Article

# Patterns of Transposable Element Distribution Around Chromatin Ligation Points Revealed by Micro-C Data Analysis

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**Abstract:** Background: Transposable elements constitute a significant portion of eukaryotic genomes, yet their role in chromatin organization remains poorly understood.

Methods: This study computationally investigates the density patterns of transposons around chromatin contact points identified from public Micro-C chromatin conformation data from human cell culture. The density peak patterns of various transposable families and subfamilies were studied within a 100kb window centered on contact points. The analysis was focused on the most abundant transposons, such as Alu and LINE-1.

Results: The computational analysis revealed highly pronounced, non-random density patterns of transposons around the chromatin contact points. The patterns were produced by aligning all ligation points and plotting the average density around them. The patterns were strikingly different between transposable element families and substantially different between the members of the families. The patterns were found to be reproducible across independent studies and biological replicates. Among major families and subfamilies there were no members that didn't have reproducible density patterns around the contact points. Randomly generated coordinates produced less pronounced patterns, which were not correlated between replicates as expected for the negative control. Some families showed enrichment and some - depletion at contact points, while 100Kb window-wide patterns remained correlated between biological replicates. The patterns were oriented relative to the chromosomal orientation. Additionally, the patterns were oriented relative to the transposon sequence direction.

# Keywords: transposable elements, chromatin conformation, chromatin folding, chromatin, Micro-C

### 1. Introduction

Transposable elements (TEs) constitute a significant portion of eukaryotic genomes, often comprising more than half of the genomic content in many species. Once considered "junk DNA", these mobile genetic elements have been increasingly implicated in various aspects of genome function and evolution [1, 2]. Previously, we have emphasized that transposable elements can serve an important positive biological function as anchors and regulators of chromatin folding [3–10]. Recent advancements in chromosome conformation capture techniques, particularly Micro-C and Hi-C, have enabled high-resolution mapping of chromatin interactions [11], revealing the three-dimensional organization of the genome.

The rationale behind our investigation stems from the hypothesis that repetitive ele-42ments, due to their sequence homology, might provide sequence-specific anchors for43chromatin folding. This aligns with emerging views of spatial-temporal genome regula-44

tion, where chromatin organization plays a key role in cellular responses [12]. This hy-45 pothesis is based on the idea that homologous sequence structures may form homolo-46 gous contacts in the nucleoplasm. In other words, two identical double helices would 47 adhere to each other. Since TEs are highly repetitive, we hypothesized that they form 48 adhesive contact points in chromatin folding. To test this, we examined the distribution 49 of TEs around the chromatin contact points identified through micro-C experiments. 50Since the contact points are measured via ligation and sequencing, the exact positions of 51 ligation points (LPs) can be reconstructed from the Micro-C data. Then, we mapped the 52 density patterns of various families and subfamilies of TEs. 53

Our study focuses on two of the most abundant TE families in the human genome: 54 Alu elements, short interspersed nuclear elements (SINEs) approximately 300 base pairs 55 in length, and LINE-1 (L1) elements, long interspersed nuclear elements that can span 56 several kilobases. Additional families were also studied. 57

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### 2. Methods

### 2.1. Public chromatin conformation capture datasets

We analyzed chromatin interaction data from four independent datasets to identify ligation points (LPs). Dataset 1 (DS1) and Dataset 2 (DS2) were produced by Micro-C: SRR12625672 and SRR12625674 (biological replicates from HUDEP cell line, ~43M paired-end reads each, 150nt). https://www.ncbi.nlm.nih.gov/sra/SRR12625672. The Micro-C protocol employs micrococcal nuclease digestion, achieving nucleosome-level (~200bp) resolution.

Datasets 3 and 4 (DS3 and DS4) were produced by Hi-C: SRR27906244 and SRR27906243 from the GM13977 cell line (~75M paired-end reads each). Hi-C libraries were constructed using a standard protocol with restriction enzyme digestion (MboI, MseI, or NlaIII), which typically produces 2-6 kilobase fragments, followed by biotin incorporation and proximity ligation. The Hi-C data were obtained from NCBI Bioproject PRJNA1074296 (https://www.ncbi.nlm.nih.gov/bioproject/1074296) and processed using JuicerTools v1.14.08. Despite the methodological differences between Micro-C and Hi-C, all datasets were analyzed using identical parameters for TE density patterns and correlations.

## 2.2. Identification of ligation points.

Although contact points in the chromatin conformation capture are called points, they are imprecisely defined and are actually represented by short fragments. Yet, it is possible to define precisely to a single nucleotide which sequences were ligated in the assay. The first step in the analysis was to identify the exact positions of the ligation points on the genome. The method identifies ligation points with single-nucleotide precision by using the alignment of paired-end reads to the genome. In Micro-C, most paired reads overlap, forming a continuous sequence. This sequence corresponds to the real physical fragment (called here ligated\_fragment) produced by ligation in chromatin conformation capture. For each paired read, we reconstruct the ligated\_fragment sequence by aligning paired reads. Next, we align the ligated\_fragment against the genome. Only those paired reads are retained where two parts of the ligated\_fragments (called here arms) align to the same chromosome and are positioned at least 1Kb apart. Once the arms are aligned, the exact position of coordinates of the chromosome that were ligated to each other in the assay are identified (called here, LP1 and LP2). The pipeline for the identification of the ligation points was called LigP\_finder\_v2. The Python code for LigP\_finder\_v2 can be downloaded from https://github.com/maxrempel/DRRF/tree/main/LigP\_finder-main

| All datasets were processed using whole-genome alignment in LigP_finder_v2.   | 93   |
|---|--|
| When applied to Hi-C datasets, the program produced fewer ligation points since the reads in Hi-C don't overlap, but the number of found ligation points was sufficient to produce TE density pattern plots and correlations. In general, the amount of data from the experiments was in excess since the plots didn't visually deteriorate even when a single small chromosome was used for pattern plotting. The pattern was consistent across chromosomes, although the density of ligation points varied widely, about 10-fold between chromosomes, reflecting variable compaction of chromosomes in given cell lines.  | 94<br>95<br>96<br>97<br>98<br>99<br>100<br>101                     |
| To test whether the program was correct, we ran tests using synthetic data and did a spot-check of the results. Also, the final program was described in English algorithm, and reprogrammed from scratch by another remote programmer. The recreated program produced identical results, demonstrating that both programs are correct.   | 102<br>103<br>104<br>105   |
| The Ligation points, LPs, were obtained with hg38, unmasked version: https://hgdown-load.soe.ucsc.edu/goldenPath/hg38/bigZips/hg38.fa.gz  | 106<br>107   |
| We used the hg38 transposable element (TE) annotation from UCSC: https://hgdown-load.soe.ucsc.edu/goldenPath/hg38/bigZips/hg38.fa.out.gz  | 108<br>109   |
| 2.3. Plotting density patterns  | 110  |
| For each identified LP, we examined the genomic regions extending 50kb in both direc-<br>tions, which we termed "exbors" (from EXtended harBORS). Within these regions, we<br>cataloged the density of transposable elements using bedtools, and function intersect. In<br>that, the rigth and left-oriented elements were analysed separately. We focused on the<br>most abundant TE families and subfamilies based on their genomic copy numbers.   | 111<br>112<br>113<br>114<br>115                                    |
| To quantify TE density patterns, we calculated TE density in 1kb bins across the 100kb window centered on each LP. Importantly, we conducted separate analyses for elements on the plus and minus strands to investigate potential strand-specific (same as orienta-tion-specific) patterns. To visualize these density patterns, we smoothed the density curves with the Kernel Density Estimation (KDE) smoothing method.   | 116<br>117<br>118<br>119<br>120                                    |
| For negative controls, we performed random control analyses. This involved generating random LP positions across the genome in excess and selecting those random coordinates that were located in non-repeat-masked (unique) regions across the whole genome. Then, real and random coordinates of LPs were compared using density and correlation plots. To control for programming errors, three programmers (DO, AV and MM) reproduced the prlots independently, and the plots came out identical. Correlations between density pattern plots were calculated using the Pearson correlation and visualized as diagonal heatmaps. The Python code for LigP_finder_v2 also contains TE density pattern plotting with KDE smoothing. https://github.com/maxrempel/DRRF/tree/main/LigP_finder-main | 121<br>122<br>123<br>124<br>125<br>126<br>127<br>128<br>129<br>130 |
| Some plots are shown in the Results section, and many more plots are shown in the Supplementary Materials pdf file.   | 131<br>132   |
| 2.4. Analysis of harbor homology  | 133  |
| To measure sequence similarity between harbor pairs, we extracted ±5 Kb sequences around each ligation point (10kb total harbor size). Sequence alignment was performed   | 134<br>135   |

To measure sequence similarity between harbor pairs, we extracted ±5 Kb sequences 134 around each ligation point (10kb total harbor size). Sequence alignment was performed 135 using minimap2 with the 'sr' preset and minimum 70% identity threshold. To compare 136 within-pair versus between-pair homology, we analyzed sequence similarity between 137 harbors from the same ligation point pair (REAL) versus harbors from different ligation138point pairs (CONTROL). Both unique/low-copy sequences (regions remaining after repeat139masking) and total sequence (unmasked) were analyzed. Pairs with overlapping harbors140(distance < 10kb) were excluded from the analysis. The Python code for homology analysis</td>141is available at https://github.com/maxrempel/DRRF/tree/main/LigP\_finder-main.142

# 3. Results

For each tested transposable element (TE) family, we produced TE density patterns 144 in 100 Kb windows (called here "harbors") around chromatin ligation points (called here 145 LPs) identified in Micro-C and Hi-C chromatin conformation assays. We examined the 146 density of six major transposable element (TE) families: Alu, L1 (LINE-1), L2 (LINE-2), 147 Medium Reiterated Frequency Repeat (MER), Mammalian-wide interspersed repeats 148 (MIR), and Mammalian Long Terminal repeat (MLT) elements, as well as most frequent 149 subfamilies the most frequent TEs: L1 and Alu. Figure 1 presents the distribution of L1 150 elements around LPs (ligation points). 151



Figure 1: Distribution of L1 transposable elements around chromatin ligation point. Panel A. The 153 density of L1 elements (a major subfamily of Long Interspersed Nuclear Elements or LINEs) rela-154 tive to chromatin ligation points (LPs) in two experimental Micro-C datasets and their correspond-155 ing random controls. The x-axis represents the position relative to the LP (vertical black line at 156 50,000 bp), spanning 100 kb. The y-axis shows the density of L1 elements (count of L1 elements 157 divided by the bin size, bp). Solid lines represent L1 elements in the plus strand (therefore ori-158 ented left to right), and dotted lines represent the opposite-oriented L1 elements labeled as the 159 minus strand. Dataset 1 (blue/red) and Dataset 2 (green/purple) show experimental data for LP1 160 and LP2 regions. LP1 and LP2 ligation points were numbered from left (start) to right (end) of the 161 chromosome. Random controls 1 and 2 correspond to Datasets 1 and 2, respectively, generated 162 using randomized LP positions. The graphs are oriented from left to right on the chromosome. 163 Panel B shows a correlation heatmap comparing L1 element density patterns between different 164 datasets and strands. The color intensity represents the Pearson correlation coefficient, with lighter 165 colors indicating stronger correlations. The heatmap reveals strong correlations between biological 166 replicates (Dataset1 and Dataset2) for the same strand orientation, while plus and minus strands 167 show notably weaker correlations, quantitatively confirming the strand bias observed in panel A. 168

Figure 1 presents a view of L1 transposable element distribution around chromatin169ligation points (LPs) identified through micro-C experiments. L1, or LINE-1, is a major170subfamily of Long Interspersed Nuclear Elements (LINEs) comprising about 20% of the171human genome. The data reveal patterns that suggest a non-random association between172L1 elements and chromatin structure.173

Key observations from the figure include:

 Pattern reproducibility: Panel A shows a similarity in the distribution patterns between Dataset 1 and Dataset 2. This consistency across independent
 datasets strengthens the biological significance of the observed patterns,
 which is quantitatively confirmed in Panel B by high correlation coefficients
 (>0.8) between datasets for corresponding strands.

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- Strand-specific patterns: The plus and minus strands show distinct patterns 180 (Panel A), which is quantitatively supported by the low correlation coefficients (0.2-0.4) between the plus and minus strands in Panel B.
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- Depletion at the ligation point: There is a noticeable dip in L1 density directly 183 at the ligation point (50,000 bp). This depletion indicates the exclusion of L1 184 elements from immediate chromatin contact points. 185
- Random controls: The random control datasets in Panel A show markedly 186 different patterns characterized by less pronounced fluctuations and lack of 187 correlations between datasets and within pairs of ligation points. This contrast underscores the biological significance of the patterns observed in the experimental data.

These findings suggest a sequence-specific role of L1 elements in chromatin folding, 191 with patterns indicating their selective retention in specific chromatin environments. Future studies should focus on understanding the mechanisms driving these distribution 193 patterns. 194



**Figure 2:** Distribution of Alu transposable elements around chromatin ligation points. The density of Alu elements (a subfamily of Short Interspersed Nuclear Elements or SINEs) relative to chromatin ligation points (LPs) in two experimental datasets. The x-axis represents the distance from the LP (at 0), spanning 100 kb (-50,000 to +50,000 bp). The y-axis shows the density of Alu elements. Solid lines represent the plus strand, and dotted lines represent the minus strand. Dataset 1 (blue/orange) and Dataset 2 (green/red) show data for LP1 and LP2 regions. The y-axis shifts were added artificially to prevent overlap of the curves.

Figure 2 presents the distribution of Alu elements around chromatin ligation points (LPs) identified through Micro-C experiments. Alu elements, the most abundant member of Short Interspersed Nuclear Elements (SINEs), comprise approximately 11% of the human genome.

Key observations from Figure 2 reveal consistent patterns in Alu element distribution 207 around chromatin ligation points. The patterns are consistent across datasets and ligation 208 points. As for other TEs, a strand difference was observed, with the plus strand (represented by solid lines) displaying different density patterns compared to the minus strand 210 (dotted lines). This asymmetry is consistent in all tested datasets and was observed in 211 every other tested TE family. 212

#### 3.1. Additional Transposable Elements

Our analysis of transposable element (TE) distribution around chromatin ligation 214 points (LPs) revealed that each TE family and subfamily has a distinct density pattern 215 around ligation points. We examined the density of six major TE families: Alu, L1, L2, 216 MER, MIR, and MLT elements, as well as several subfamilies within these groups, within 217

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a 100 kb window centered on LPs identified through Micro-C experiments (Fig. 1 for L1, 218 Fig. 2 for Alu; see Supplementary Figures for all tested TE families and subfamilies). 219

The distribution patterns observed in every examined TE family had very similar 220 trends to the trends observed for L1 elements (Fig. 1), although the density patterns 221 around the ligation points were unique for each tested TE family and subfamily. 222

Also, the patterns for all the TE families and subfamilies were asymmetric relative to 223 the ligation points, oriented with respect to chromosome direction (with the short arm (p) 224 positioned towards the left, following standard genomic convention). This asymmetry in-225 dicates that transposon directionality relative to chromosome orientation participates in 226 chromatin folding. The consistency of these patterns suggests that the evolutionary selec-227 tion of transposon placement produced long-range chromosome-wide orientations of 228 transposons. Such large-scale genomic patterns may represent fundamental principles of 229 chromosome organization that persist through evolution. 230

Since every tested TE family and subfamily produced consistent density patterns 231 around the ligation points, this suggests the fundamental role of transposable elements in 232 sequence-specific chromatin folding that transcends the specific characteristics of individual TE types. The observed strand biases and dataset-specific variations appear to be general features of how TEs are distributed relative to chromatin contact sites. 235

To exclude potential methodological artifacts, we utilized Hi-C data (Datasets 3 and 236 4, see Methods) to validate our findings. While both Hi-C and Micro-C capture chromatin 237 conformation, they differ fundamentally in their digestion methods: Hi-C uses restriction 238 enzymes producing fragments of several kilobases, while Micro-C employs micrococcal 239 nuclease digestion, achieving nucleosome-level (~200bp) resolution. Despite this resolu-240tion difference, Hi-C data reproduced the key patterns of transposable element distribu-241 tion around contact points, with correlation coefficients between biological replicates 242 reaching 0.7-0.8 for same-strand comparisons and remaining below 0.2 for opposite 243 strands (Supplementary Figures). This cross-method validation confirms the strand-spe-244 cific organization of transposable elements around chromatin contact points. While abso-245 lute correlation values between datasets from different laboratories and cell lines were not 246 expected due to the tissue-specific nature of chromatin organization, the fundamental pat-247 terns were consistently reproduced: strong correlations between biological replicates and 248 between members of ligation point pairs and strong strand asymmetry. These patterns 249 were reproduced despite using cell lines from different blood lineages (HUDEP erythroid 250 progenitors (datasets 1 and 2) vs. GM13977 lymphoblastoid cells (datasets 3 and 4)), sug-251 gesting that the results reflect a common principle in chromatin organization. 252

# 3.2. Asymmetric Distribution Suggests Homological Adhesion

We noticed that the asymmetry of transposable element density patterns relative to chromosome orientation suggests tandem-like organization. However, we found no periodic repetition of these patterns, suggesting an aperiodic repetition of patterns that maintains directional consistency. Coincidentally, this aligns with Erwin Schrödinger's description of the hereditary material as aperiodic crystal [13].



**Figure 3:** Model of homological adhesion in chromatin folding. The figure shows how patterns of identical transposon sequences (arbitrarily placed colored arrows) can serve as contact points through sequence-specific homological adhesion. The 100 Kb window demonstrates the alignment of identical transposons creating interactions between distant chromatin regions.

We propose that this aperiodic but directional repetition of TE patterns functions in 264 chromatin architecture through homological adhesion, where identical sequences, partic-265 ularly transposable elements, can form contact points through sequence-specific adhesion 266 of identical sequences. Here, we propose for the first time that two parallel double helices 267 of DNA can adhere to each other when their sequences are identical, forming the molec-268 ular basis for homological adhesion. The density plots reveal that the patterns are asym-269 metric and face in one direction on chromosomes. Since they face in one direction on chro-270 mosomes, they could serve as specific anchors for homological adhesion and the for-271 mation of large-scale helices. 272



**Figure 4:** Distribution of sequence homology within and between ligation point pairs. Panel A shows homology in unique and low-copy sequences left after masking. Panel B shows total sequence homology from the unmasked sequence. For REAL comparisons, we measured homology between two harbors from the same ligation point pair (within the LP pair). For CONTROL comparisons, we measured homology between harbors from different ligation point pairs (between LP pairs).

We next tested if the sequences that come together in chromatin display homology. As a negative control, we measured homology between unpaired unligated harbors from 281 different harbor pairs. Comparing harbors (ligated fragments) from the same harbor pair 282 versus different pairs revealed a strong enrichment of homology in ligated regions. In 283 unique and low-copy sequences, within-pair homology reached 70%, while between-pair 284 homology remained near zero (mean 0.47% vs 0.02%). The signal persisted in an un-285 masked sequence, with within-pair homology reaching 100% compared to a maximum of 286 40% between pairs. This provides quantitative evidence for sequence-specific homological 287 dsDNA-dsDNA adhesion in chromatin contacts. 288

### 4. Discussion

The patterns of TE distribution around chromatin ligation points observed here provide insight into the potential role of TEs in chromatin folding. We discovered consistent 291 strand asymmetry across TE families, which indicates systematic evolutionary pressures 292 governing transposon organization. While general principles of transposon insertion have 293 been described [14], this chromosome-wide directional bias has not been previously reported. 295

The density patterns around ligation points were distinct for each TE family. Alu 296 elements showed proportionally scaled patterns with multiple peaks, reflecting their 297 higher copy number, while L1 elements displayed fewer but more pronounced peaks. 298

## 4.1. Homological adhesion

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The discovery of the density patterns of transposable elements around the chromatin 300 contact sites and of the homology between contacting DNA sequences offers support for 301 homological adhesion. We will nickname it "homadhesion" for brevity. Homadhesion is 302 the adhesion of two DNA duplexes (double helixes) to each other if they have similar 303 sequences. 304

Several studies provide support for sequence-dependent attraction between DNA 305 duplexes. Yoo et al. [15] demonstrated sequence-dependent attraction between doublestranded DNA molecules using molecular dynamics simulations and single-molecule 307 FRET experiments. They showed that DNA duplexes can attract each other over distances 308 up to 2-3 nm in the presence of polyamines like spermine. Importantly, AT-rich sequences 309 showed stronger attraction than GC-rich sequences, and DNA methylation enhanced 310 these interactions. 311

Lee et al. [16] provided theoretical support for sequence-dependent attraction between intact DNA duplexes. Their model incorporated electrostatic forces and sequencedependent DNA shape variations, predicting lower interaction energies for pairs of DNA fragments with parallel homologous sequences compared to those with uncorrelated sequences. This interaction was modeled without strand separation, considering the aqueous environment through parameters that account for electrostatic screening in solution. 317

Barzel and Kupiec reviewed the evidence for the pairing of similar DNA sequences 318 across different organisms [17]. They noted that in yeast, matching DNA sequences can 319 locate each other and recombine efficiently even when in different genomic locations despite the large amount of genomic DNA present. Their review suggested that similar sequences are paired as part of the genome's basic organization. 322

For the adhesion mechanism itself, several forces could be considered: electrostatic 323 (including ionic), hydrophilic-hydrophobic, hydrogen bonding, van der Waals, and other 324 weak forces. The dynamic nature of chromatin condensation and decondensation sug-325 gests that DNA duplexes remain nearly intact during these interactions. While hydrated 326 DNA structure may be primary in mediating adhesion, other nucleoplasmic components 327 likely participate, including histones, other proteins, and low molecular weight chemicals. 328 The highly negatively charged DNA duplexes must overcome electrostatic repulsion to 329 adhere. This requires neutralization by positive ions present in the nucleoplasm, including 330 protons (H+), hydronium ions (H3O+), positively charged histones and other proteins, and 331 ions such as Na<sup>+</sup>, K<sup>+</sup>, Mg<sup>2+</sup>, and polyamines like spermine and spermidine. 332

## 5. Conclusions

This discovery of consistent density patterns of transposable elements around chromatin contact points suggests a functional involvement of TEs in sequence-specific chromatin folding and homological adhesion. The observed asymmetry of patterns relative to chromosome orientation and transposon orientation suggests the existence of chromosome-scale sequence organization that was not previously known. 334

This opens new opportunities for research into the role of transposable elements in 339 sequence-specific chromatin folding and, through that, in genome regulation. The organ-340 ization of chromatin through networks of interacting domains has been previously de-341 scribed [18], though not through the lens of transposon-mediated interactions we report 342 here. Future work should focus on elucidating the mechanisms underlying the observed 343 density patterns and investigating their potential implications for genome regulation and 344 function. Exploring these patterns across different cell types, developmental stages, and 345 organisms could provide valuable insights into the evolutionary conservation and func-346 tional significance of TEs in chromatin folding [19]. 347

The sequence-specific principles of genome organization are known to operate across 348 multiple scales, from local loop formation to chromosome territories. Dixon et al.[20] 349 demonstrated how chromatin organization changes systematically during cellular differentiation. The importance of repetitive elements in nuclear organization was highlighted 351 by Cournac et al.[21], showing correlations between similar repetitive elements and 3D 352

|     | <ul> <li>folding patterns. Studies of chromosome territories by Cremer et al. [22] revealed principles of nuclear organization. Rowley et al. [23] established evolutionarily conserved principles of 3D chromatin organization. The emerging picture of genome architecture was developed by Bonev et al. [24], who mapped dynamic changes in genome organization during development. Our finding of chromosome-orientation-dependent patterns in TE densities around contact points suggests an additional layer of sequence-encoded structural information that may help explain these organizational principles. As our understanding of genome organization continues to evolve, it is becoming increasingly clear that a comprehensive view of genomic function must include consideration of transposable elements. This work lays the foundation for future investigations of the role of transposable elements in chromatin folding.</li> <li>Acknowledgments. We thank Jekaterina Erenpreisa for the discussion. The work and the DNA Resonance Research Foundation (DRRF) were funded by MM. The publication fee was paid with AV's reviewer's vouchers. AV acknowledges support of his work by the RUDN USAL Program.</li> </ul> | <ul> <li>353</li> <li>354</li> <li>355</li> <li>356</li> <li>357</li> <li>358</li> <li>360</li> <li>361</li> <li>362</li> <li>363</li> <li>364</li> <li>365</li> <li>366</li> </ul> |
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|     | Author contributions: AV, DO, and MM did genomic computation.  | 367   |
|     | <b>Conflicts of Interest</b> : The authors declare no conflicts of interest.   | 368   |
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# Supplement

Alu mobile element density around LP names Exbor1\_Dataset1 Exbor1\_Dataset1 Exbor1\_Dataset2 Exbor2\_Dataset2 Density strand + -10000 0 10000 Distance from ligation point -50000 -40000 -30000 -20000 

Fig. S\_ALU







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Fig. S\_MIR



**Fig. S\_MLT** 450













Fig. AluSg



Fig. S\_AluSx



Fig. S\_L2a



Fig. S\_L2b



Fig. S\_L2c



Fig. S\_MIR subfamily



Fig. S\_MIRb



Fig. S\_MIRc



Fig. S\_DS3-DS4 Transposon L2b DS3 is correlated with DS4 (strands averaged, strand info ignored)



Fig. S\_DS3-DS4 Transposon L2b DS3 is correlated with DS4 (strads plotted separately)

# Correlation

The following graphs are correlations for density curves between Datasets D1, D2, D3, D4. The transposon subfamily is marked on the top of each graph.

Correlation Analysis of Transposable Element Distributions These correlation plots demonstrate the reproducibility of transposable element density patterns around chromatin contact points across different datasets and methods. Each plot shows pairwise correlations between biological replicates, with lighter colors indicating stronger correlations. Plus (+) and minus (-) strands are analyzed separately to reveal strand-specific patterns. Correlation values between corresponding strands of biological replicates typically reach 0.7-0.8, while correlations between opposite strands remain below 0.2, quantitatively confirming strand specificity. Results are shown for Alu (X), L1 (Y), L2a-c (Z) and their subfamilies. The consistent patterns across Hi-C (DS3, DS4) and Micro-C (DS1, DS2) datasets validate the biological authenticity of these organizational features.



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# Random Controls

| Correlation plot for AluJb mobile element |                   |                    |                   |                    |                   | - 1.0              |                   |                    |       |
|---|-------------------|--------------------|-------------------|--------------------|-------------------|--------------------|-------------------|--------------------|-------|
| LP1_Dataset1_plus                         | 1                 | -0.049             | -0.078            | -0.013             | 0.0086            | -0.1               | -0.11             | 0.32               |       |
| LP1_Dataset1_minus                        | -0.049            | 1                  | -0.032            | -0.12              | 0.077             | 0.15               | -0.023            | -0.099             | - 0.8 |
| LP2_Dataset1_plus                         | -0.078            | -0.032             | 1                 | -0.073             | 0.099             | 0.082              | 0.024             | -0.035             | - 0.6 |
| LP2_Dataset1_minus                        | -0.013            | -0.12              | -0.073            | 1                  | 0.091             | -0.058             | -0.067            | -0.095             | - 0.4 |
| LP1_Dataset2_plus                         | 0.0086            | 0.077              | 0.099             | 0.091              | 1                 | 0.17               | -0.075            | 0.032              | 0.4   |
| LP1_Dataset2_minus                        | -0.1              | 0.15               | 0.082             | -0.058             | 0.17              | 1                  | -0.21             | -0.097             | - 0.2 |
| LP2_Dataset2_plus                         | -0.11             | -0.023             | 0.024             | -0.067             | -0.075            | -0.21              | 1                 | 0.051              | - 0.0 |
| LP2_Dataset2_minus                        | 0.32              | -0.099             | -0.035            | -0.095             | 0.032             | -0.097             | 0.051             | 1                  | - 0.2 |
|   | LP1_Dataset1_plus | LP1_Dataset1_minus | LP2_Dataset1_plus | LP2_Dataset1_minus | LP1_Dataset2_plus | LP1_Dataset2_minus | LP2_Dataset2_plus | LP2_Dataset2_minus | -0.2  |

# Fig.S. Random controls