Chromatin Compartments and Selection on X

How Edges of Active Chromatin Align with Selection Regions in Primates

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Abstract

This is a dummy abstract, dreamt up by chatGPT. This thesis investigates the 3D chromatin architecture of the X chromosome in baboons, macaques, and humans, focusing on chromatin compartments during spermatogenesis. Using publicly available Hi-C data, interaction maps were created to identify Principal Component 1 (PC1) compartments, revealing distinct compartmentalization patterns among species. The analysis included transition zones, where chromatin shifts between compartment types, and their correlation with positively selected regions. By comparing these zones with evolutionarily significant regions, the study explores how chromatin structure influences evolutionary pressures. Key findings include conserved chromatin features that may help retain non-advantageous alleles, suggesting a role for selfish genetic elements in genome evolution. This research offers new insights into the relationship between chromatin architecture and evolutionary dynamics across primate species.

## Introduction

### Sexual reproduction (spermatogenesis, meiosis)

The production of gametes in a sexually reproducing organism is a highly complex process that involves numeruous elements. Spermatogenesis, the process of forming male gametes, involves four stages of differentiation from a germ cell through *spermatogonia*, *pachytene spermatocyte*, and *round spermatids* to *spermatozoa*, or *sperm* (Wang et al. 2019), and it is the very basis of male reproduction. The specialized cell division of meiosis neatly handles the pairing, recombination, and segregation of homologous chromosomes, thereby ensuring proper genetic distribution. Deeply understanding the steps of molecular steps of reproduction and how our genetic material is inherited is essential in biology, bringing insight to areas such as speciation, population diversity, and (male) infertility.

### Selfish genes (and randomness)

The conventional story of meiosis in gametogenesis is one of random segregation of the sex chromosomes. They split into haploid gametes, where each chromosome has an equal chance of being passed on to a gamete. That seems like a fair game, but what if some genes are cheating the system by making others less viable. A meiotic driver is a selfish gene element that modulates meiosis and preferentially transmits its own allele through meiosis, regardless of the downstream fitness effects it may have (good or bad) on the organism it is part of. This phenomenon challenges the traditional understanding of selection, extending its scope beyond the fitness effects on an organism to include selective pressures at the molecular level. For example, if some genes on the X chromosome create a disadvantage for gametes that *do not* contain those genes, making sure the Y chromosome is not as viable as the X, resulting in a sex imbalance and possibly numerous other downstream effects. That is exactly what is coined *sex chromosome meiotic drive* (Jaenike 2001), a result of selfish genetic elements.

Motivated by previous results in the Munch Research group (Munch 2024) on hybrid incompatibility and extended common haplotypes (Skov et al. 2023; Sørensen et al. 2023) that could be explained by meiotic drive, we wanted to investigate how these patterns correlate with chromatin compartments.

### High-Throughput Chromosome Conformation Capture (Hi-C)

Our DNA can be divided into different orders of structure. *3C* focus on identifying the highest orders of organization inside the nucleus, that is, when the 30 nm thick coil of chromatin fibers folds into loops, Topologically Associating Domains (TADs), and chromatin compartments. Here, we narrow our focus on the largest of the structures, *compartments*, that is known to determine availability to transcription factors, thus making an *A* compartment *active*—and the *B* compartment *inactive*. The introduction of the Hi-C (high-throughput 3C) method (Lieberman-Aiden et al. 2009) opened new possibilities for exploring the three-dimensional organization of the genome.

## Methods

In this project, we formulate two objectives:

**A**: Reproduce the Hi-C interaction maps and eigendecomposition from (Wang et al. 2019), with some modifications. We briefly use *HiCExplorer*, but change the analyses to use the *Open2C Ecosystem* (Open Chromosome Collective 2024) which have a Pyton API as well as command-line functions, which can be paired very well with Jupyter Notebooks. The majority of the data analysis was run with a *gwf* workflow, and the commands that were visually inspected were run in Jupyter Notebooks.

**B** Compare with regions of selection that are found in *papio anubis*, and maybe in *human* too. Investigate the biological meaning of the results.

All computations were performed on GenomeDK (GDK) [ref], an HPC cluster located on Aarhus Uninversity, and most of the processing of the data was nested into a *gwf* workflow [ref], a workflow manager developed at GDK. I would like to thank GDK and Aarhus University for providing computational resources and support that contributed to these research results.

The whole of this project is carried out with reproducibility in mind, so an effort (and quite a significant amount of time) has been put into documenting code and organizing the project for readbility and transparency through a Quarto project [ref]. Therefore, all code, virtual environments and text is made available as a Quarto book, rendered directly from the GitHub repository with GitHub Pages []. To make this possible, the Quarto documentation has been extensively studied and discussed with *KMT* [ref, aknowledge].

### Downloading Data and Project Structure

To reproduce the results from (Wang et al. 2019), I chose to use their raw data directly from the SRA portal [ref]. I filtered the data to contain all their paired-end Hi-C reads, and included only macaque samples. The data set also contains RNAseq data, and the same tissues for both macaque and mouse. The meta data for the data set was extracted into a runtable SRA-runtable.tsv. To get an overview of the data accessions used in this analysis, we will first summarize the runtable that contains the accession numbers and some metadata for each sample ([Table 1](#tbl-runtable)). It adds up to ~1Tb of compressed fastq files, holding ~9.5 billion reads, roughly evenly spread on the 5 tissue types.

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| Table 1: The most relevant columns of the SRA-runtable.tsv file   |  | source\_name | BioSample | Run | GB | Bases | Reads | | --- | --- | --- | --- | --- | --- | --- | | 16 | fibroblast | SAMN08375237 | SRR6502335 | 29.771059 | 73,201,141,800 | 244,003,806 | | 17 | fibroblast | SAMN08375237 | SRR6502336 | 22.755361 | 65,119,970,100 | 217,066,567 | | 18 | fibroblast | SAMN08375236 | SRR6502337 | 21.434722 | 52,769,196,300 | 175,897,321 | | 19 | fibroblast | SAMN08375236 | SRR6502338 | 21.420030 | 52,378,949,100 | 174,596,497 | | 20 | fibroblast | SAMN08375236 | SRR6502339 | 10.207410 | 28,885,941,600 | 96,286,472 | | 9 | fibroblast | SAMN08375237 | SRR7349189 | 52.729173 | 139,604,854,200 | 465,349,514 | | 10 | fibroblast | SAMN08375236 | SRR7349190 | 53.085520 | 142,008,353,400 | 473,361,178 | | 21 | pachytene spermatocyte | SAMN08375234 | SRR6502342 | 60.258880 | 150,370,993,500 | 501,236,645 | | 22 | pachytene spermatocyte | SAMN08375234 | SRR6502344 | 27.146048 | 65,697,684,300 | 218,992,281 | | 23 | pachytene spermatocyte | SAMN08375234 | SRR6502345 | 26.202707 | 63,490,538,700 | 211,635,129 | | 0 | pachytene spermatocyte | SAMN09427370 | SRR7345458 | 55.970557 | 153,281,577,900 | 510,938,593 | | 1 | pachytene spermatocyte | SAMN09427370 | SRR7345459 | 53.982492 | 144,993,841,200 | 483,312,804 | | 11 | pachytene spermatocyte | SAMN08375235 | SRR7349191 | 51.274476 | 137,821,979,100 | 459,406,597 | | 24 | round spermatid | SAMN08375232 | SRR6502351 | 20.924497 | 55,095,075,300 | 183,650,251 | | 25 | round spermatid | SAMN08375232 | SRR6502352 | 41.133960 | 115,578,475,800 | 385,261,586 | | 26 | round spermatid | SAMN08375232 | SRR6502353 | 36.444117 | 96,195,161,400 | 320,650,538 | | 2 | round spermatid | SAMN09427369 | SRR7345460 | 38.244654 | 104,105,827,200 | 347,019,424 | | 3 | round spermatid | SAMN09427369 | SRR7345461 | 53.996261 | 144,532,309,500 | 481,774,365 | | 12 | round spermatid | SAMN08375232 | SRR7349192 | 52.384556 | 140,431,608,000 | 468,105,360 | | 29 | sperm | SAMN08375229 | SRR6502360 | 26.653940 | 64,752,370,800 | 215,841,236 | | 30 | sperm | SAMN08375228 | SRR6502362 | 23.973440 | 58,369,232,700 | 194,564,109 | | 13 | sperm | SAMN08375229 | SRR7349193 | 52.806276 | 141,148,572,300 | 470,495,241 | | 14 | sperm | SAMN08375229 | SRR7349195 | 22.444378 | 60,523,788,600 | 201,745,962 | | 15 | sperm | SAMN08375229 | SRR7349196 | 38.253606 | 104,119,671,000 | 347,065,570 | | 27 | spermatogonia | SAMN08375231 | SRR6502356 | 22.845286 | 58,909,579,800 | 196,365,266 | | 28 | spermatogonia | SAMN08375231 | SRR6502357 | 17.947471 | 46,888,332,900 | 156,294,443 | | 4 | spermatogonia | SAMN09427379 | SRR7345462 | 18.686342 | 52,032,780,000 | 173,442,600 | | 5 | spermatogonia | SAMN09427379 | SRR7345463 | 29.956561 | 82,384,836,000 | 274,616,120 | | 6 | spermatogonia | SAMN09427379 | SRR7345464 | 39.145759 | 105,153,716,100 | 350,512,387 | | 7 | spermatogonia | SAMN09427378 | SRR7345465 | 35.816184 | 96,048,594,600 | 320,161,982 | | 8 | spermatogonia | SAMN09427378 | SRR7345467 | 28.396816 | 77,248,140,900 | 257,493,803 | |

### Handling coolers (Or: preparing coolers)

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| ***[A flowchart showing the pipeline from .fastq to .mcool. The first 6 steps were done with a Probably BioRender or Inkscape.]***  Figure 1 |

#### The *gwf* workflow targets

A *gwf* workflow was created to handle the first part of the data processing, and each accesion number (read pair, mate pair) from the Hi-C sequencing was processed in parallel, so their execution was independen on the other samples.

##### Downloading the reads

The reads were downloaded from NCBI SRA portal [ref] directly to GDK using sra-downloader [ref] through docker [ref] as .fastq.gz files.

##### Handling the reference

The latest reference genome for rhesus macaque (*macaca mulata*), *rheMac10* (or *Mmul\_10*, UCSC or NCBI naming conventions, respectively) was downloaded to GDK from UCSC web servers with wget [ref]. To use bwa (Burrow Wheeler’s Aligner) [ref] for mapping, rheMac10 needs to be indexed with both bwa index with the --bwtsw option and samtools faidx, which results in six indexing files for bwa mem to use.

##### Mapping paired-end reads

##### Pair and sort the reads

##### Filter (deduplicate) pairs

##### Create interaction matrices (coolers)

##### Pooling samples (Merging coolers)

The strategy to get the best signal was by pooling the interaction matrices. cooler merge was used to merge all samples in each sub-folder (cell type) to just one interaction matrix for each cell type. The reason for that is that we choose to trust Wang et al. (2019) when they say that compartments are highly reproducible between replicates, and by merging all replicates, we will have a more robust signal.

##### Create multi-resolution coolers (zoomify)

##### Matrix balancing (Iterative correction)

##### Eigendecomposition

## Results

Here are the glorious results

## Discussion

Here is the discussion

## Bibliography

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