be PCR duplicate filtered. All of 103 these steps are carried out automatically by CCseqBasic with minimal user input. The main 104 output of CCseqBasic is a convenient UCSC-loadable data hub which contains reporter tracks 105 for each viewpoint (Fig. 4), complete with sample quality control metrics for 3C library quality 106 and enrichment efficiency (Fig. 3b), as well as sequence quality control in the form of fastQC 107 reports. Additionally, all generated data is stored in an easy-to-follow hierarchical folder structure 108 allowing re-analysis, or immediate further processing with CaptureCompare or Tri-C.

109

110 Tri-C: Investigation of 3-way interaction hubs.

111 Most 3C methods interrogate single, pairwise interactions to generate a population view of 112 interaction frequencies. Recently, interest has been given to the formation of multi-way 113 interaction hubs^{28–30}. By enriching for chimeric fragments containing multiple ligation junctions, 114 Tri-C allows investigation of these multi-way interactions from single-alleles⁷. Tri-C interaction 115 matrices are generated using an optional flag, which can be implemented during either the initial 116 run or re-analysis with CCseqBasic. When using Tri-C analysis, users can specify the genomic 117 regions to analyse, sonication fragment size, and bin size to match their specific experiment. 118 The output of this analysis is a pdf of an interaction matrix for the region surrounding the 119 viewpoint of interest (Fig 5.).

120

121 CaptureCompare: Statistical analysis and sample comparison.

122 A major strength of Capture-C methods is their ability to confidently distinguish PCR duplicates. 123 while simultaneously generating data from multiple replicates and cell types by multiplexing 124 samples³. This allows experiments to be designed with advanced statistical analysis in mind. 125 such as the comparison between cell types, differentiation states, genetic models, drug 126 treatment or any other biological question of interest. As with analysis by CCseqBasic, 127 CaptureCompare is implemented using a command line interface and AltSort-generated 128 parameter file. CaptureCompare extracts the raw interaction count data for each viewpoint from 129 replicates of multiple conditions and generates a UCSC loadable hub containing normalised 130 mean interactions, standard deviation, and difference in interaction level for different conditions 131 on a per RE fragment basis (Fig. 4). Additionally, CaptureCompare generates and implements

a modifiable R script for DESeq2 analysis³¹, which compares RE fragments to determine if
observed differences between conditions reach significance. CaptureCompare also automates
the preparation of the necessary files for more complex interaction frequency analysis via the
Bayesian modelling tool peaky²¹. Finally, to facilitate presentation, a pdf is generated for each
viewpoint with publication-style windowed interaction profiles (Fig. 4).

137

138 CaptureSee: A simple interface for navigating and sharing experiments.

139 On-going methodological improvements have increased the ability of Capture-C methods to 140 target thousands rather than tens to hundreds of loci. Exploration of these large datasets is 141 difficult when using the UCSC browser. To facilitate easy interaction with 3C data, and provide 142 a platform for sharing results, we developed the web-based platform CaptureSee. CaptureSee 143 takes the output hub from CaptureCompare to generate a browsable repository of viewpoints in 144 both per RE fragment and windowed formats (Fig 6.). Within CaptureSee users can guickly jump 145 between capture viewpoints in the genome, to see both the total number and percent of cis 146 reporter counts for each viewpoint on a per sample basis, and filter DESeg2 results to find 147 significantly interacting RE fragments.

148

149 **DISCUSSION**

150 The importance of genome folding for gene regulation has come to the fore in recent years. An 151 extensive array of techniques in fixed and native cells has been developed to examine this 152 folding based on the premise of proximity ligation. Despite being based on a simple premise, 3C 153 sequencing generates a complex bioinformatic problem of mapping and interpreting chimeric DNA fragments. To meet this need, an equally wide range of tools have been developed to 154 155 either design experiments^{12,15,16}, map sequence reads^{3,17}, or interpret interaction profiles^{18–22}. 156 While some of these tools have been developed with a non-bioinformatically trained user in 157 mind, many are complex to implement. The absence of a coherent analysis workflow 158 necessitates the definition and testing of appropriate solutions and tools for each step of design, 159 analysis, interpretation and visualisation which must then be assembled into a functioning 160 bioinformatics pipeline, before such experiments can be productively undertaken. This has 161 slowed the general uptake into other fields, such as genetics, clinical studies and developmental 162 biology, though Capture-C methods have proven utility in these areas^{10,32–41}.

163

We have generated a systematic, semi-automated workflow for the design and analysis of Capture-C based 3C experiments. This pipeline is constructed with the end-user in mind, including easy to navigate web tools where computationally possible. For the command line tools the default settings will suffice for general use, and user input is only required at three stages. More advanced users will find their needs met, with options and flags available at all

- 169 stages to customise analysis such as the choice in mapping algorithm or using SNPs to
- 170 produce allele-specific profiles³. Furthermore, the availability of well-documented open source
- 171 software in commonly used programming languages (Unix, perl, R) means further modifications
- 172 can be easily incorporated. Therefore, CaptureCompendium represents a significant advance
- 173 for the implementation of capture-based 3C methodologies that will enable widespread use of
- 174 these powerful techniques to explore genomes structure.

175 **METHODS**

176 Capture-C data presented in this work are available on GEO (GSE129378) and can be 177 visualised at (http://capturesee.molbiol.ox.ac.uk/projects/capture compare/1387). The Tri-C 178 interaction profile was generated from published data⁷ (GSE107755). CaptureCompendium with 179 links to software, installation guide, usage manual and tutorial vignettes is available online 180 (http://userweb.molbiol.ox.ac.uk/public/telenius/CaptureCompendium/). Capsequm2, AltSort 181 and CaptureSee are available as web tools with full descriptions on the websites 182 (http://apps.molbiol.ox.ac.uk/CaptureC/cgi-bin/CapSegum.cgi and 183 https://capturesee.molbiol.ox.ac.uk/). To facilitate saving and sharing data, CaptureSee requires 184 users to register for a free account. CCseqBasic is modified from previous published versions³, 185 and now contains Tri-C. CCseqBasic is available on GitHub (https://github.com/Hughes-186 Genome-Group/CCseqBasicS) and implements a suite of tools to analyse data. Briefly adaptor 187 with TrimGalore sequences а trimmed 188 (https://www.bioinformatics.babraham.ac.uk/projects/trim_galore/), reads are flashed using 189 FLASH²⁷, then *in silico* digested, and mapped with either bowtie or bowtie^{242,43}. Read quality is 190 measured at multiple stages of processing FastQC using 191 (http://www.bioinformatics.babraham.ac.uk/projects/fastgc/). Reads are classified as capture or 192 reporter depending on mapping location and PCR duplicates are filtered based on duplication 193 of random end pairs generated by sonication. Skew arising from co-capture from 194 oligonucleotides at two independent viewpoints is excluded by masking all viewpoints included 195 within Capture-C experiment. CaptureCompare а is available on github 196 (https://github.com/Hughes-Genome-Group/CaptureCompare), and uses custom perl and R 197 scripts to analyse each viewpoint independently by normalising reads per 100,000 cis 198 interactions, combining replicates using BEDtools⁴⁴, generating mean and subtraction data and 199 inputting read counts into DESeq2³¹ treating each restriction fragment as an individual object. 200 Bedgraphs are converted to bigwigs with deepTools2⁴⁵. Windowed plots are generated with 201 adjustable bin and window sizes using custom R scripts.

202

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- 216

217 Competing Interests

218 J.R.H and J.O.J.D are founders and shareholders of Nucleome Therapeutics.

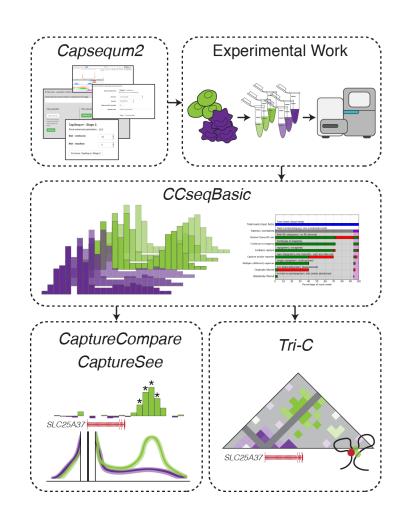
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322		

323 **FIGURES**

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325 326

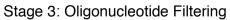
Figure 1 I Overview of CaptureCompendium. Biotinylated oligonucleotides for 3C experiments can be designed using Capsequm2. Following the experimental work, individual samples are analysed and QC is performed in CCseqBasic. Whole experiments, with replicates and comparators are analysed in CaptureCompare and visualized in CaptureSee. Multiinteraction hubs in Tri-C can also be detected using CCseqBasic.

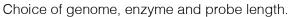
Stage 1: Co-ordinate input

Enter the Capture-C oligo design parameters below								
Upload positions file	Browse slc25a37.txt (BED format / max. 1000 positions):							
Genome:	Human (hg19)							
Enzyme:	Dpnll (GATC)							
Oligonucleotide size (bp):	70							
Dataset name:	Capture_Compendium							
	Stage 1 - Construct oligos							

Stage	2: I	BLAT	Calcul	lation
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CapSequm - Stage 2.							
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Blat - minScore:	10	·					
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Continue: CapSequm Stage 2							





Filter data - usi	ing MIG or /	AltSort:										
UCSC genome browser session url (optional at this stage):												
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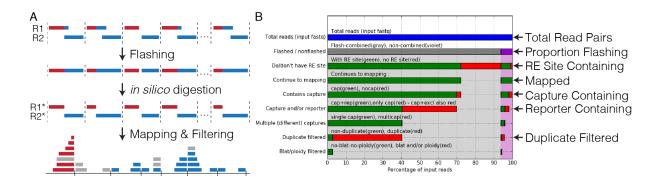
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- 333
- **Figure 2 | CapSequm2 provides a user-friendly web interface for oligonucleotide design.**

335 The online tool allows design to desired capture locations with selection of genome from Human

336 (hg18/19/38), Mouse (mm9, mm10), Chimpanzee (panTro3), Zebrafish (danRer7, danRer10),

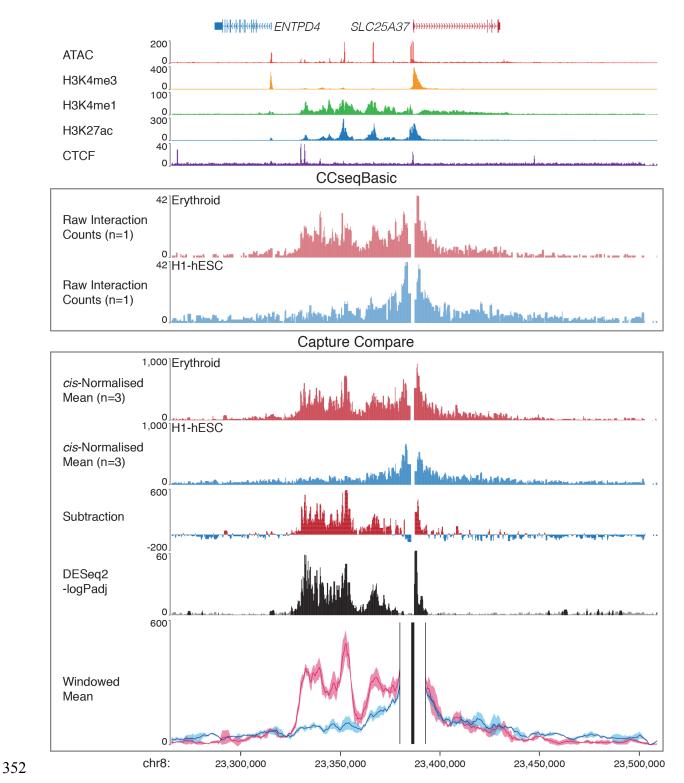
Drosophila melanogaster (dm3), Chicken (galGal4, galGal5), & Sickleback (BROADS1), of enzyme (*Dpn*II, *Nla*III, *Bam*HI, *Bgl*II, *Hind*III), and of oligonucleotide size. Designed oligonucleotides are then quality filtered using MIG or AltSort which allows integration with chromatin data. AltSort automatically selects oligonucleotides on recommended BLAT density,

341 duplication, and repeat values which can be manually adjusted.



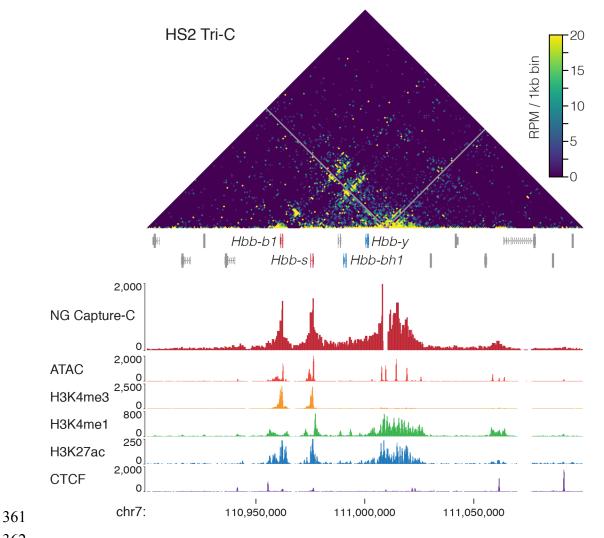
342 343

344 Figure 3 I CCseqBasic extracts informative data from 3C sequence reads. (a) 3C libraries 345 are analysed by reconstructing original fragments by flashing. Subsequent in silico digestion of 346 reads resolves chimeric fragments, and allows mapping with bowtie. Mapped reads are 347 identified as "Capture" (red): coming from a targeted viewpoint, or "Reporter" (blue): coming from an interacting fragment. Capture-reporter read pairs are then filtered for PCR duplicates 348 349 and off target capture sites (grey). (b) 3C library and capture quality metrics are reportable as 350 fractions of total reads to allow easy tracking of experimental success, or where necessary, 351 facilitate troubleshooting.



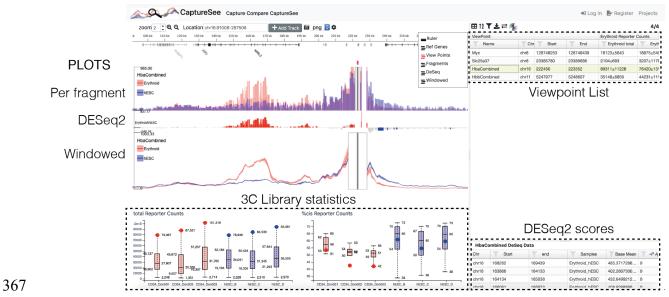
353

Figure 4 I CaptureCompare allows statistical analysis of experiments. The outputs of CCseqBasic and CaptureComparecan be directly loaded into UCSC for interpretation with chromatin epigenetic data. Example ouput is a Capture-C experiment from the viewpoint of the human mitoferrin gene *SLC25A37*. Output tracks include raw reporter counts, *cis*-normalised reporter counts, subtraction between comparison samples, and -logPadj values from DESeq2 on a per fragment basis. A pdf of windowed mean interactions with standard deviations is also provided. For DESeq2, significantly different fragments have been coloured black (FDR q<0.05).



362

Figure 5 I Detection of multi-way interaction hubs with Tri-C analysis. The output interaction map from Tri-C analysis of the viewpoint from the HS2 element of the mouse betaglobin Locus Control Region (LCR) shows concurrent interaction between HS2, the active betaglobin genes, and other elements within the LCR.



368

Figure 6 I Data exploration and sharing with CaptureSee. Data is loaded into CaptureSee using the CaptureCompare generated public hub. Within CaptureSee users can quickly move between viewpoints; with both per fragment and widowed views. In addition to 3C profiles, reporter counts and percent *cis* interactions are provided for each sample 3C library. DESeq2 logPadj scores are provided on a per fragment basis and can be filtered for level of significance.