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Rouse model with transient intramolecular contacts on a timescale of seconds recapitulates folding and fluctuation of yeast chromosomes

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ABSTRACT

DNA folding and dynamics along with major nuclear functions are determined by chromosome structural properties, which remain, thus far, elusive in vivo. Here, we combine polymer modeling and single particle tracking experiments to determine the physicochemical parameters of chromatin in vitro and in living yeast. We find that the motion of reconstituted chromatin fibers can be recapitulated by the Rouse model using mechanical parameters of nucleosome arrays deduced from structural simulations. Conversely, we report that the Rouse model shows some inconsistencies to analyze the motion and structural properties inferred from yeast chromosomes determined with chromosome conformation capture techniques (specifically, Hi-C). We hence introduce the Rouse model with Transient Internal Contacts (RouseTIC), in which random association and dissociation occurs along the chromosome contour. The parametrization of this model by fitting motion and Hi-C data allows us to measure the kinetic parameters of the contact formation reaction. Chromosome contacts appear to be transient; associated to a lifetime of seconds and characterized by an attractive energy of −0.3 to −0.5 k_BT. We suggest attributing this energy to the occurrence of histone tail-DNA contacts and notice that its amplitude sets chromosomes in ‘theta’ conditions, in which they are poised for compartmentalization and phase separation.

INTRODUCTION

The genome is organized into higher-order functional domains and territories (1) that contribute to the regulation of gene expression and to cell differentiation (2). In interphase of budding yeast cells cycle, chromosomes adopt a Rabl-like conformation (3), which is characterized by the positioning of centromeres and the nucleolus at opposite ends in the nucleoplasm. This large-scale organization has adequately been reproduced in silico by modeling chromosomes as homogeneous polymers (homopolymers) with some structural constraints associated to peripheral tethering of telomeres and centromeres (4–7). A detailed analysis of chromosome fluctuations with the Rouse model, which is a generic polymer model relevant to crowded environments, further corroborated the relevance of this homopolymer model in yeast (8). Indeed, the dynamics of chromosome loci, as measured by the temporal variation of the mean square displacement (MSD), appeared to follow an anomalous response characterized by:

\[
\text{MSD} (\tau) = \Gamma \tau^\alpha
\]  

(1)
with $\alpha$ the diffusion exponent, $\Gamma$ the amplitude, and $\tau$ the time interval. The exponent $\alpha$ was in the range of 0.4–0.6 and the amplitude $\Gamma$ appeared to be homogeneous in the genome except for the proximity of telomeres and centromeres (9–16). Accordingly, the Rouse model predicts that the MSD should increase over time with an exponent of 0.5 and that the amplitude of fluctuations is determined by the local properties of chromosomes, specifically the bending flexibility and solvent friction resisting monomer displacements (17). Using the Kuhn length $b$, which is twice the persistence length of the polymer, as a proxy of the bending stiffness, and the monomer friction coefficient $\zeta$, the Rouse model predicts the MSD in 2D:

$$\text{MSD} (\tau) = \frac{16b^2k_BT}{3\pi\zeta} \tau^{0.5}$$

with $k_B T$ the Boltzmann thermal energy. Note that this expression is obtained with the assumption that the chain contains a large number of monomers and behaves as a 'phantom polymer', i.e. without effects of volume exclusion between monomers. Taking the standard approximation that the friction coefficient $\zeta$ is proportional to the Kuhn length (17), the amplitude of MSD fluctuations $\Gamma$ is expected to provide a means to infer the bending flexibility of chromosomes in vivo. However, direct adjustment with the Rouse theory showed that the in vivo value of $\Gamma \approx 0.01 \mu m \cdot s^{-0.5}$ corresponded to a Kuhn length as small as 1–5 nm (9). This estimate appeared to be incompatible with structural and mechanical models of nucleosome fibers (18), as well as the recurrent detection of irregular 10-nm fibers by electron microscopy (EM) of thin nuclear sections (19–21). This inconsistency shows that the precise mechanisms underlying chromosome motion remain largely unclear.

The study of chromosome higher-order structure based on chromosome conformation capture techniques (specifically, Hi-C) has recently shed light on the contribution of two essential mechanisms, which are not included in the simple Rouse model: active processes involving loop extrusion or transcription, and segregation through the random association of monomers along the chain (22–24). Theoretical studies have suggested that both mechanisms could modulate the amplitude of chromatin fluctuations (25,26). Consequently, we were motivated to clarify if and whether additional mechanisms had to be integrated into the Rouse model in order to recapitulate chromosome dynamics in yeast. In order to test polymer models in controlled settings, we first design a biomimetic system to validate the relevance of the Rouse model to analyze DNA and chromatin fluctuation in bulk. This study shows that fluctuation data are quantitatively fitted with the Rouse model parametrized with mechanical parameters of DNA and chromatin. Conversely, in living yeast, the inconsistency of the Rouse model to fit fluctuation and Hi-C data forced us to introduce an equilibrium polymer model, the RouseTIC model, in which random Transient Internal Contacts along the chromosome contour were considered. After a rigorous fitting of the kinetic parameters of the chromosome contact formation reaction, we show that this model recapitulates Hi-C and MSD measurements. Chromosome contacts appear to be transient, lasting a few seconds, and characterized by an interaction energy in the range of $-0.3$ to $-0.5 \ k_BT$. We finally speculate that these labile interactions along the chromosome contour may originate from unspecified histone tail interactions and constitute a key mechanism for the formation of dynamic compartments in eukaryotes.

### MATERIALS AND METHODS

#### DNA and chromatin preparation and characterization

Unless stated, chemical and biochemical reagents were purchased from Sigma-Aldrich or New England Biolabs, respectively. Chromosome fragments were purified from U2OS cells, which were synchronized at the beginning of S phase, then released in the presence of dUTP-Cy3 and scraped from glass surfaces in order to allow fluorescent nucleotide entry into the cells and incorporation into the genome during DNA replication (see (27) for details of the protocol). Cells were eventually embedded in agarose plugs, and treated with 5% SDS and 100 $\mu g/ml$ proteinase-K during two days in order to purify genomic DNA. The resulting chromosome fragments were recovered from agarose plugs using $\beta$-agarase treatment during 2 h. Nucleosome arrays were assembled with a reconstitution kit (Active Motif) using $\sim 10$ ng of purified chromosome fragments mixed with 1 $\mu g$ of unlabeled $\lambda$-DNA.

Structural characterization of chromatin fibers was carried out by electron microscopy using the protocol of (28). Briefly, 5 $\mu l$ of reaction product diluted to $\sim 10^{-12}$ M in 10 mM Tris–HCl pH 7.5, 5 mM MgCl$_2$, 50 mM NaCl was deposited on a 600-mesh copper grid covered with a thin carbon film, activated by glow-discharge in the presence of pentyamine. Grids were washed with aqueous 2% (w/v) uranyl acetate, dried and observed in the annular dark-field mode, using a Zeiss 902 transmission electron microscope. Images were captured at a magnification of 85 000 or 140 000 with a Veletta CCD camera and analyzed with iTEM software (Olympus Soft Imaging Solution).

For imaging experiments, DNA or chromatin fibers were diluted 1000-fold in a low salt buffer (Tris–HCl 89 mM, 89 mM boric acid, 2 mM EDTA; pH 8.3) supplemented with 360 or 40 kDa PVP (Sigma-Aldrich). For DNA motion analysis, the viscosity was 5.4 or 2.3 mPa.s with the same mass:volume fraction of 2% of 360 or 40 kDa PVP, respectively. The 360 kDa PVP concentration was set to 3.2% and the viscosity to 15 mPa.s for chromatin loci tracking experiments. The overlapping concentration of these PVP solutions has been estimated in ref. (29). Note that very low concentrations of 100 nm carboxylated polystyrene fluorescent beads (Invitrogen) were added to the buffer in order to measure the viscosity by particle tracking.

#### Yeast culture

We used a fluorescent yeast strain containing one fluorescent locus with 224 tetO sequences to maximize signal-to-noise ratio located at genomic position 530 kb on chromosome XII (13). Cells were grown overnight at 30°C in YP media containing 2% glucose. They were then diluted at 10$^6$ cells/ml and harvested when OD$_{600}$ reached 4 $\times$ 10$^6$ cells/ml. Cells were spread on slides coated with corresponding patch containing 2% agar and 2% glucose. Cover
Microscopy and trajectory analysis

DNA/chromatin or chromosome loci were tracked with a Zeiss AxioObserver or a Nikon TI-E/B inverted microscope, respectively. The Nikon status was equipped with an EM-CCD IxonULTRA DU897 camera (Andor), a 488 nm laser illumination (Sapphire 488, Coherent), and 100× oil immersion objective (numerical aperture = 1.4) with a 1.5× magnification placed ahead of the camera. Pixel size was 107 nm. The Zeiss status was equipped with an Zyla 4.2 sCMOS camera (Andor), a Sola light engine (Lumencor), a 40X oil immersion objective (NA = 1.4) placed ahead of the camera. Pixel size was 162.5 nm.

Trajectories with more than ~100 consecutive positions were subsequently extracted with the TrackMate Plugin (30). We processed the MSD and the kurtosis K, as defined by (31):

\[
\text{MSD}(\tau) = \left\langle (x(t) - x(t + \tau))^2 \right\rangle
\]

\[
K(\tau) = \left\langle \left( (x(t) - x(t + \tau))^4 \right) - \text{MSD}(\tau)^2 \right\rangle / \text{MSD}(\tau)^2
\]

with \(\tau\) the time interval. We also extracted the step distribution function (SDF) for a given time interval \(\tau\). For each trajectory, these MSD and kurtosis were computed for \(\tau\) smaller than 30% of the total duration of the track in order to obtain statistically significant values in single cell datasets (32).

Monte Carlo Simulations of nucleosome arrays

The chromatin fiber was modelled as a string of coarse-grained DNA linkers and nucleosome core particles following the methodology described in (33). The nucleosomal DNA consists of 14 segments of 10.5 bp, anchored to the histone core by 14 minor groove locations, usually referred as Super Helix Locations (34). The linker DNA is made of an integer number of identical segments, each one containing a number of base pairs as close as possible to 10.5 bp. For example, we modeled a 22 bp linker with two segments of 11 bp or a 18 bp linker with two segments of 9 bp. Note that the length \(l\) of each linker segment is proportional to the number \(n\) of base pairs with \(l = n \times 0.32 \text{ nm}\) and the helical angle between consecutive base pairs is 34°. The articulation between linker DNA segments is modelled as a ball-and-socket joint with an energy penalty corresponding to the bending and twisting restoring torques of the DNA double helix. The bending energy is \(E_b = g_b(1 - \cos \theta)\) where \(\theta\) is the bending angle between two consecutive segments and \(g_b\) is calculated from \(E_b = (b - l)/(b + l)\) with \(b = 100 \text{ nm}\) and \(L\) the Langevin function. The twisting energy is given by \(E_t = g_t \phi^2/2\) in the harmonic approximation where \(\phi\) is the twist angle of two consecutive segments. The twisting energy constant \(g_t\) between two connected segments of length \(l\) is given by \(g_t/k_B T = l_r/l\) with \(l_r = 95 \text{ nm}\) the twist persistence length of naked DNA. At each step of the simulation, we chose one segment at random and rotated all the next segments around a random axis with a random angle (Pivot move). The resulting conformation was accepted following the Metropolis criterion.

Extraction of chromatin Kuhn length from nucleosome array Monte Carlo simulations

We used the end-to-end distance distribution as readout of Monte-Carlo simulations to extract the Kuhn length of every nucleosome array configuration (see details in Supplementary Material 1). Briefly, we assumed that nucleosome arrays behaved according to the Worm-Like Chain model, which is defined by the persistence length \(l_p\) as half the Kuhn length and the contour length \(L\) of the chain. We developed a specific approach to estimate these two parameters independently from the analysis of the end-to-end distance distribution based on (35). The second and fourth moments of the distribution can indeed be expressed as a function of \(L\) and \(l_p\), and are the most relevant whenever the contour length \(L\) is comparable to \(4l_p\). Therefore, we simulated sufficiently, but not overly, long fibers with up to 100 nucleosomes and then, using an inversion method, extracted both \(L\) and \(l_p\) from the second and fourth moments. In order to limit end effects, we performed this extraction for central sub-chains containing 5 to 95 nucleosomes (Supplementary Figure S2). The stationary value of the persistence length detected for a contour length \(L \sim 4l_p\) was used to estimate \(l_p\).

Kinetic Monte Carlo simulations of coarse-grained homopolymers with internal contacts

We used a coarse-grained self-avoiding homopolymer model for chromosome structure and fluctuations, as described in (36). Briefly, we modeled a typical chromosome containing \(10^6\) bp by a polymer chain of \(N\) monomers, each of size \(b\) and corresponding to \(n\) base pairs \((N^*n = 10^6)\). The dynamics of the chain were simulated on a face-centered-cubic lattice with periodic boundary conditions using a simple kinetic Monte-Carlo scheme with local moves: at each trial move, a monomer is randomly picked and randomly displaced to a nearest-neighbor (NN) site on the lattice. The move is accepted if the trial position is empty and maintains the chain connectivity. The density of monomer in the simulation box was set to 15% to effectively account for the presence of other chromosomes.

For the simulations of the Rouse model with transient internal contacts, we assumed that two monomers that occupied NN sites have the possibility to form a rigid pair at binding rate \(k_b\) and that this complex can be disassembled with a rate of unbinding \(k_u\). We assumed that a monomer can be involved at most in only one pair. A paired monomer could then move only if it remained connected to its partner, i.e. only if its new location remained a NN site of its partner.
For each investigated situation, we initiated the system with random configurations and then let it reach equilibrium before measuring the average contact probabilities, the mean squared distances, mean squared displacements and kurtosis by averaging over 1000 trajectories. The time step of the simulation was 0.55 ms.

In order to evaluate the effective energy of contact between monomers, we used the detailed balance of the Monte Carlo simulation between a micro-state $i$ where two monomers were NN on the lattice and a micro-state $j$ where they were not in contact. According to Boltzmann statistics, we have $P(i \to j)/P(j \to i) = \exp(\frac{E_0}{k_R T})$. Considering reaction rates $k_b$ and $k_u$, the detailed balance yields during a time step $dt$:

$$P(i \to j) = \frac{1}{dt} \left( \frac{k_b}{k_u + k_b} \right) k_u dt$$

$$+ \left( \frac{k_u}{k_u + k_b} \right) (1 - k_b dt) = \frac{k_u}{k_u + k_b}$$

leading to $E_0 = k_R T \log \left( \frac{k_u}{k_u + k_b} \right)$.

RESULTS
DNA flexibility can be determined from the analysis of its fluctuations

We first wished to validate our assumption that the bending stiffness of a polymer, equivalently its Kuhn length, can be inferred from the analysis of its spatial fluctuations with the Rouse model. Because DNA Kuhn length of $b \sim 100$ to 110 nm in low salt conditions is well-documented in the literature (37), we used naked DNA as a model biopolymer. In order to generate long deproteinized DNA molecules containing randomly-distributed short fluorescent stretches, we purified chromosome fragments from human osteosarcoma cells treated with fluorescent nucleotides during DNA replication (see Materials and Methods). The size of the stretches of $\sim 50$ kb and the contour length of the molecules of 500 kb or more have been characterized by DNA combing and optical mapping in nanochannels, respectively (27). Labeled DNA molecules were then dispersed in a crowded solution containing a high concentration of the neutral polymer poly-vinylpyrrolidone (PVP), equal to three times the overlapping concentration, i.e. the threshold concentration for which PVP chains become intermingled (Figure 1A, (29)). DNA loci movements were finally recorded during $> 60$ images using wide field fluorescence microscopy at an inter-frame interval of 39 ms. The MSD in 2D was finally extracted for each trajectory (gray datasets in Figure 1B).

The average MSD response shown with the red dataset in Figure 1B was consistent with the Rouse model using anomalous diffusion parameters ($\Gamma, \alpha$) of $(0.26 \pm 0.02 \mu m^2/s^{0.53}, 0.53 \pm 0.03)$, as plotted with the corresponding black curve, using 95% confidence intervals for the uncertainty. Taking $b \sim 100$ nm for DNA (37) and using the friction coefficient of a cylindrical monomer of diameter $d \sim 2$ nm for DNA $\zeta = 3\pi n b ln(b/d)$ with $n$ the solvent viscosity of 5.4 mPa.s, we deduce from equation (2) that $\Gamma = 0.23 \mu m^2/s^{0.5}$, in good agreement with experimental data. More quantitatively, we could estimate the Kuhn length by measuring the sum of the squared residuals between the fit and the MSD curve (Figure 1C), which reached a marked minimum for $b = 119$ nm. In addition, the same model enabled fitting the step distribution function (SDF) for three time intervals of 0.12, 0.23 and 0.35 s (Supplementary Figure S3).

We next performed a series of experiments using a PVP concentration three times lower than the overlapping concentration (see Methods), i.e. in conditions where crowding becomes negligible because the PVP chains are no longer intermingled (38). For this, we used the same PVP concentration but the chains were characterized by a molecular weight 10 times smaller. The resulting viscosity of $\eta = 2.3$ mPa.s was expectedly lower. In this regime where hydrodynamic interactions between DNA monomers contribute to the dynamics of the chain, the Zimm model is expected to be valid with the corresponding temporal variation of the MSD (39):

$$\text{MSD}_b(\tau) = \frac{2b}{\tau^2} \times \left( \frac{\sqrt{3\pi k_b T}}{\eta} \right)^3 \times \exp(-\tau/\tau^b)$$

with $\Gamma(1/3) = 2.679$. Note that equation (6) is dependent on the viscosity $\eta$ and unaffected by DNA persistence length. The Zimm model was in agreement with our measurements (green line in Figure 1D), whereas the Rouse model yielded an inconsistent prediction using a Kuhn length of 100 nm (black curve in Figure 1D). In addition, the Zimm model was not consistent with the dynamics of DNA in crowded conditions (green curve in Figure 1B). We note that the consistency of the Zimm model to analyze fluctuation data in ‘uncrowded’ conditions was already reported in published fluorescence correlation spectroscopy studies (40,41). We conclude that the motion of DNA loci can be accurately analyzed with polymer models, and that DNA bending stiffness can be inferred from realtime microscopy studies using the Rouse model in the case of crowded environments.

Nucleosome array fluctuations are consistent with the Rouse model

We then focused on the analysis of nucleosome array fluctuations in vitro. Because the Kuhn length of chromatin remains debated in the literature (18), we separated the presentation of experimental results from the analysis of the amplitude of spatial fluctuations with the Rouse model. We assembled nucleosome arrays on the same long DNA fragments with fluorescent stretches using human core histones and chromatin assembly factors to produce nucleosome arrays with a repeat length of $\sim 168$ bp, as characterized by micrococcal nuclease digestion (Supplementary Figure S4). This nucleosome repeat length matches that of yeast chromatin, which is known to have an average nucleosome repeat of 167 bp that is equivalent to 20–22 bp of linker DNA (42). Electron microscopy revealed the characteristic pattern of 10 nm fibers, with gaps and random clustering of assembled chromatin (Figure 2A, see Methods). The resulting chromatin fibers were diluted in a crowded buffer, and the motion of chromatin loci was tracked by microscopy to measure the MSD and the SDF (Figure 2B, C). The aver-
Figure 1. Analysis of DNA fluctuations in vitro. (A) Schematics of the experiment: long DNA molecules with fluorescent stretches, which are shown with black and red segments, respectively, are dispersed in a crowded solution composed of a high concentration of PVP chains (gray lines). (B) The graph presents the temporal evolution of the MSD as a function of time for fluorescent DNA loci (gray dataset). The average response (red squares) is fitted with the Rouse model (black line), but not with the Zimm model (green line). (C) The graph shows the residuals of the fit with the Rouse model as a function of the Kuhn length. (D) The red dataset represents the average MSD over time for DNA loci dispersed in a solution without crowding. The green and black solid lines show the predictions of the Zimm and Rouse models, respectively, obtained by setting the Kuhn length to 100 nm and the viscosity to 2.3 mPa.s.

By plugging the measured MSD in Figure 2B, we obtained a good fit between the experimental SDF and that predicted by the Rouse model (curves and corresponding fits in Figure 2C). Finally, exploiting the spontaneous destabilization of nucleosome arrays in diluted conditions (43), we compared the fluctuations of nucleosome arrays after assembly, and two days after their reconstitution, and detected a motion characterized by an amplitude \( \Gamma \) 3.5-fold enhanced (data not shown). This response matched that of purified DNA molecules in these experimental conditions, showing that the amplitude of the fluctuations of nucleosome arrays are lower than that of DNA.

Inferring the mechanical properties of nucleosome arrays to test the Rouse model

In order to check the relevance of the Rouse model for chromatin, we had to determine the monomer friction coefficient \( \zeta \) and the Kuhn length \( b \) from structural models of nucleosome arrays and integrate these parameters in equation (2). In order to sample the conformational space of nucleosome arrays thoroughly, we performed simulations with the nucleosome conformation of the crystal structure, i.e. with two turns of DNA wrapped around the histone core, hereafter referred to as the ‘closed negative’ structure (see Materials and Methods, Figure 3A, (44)). We also considered an ‘open’ nucleosome state, in which the most distal histone–DNA binding sites at super-helix location ±6.5 are disrupted (34), as detected in molecular biology assays (44) as well as single molecule techniques ((45), Figure 3B). For both nucleosome configurations, nucleosome arrays were simulated with variable lengths of DNA linkers in the range 18 to 22 bp, i.e. for nucleosome repeat lengths of \( \sim 167 \pm 2 \) bp. Finally, in order to account for the structural defects of nucleosome arrays observed by electron microscopy (Figure 2A), we introduced some randomness in the positioning of nucleosomes associated to a deviation
Figure 2. Nucleosome array fluctuations in vitro. (A) Electron micrographs of chromatin fibers reconstituted on λ-DNA molecules. (B) The graph shows the temporal evolution of the MSD of chromatin fibers dispersed in a crowded solution with individual trajectory reported in gray and the average response in red with the standard error. The viscosity of the solution was 15 mPa.s. The black curve is the fit with a power-law response characterized by an anomalous exponent of 0.54. (C) The plot shows the step distribution function extracted from the same set of data as in (B) for three consecutive time lags indicated in the caption. The corresponding Gaussian fits are directly expressed from equation (7).

Figure 3. Inferring the mechanical properties of nucleosome arrays. (A) Structure of a nucleosome array with closed negative nucleosomes and a linker length of 20 bp. Red spheres correspond to histone octamers. Gray segments represent DNA and the blue segments specifically show entry exit DNAs. (B) Structure of a nucleosome array with open nucleosomes and a linker length of 20 bp. (C) Structure of a nucleosome array with open nucleosomes and a variable linker length of 20 ± 5 bp. (D) The graph shows the end-to-end distance distribution as extracted from the simulation for a nucleosome array containing 20, 30, and 60 closed negative nucleosomes and a linker length of 20 bp. (E) The plot shows the contour length vs. Kuhn length deduced from the analysis of the moments of the end-to-end distance distribution. The colored dots correspond to the fit obtained with chromatin fibers containing 5–95 nucleosomes (plotted with dark blue to red colors, respectively). The three arrows show the results inferred from the analysis of the histograms shown in (D). The red line corresponds to the conditions for which the contour length is equal to two times the Kuhn length.
range of 3 or 5 bp, i.e. linkers of 17 to 23 bp or 15 to 25 bp, respectively (Figure 3C). For each condition, we simulated 1000 independent chromatin fibers structures containing 100 nucleosomes. Note that nucleosome-nucleosome interactions were neglected in the simulations because our experiments have been carried out in low salt conditions with minimal oligomerization or intramolecular contacts, as inferred from analytical sedimentation (46).

We then extracted the Kuhn length of each array using the end-to-end distance distribution as readout of the simulations (Figure 3D, see Materials and Methods). We found that the Kuhn length was larger for closed negative vs. open nucleosomes, altogether spanning 85–140 nm (third column of Table 1 and Supplementary table S2). Specifically, for regular arrays of negative nucleosomes with a repeat length of 165–169 bp, the Kuhn length was roughly constant in the range 120–140 nm, whereas fibers with open nucleosomes appeared to be slightly more flexible with a Kuhn length spanning 85–110 nm (Supplementary Figure S2 and Table S1). These values of the Kuhn length matched the predictions of other simulations for closed negative nucleosomes in the range of 102–154 nm for nucleosome repeat length of 169 and 168 bp, respectively (47,48). Measurements based on Hi-C in yeast also indicated that the Kuhn length was in a comparable size range of 110–230 nm (49). Last but not least, analytical models support the stiffness of fibers with short linkers of ~20 bp with a Kuhn length of 100 nm or more (50).

These simulations also provided an estimate of the density of nucleosomes per 11 nm knowing the contour length of less than 5 nm. We tested the consistency of this estimate with published Hi-C data in the form of probability of contact normalized at 20 kbp vs. genomic distance (black, blue, red and burgundy curves in Figure 4B). For this, we ran a coarse-grained lattice-based polymer simulation (see Methods) using $n = 200$ bp per monomer, a Kuhn length $b = 5$ nm, and the concentration of DNA set to $3 \times 10^{-3}$ bp/nm$^3$ as input parameters (53). We observed that the occurrence of distant contacts in the simulation was underestimated by more than one order of magnitude (cyan curve in Figure 4B). Furthermore, despite the contradiction with MSD data, we could run the simulation for a Kuhn length of $b \sim 50$ nm for $n = 2$ kb per monomer, because this value is more consistent with in vitro data (49). Yet, this parametrization only marginally improved the fitting of Hi-C data (green curve in Figure 4B). Last, using our large ensemble of displacement measurements on chromosome XII, we compiled the SDF and detected a deviation to the Gaussian behavior expected for the Rouse model with a peaked distribution for time intervals of 0.4 and 0.8 s (Figure 4C, D). Overall, the Rouse model appeared to show some inconsistencies with structural and motion data in living yeast, suggesting that additional molecular parameters should be integrated in the model.

Transient chromatin contacts account for chromosome dynamics and folding in yeast

Recent EM studies of thin nuclear sections of budding yeast have hinted to the existence of intra-chromosomal contacts (54). Although intramolecular contacts have not been thoroughly discussed in yeast, their contribution to higher-order
Table 1. Mechanical, structural and hydrodynamic modeling of nucleosome arrays by Monte Carlo simulations using different nucleosome conformations and repeat lengths. The last column corresponds to the amplitude of chromatin loci fluctuations according to the Rouse model.

<table>
<thead>
<tr>
<th>Nucleosome configuration</th>
<th>Linker size (bp)</th>
<th>Kuhn length (nm)</th>
<th>Nucleosome per 11 nm</th>
<th>Hydrodynamic diameter (nm)</th>
<th>MSD amplitude $\Gamma$ ($\mu$m$^2$/s$^{0.5}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Closed negative</td>
<td>20</td>
<td>118 ± 6</td>
<td>2.6</td>
<td>215 ± 15</td>
<td>0.057 ± 0.03</td>
</tr>
<tr>
<td>Open</td>
<td>20</td>
<td>110 ± 3</td>
<td>1.8</td>
<td>96 ± 9</td>
<td>0.079 ± 0.06</td>
</tr>
<tr>
<td>Closed negative</td>
<td>20 ± 3</td>
<td>96 ± 5</td>
<td>3.0</td>
<td>228 ± 13</td>
<td>0.045 ± 0.03</td>
</tr>
<tr>
<td>Open</td>
<td>20 ± 3</td>
<td>106 ± 5</td>
<td>2.0</td>
<td>112 ± 9</td>
<td>0.071 ± 0.05</td>
</tr>
<tr>
<td>Closed negative</td>
<td>20 ± 5</td>
<td>92 ± 5</td>
<td>3.4</td>
<td>305 ± 14</td>
<td>0.037 ± 0.02</td>
</tr>
<tr>
<td>Open</td>
<td>20 ± 5</td>
<td>95 ± 3</td>
<td>2.4</td>
<td>145 ± 10</td>
<td>0.056 ± 0.05</td>
</tr>
</tbody>
</table>

Figure 4. Analysis of MSD and Hi-C data in yeast with the Rouse model. (A) The graph shows 110 MSD responses in gray, the average response in red with the standard error to the mean, and the corresponding fit with the Rouse model in black. (B) The plot presents normalized contact frequency vs. genomic distance inferred from Hi-C measurements by Mercy et al. in red (68), Duan et al. in black (69), Marie-Nelly et al. in blue (70) and Belton et al. in burgundy (71). The orange sector serves to define the limits of experimental responses. The green and cyan datasets are the results of simulations for one homopolymer freely diffusing in space and characterized by a Kuhn length of 50 or 5 nm, respectively. (C) The plot presents the SDF at four consecutive time lags, as indicated in the legend. The trajectories are the same as in A. (D) Same data as in (C) in log-lin.

chromosome folding in mammalian cells or drosophila has been proposed many times (23,55,56). We thus set out to explore if a Rouse model with transient internal contacts (RouseTIC) could fit chromosome structure and motion in vivo. We used the same simulations scheme, but introduced non-specific self-interaction in between the monomers (36). We defined $k_b$ and $k_u$ the binding and unbinding rates between monomers, respectively. In addition to these two kinetic parameters, the RouseTIC model relies on the definition of a concentration of DNA set to $3 \times 10^{-3}$ bp/nm$^3$, the Kuhn length and the characteristic time for the fluctuations.

In order to estimate $k_b$ and $k_u$, we first performed simulations of the contact probability normalized at 20 kb versus genomic distance and compared them with Hi-C measurements. These simulations were performed setting the number of base-pair by Kuhn length to 2 kb and 50 nm, i.e. with a density of 2.6 nucleosomes/11 nm that roughly match the results of our simulations in Table 1 and those
Figure 5. Analysis of MSD and Hi-C data in yeast with the RouseTIC model. (A) The graph shows the comparison of experimental and simulated Hi-C contact frequency versus genomic distance. Experimental data are within the two orange limits defined in Figure 4B. The output of simulations fit experimental data for \( k_b/k_u \) in the range 0.3 to 0.7, as indicated in the legend at the top. (B) The plot shows the Kuhn length as a function of \( k_b/k_u \) that best fits 3D mean square distance measurements with the RouseTIC model. The inset shows an example of fit for \( k_b/k_u = 0.7 \). Note the error bars for the determination of the Kuhn length corresponds to acceptable fits within the gray limits in the inset. (C) The orange dataset shows the experimental MSD versus time for nucleosome arrays in vitro, as reported in Figure 2B, and the corresponding black curve is obtained with the RouseTIC model for a simulation time step of 0.55 ms. (D) The plot presents the MSD as a function of time for chromosomes in living yeast. The dashed orange line is the experimental measurement of \( 0.01 \times t^{0.54} \) and the upper and lower gray lines correspond to limits characterized by \( 0.013 \times t^{0.54} \) and \( 0.008 \times t^{0.54} \). The blue, black and green curves are obtained from the RouseTIC model with \( k_b/k_u = 0.7 \) and different values of \( k_u \), as indicated in the legend.

of experiments of \( \sim 2.0 \) nucleosomes/11 nm based on Hi-C (49). Notably, Hi-C data map the steady-state conformation of a polymer; hence the occurrence of contacts can be recapitulated with one equilibrium kinetic parameter, namely the reaction constant \( k_b/k_u \). These simulations were in good agreement with the experiments for \( k_b/k_u \) in the range 0.3–0.70 (Figure 5A). As an argument of self-consistency, we checked the validity of the RouseTIC model by comparing its predictions to distance measurements in between chromosome loci in fixed cells derived from ref. (57) (Figure 5B). For \( k_b/k_u \) spanning 0.3–0.7, we adjusted the Kuhn length and confirmed that its value of \( b = 53 \pm 5 \) nm was compatible with the data.

We then wished to analyze motion data. In order to adjust the time step of the simulation, we ran simulations with no interaction and compared their predictions to our in vitro MSD measurements (Figure 5C). The temporal parameter of the simulation was 0.55 ms. We finally focused on the motion of chromosome loci in living yeast with the RouseTIC model and \( k_b/k_u \) in the range 0.3–0.70 (Figure 5D). Because the amplitude of fluctuations appeared to primarily depend on the unbinding rate \( k_u \), we could determine both reaction kinetic parameters. We proceeded by setting \( k_b/k_u \) in the range 0.3–0.7 and then varied \( k_u \) to fit MSD data (Figure 5D and Supplementary Figure S5). In this parameter space, the typical unbinding time \( 1/k_u \) varied from tens of seconds to seconds (inset of Figure 5D). Notably, the fit of MSD data was significantly better for \( k_b/k_u \) of 0.7 than 0.3 (see Supplementary Figure S5).

In order to reinforce the relevance of the RouseTIC model, we focused on higher-order moments of the SDF (Figure 4C, D). The anomaly of the SDF with respect to
Figure 6. Analysis of chromatin motion based on the kurtosis. (A) The kurtosis is plotted as a function of time for the locus on chromosome XII or for reconstituted nucleosome arrays (blue and black datasets, respectively). The dashed black line corresponds to the response for a Gaussian process, as in e.g. Brownian motion or the Rouse model (31). The dashed gray and blue curves are deduced from simulations of the RouseTIC model using kinetic parameters that fit Hi-C and MSD data, which are reported in the caption. (B) Same graph as in (A) with the two experimental datasets divided by 0.9.

Figure 7. Molecular scenario for the RouseTIC model. (A) The sketch represents the Rouse TIC model with the definition of the reaction rates $k_u$ and $k_b$. The fiber displayed in the inset is obtained for $k_u/k_b = 0.5$. (B) Simulation of a chromatin fiber with a linker size of 20 bp and closed negative nucleosomes with a concentration of nucleosomes set to $3 \times 10^{-3}$ bp/nm$^3$. 127 nucleosomes are highlighted in green because they potentially form intramolecular contacts, as defined by an inter-nucleosome distance in $\mu m$ of <15 nm, i.e. comparable to the typical length of histone tails, and a genomic distance in cis of more than 1650 bp (10 nucleosomes). (C) The same simulation is carried out for a fiber with linker size of 22 bp and open nucleosomes. 456 nucleosomes are marked in green.
the Gaussian response can be measured from the kurtosis, which is computed from the fourth-order moment of the SDF (see Equation (4) in Materials and Methods). In the Rouse model, the kurtosis is equal to 2 ((31), dashed black line in Figure 6A), whereas an excess of kurtosis is detected for step distribution functions more peaked than a Gaussian function, as noticed for time intervals of 0.4 and 0.8 s in Figure 4C. Expectedly, the trajectory-averaged kurtosis appeared to be equal to 2.4 in the small time limit for chromosomes in vivo (blue solid line in Figure 6A). In addition, we noticed a clear temporal signature in the kurtosis, which decreased after 1 to 3 s. Conversely, the kurtosis measured for purified chromatin fibers was lower and equal to \( \sim 1.75 \), and roughly constant over a narrower temporal window (black solid line in Figure 6A).

We finally computed the kurtosis of the simulated trajectories with the RouseTIC model. We used the fitted parameters of \( k_b/k_u = 0.4 \) and 0.7 and \( k_u = 0.2 \) and 1 s\(^{-1}\), respectively, because these values fit MSD and Hi-C data. Simulations showed an excess of kurtosis equal to \( \sim 2.7 \) in the small time limit (dashed lines in Figure 6A), and a decrease after a typical time defined by the unbinding reaction time scale \( 1/k_u \). These predictions were in qualitative agreement with our data, as could be clearly evidenced by the phenomenological normalization of in vitro and in vivo kurtosis data by the same factor of 0.9, and the resulting superposition of experiments and simulations (Figure 6B). Altogether, the analysis of the kurtosis tends to support the consistency of the RouseTIC model to reproduce structural and dynamic data of yeast chromosome in vivo with a steady-state polymer model that includes two kinetic parameters for the contact formation reaction.

DISCUSSION

Origin of molecular interactions in RouseTIC

In this study, we monitored chromatin motion in vitro and in vivo by fluorescence microscopy, and analyzed experimental data with chromatin structure models and kinetic Monte Carlo simulations of self-interacting polymers. We propose that the Rouse model with transient internal contacts accounts for the in vivo dynamics of chromosomes (Figure 7A) and determines the kinetic parameters of the association reaction for the first time in living yeast. Internal contacts appear to be detectable from the occurrence of frequent contacts along the chromosome contour (Figure 4B) and the slow motion of chromatin loci with a deviation of the step distribution from the Gaussian response in the short time limit (Figure 4C). The latter consequence is qualitatively explained by the formation of transient contacts that tend to decrease the fraction of time a monomer freely diffuses, and restrain the displacement of freely diffusing monomers inside boundaries along the chromosome contour (25). Interestingly, the existence of these transient contacts is reminiscent of the mechanism of internal friction (58), which manifests itself if intramolecular conformational barriers lead to the dissipation of energy into internal degrees of freedom. Already proposed to be relevant for chromosomes (59), internal friction is known to lead to slow fluctuations, as observed in our datasets.

Nevertheless, the molecular origin of the contacts in the RouseTIC model remains to be clarified. By simulating the structure of a fiber with 1000 nucleosomes at a volume density of 15% with the same structural models of chromatin as for deriving the Kuhn length, we suggest that numerous contacts are likely to pre-exist within chromosomes (Figure 7B, C), which could be stabilized by nucleosome interactions. By scaling the effective energy of the binding events, the RouseTIC model can help narrow down possible molecular origin of these interactions. In addition to the transient nature of binding events with a lifetime on the order of seconds, the effective energy \( E_0 \) of contacts between monomers, as determined from the ratio of binding and unbinding rates \( k_b/k_u \) (see Equation (5)), is weak in the range of \( E_0 \sim -0.3 \) to \( -0.5 \) k_BT. Such strength for \( E_0 \) is far smaller than the energy associated with ATP hydrolysis of \( \sim 20 \) k_BT, but also than the attractive stacking energy between nucleosomes of \( -2.7 \) k_BT (60). Thus, this energy cannot be directly attributed to stacking between two nucleosomes through specific contacts, such as those described between the histone H4 N-terminal tail of one nucleosome with the ‘acidic patch’ of an adjacent one (61). Rather, we hypothesize that contacts predominantly correspond to transient, weak and labile interactions. Interestingly, an interaction energy \( E_0 \) of \( -0.3 \) to \( -0.5 \) k_BT compares to the conditions of equilibrium between histone–tail DNA attraction and DNA–DNA electrostatic repulsion in physiological salt conditions, as inferred from X-ray scattering of nucleosome core particles (62). Thus, contacts may originate from unspecific and dynamic histone tail–DNA contacts within the chromatin fiber as observed in recent cryo-EM studies on nucleosome core particles (63), potentially mediated or perturbed by chromatin-binding proteins. The higher chromatin density in the nucleus of living cells in comparison to in vitro experiments is expected to favor the formation of contacts. Yet the contribution of these contacts on chromatin fluctuations in vitro may be investigated by modulating the ionic strength of the solution. Using analytical centrifugation, it has indeed been shown that addition of 2 mM divalent salt enhances chromatin compaction and oligomerization mediated by histone tails interactions (46,64).

Relevance of contact reaction for the analysis of Hi-C data in metazoan

RouseTIC may also be relevant to chromatin compartment segregation, in which the occurrence of chromosome intramolecular contacts mediated by nucleosomes has been suggested (21). Compartments inferred from Hi-C contact maps could be modeled assuming limited attraction between the monomers (23,24), although the amplitude and origin of this interaction remains mostly elusive. The estimate of monomer-monomer interaction energy in the range of \( E_0 \sim -0.3 \) to \( -0.5 \) k_BT is notably close to the energy between monomers of \( -0.3 \) k_BT for a polymer chain in theta conditions (65), i.e. with balanced monomer-monomer and monomer-solvent interactions. Because polymers in theta conditions are poised to fold into globules or coils whether monomer-monomer interaction increases or decreases, respectively, we suggest that the residual interactions between
monomers can be modulated to trigger compartment segregation. This mechanism is likely necessary but not sufficient to model metazoan datasets due to the essential contribution of active processes involving loop extrusion or transcription (22–24). Nevertheless, our recent analysis of chromosome structure in drosophila appears to corroborate the globule to coil transition (data not shown), suggesting that nucleosome-nucleosome interactions stabilization by e.g. heterochromatin-specific proteins, such as HP1, could promote local collapse into globular state and phase separation (66,67). Despite a strong impact on chromosome large-scale organization, the number of proteins implicated in this structural transition could be limited. This hypothesis deserves further investigation using heteropolymer models (24) and combining them to chromatin structure models in order to reach a molecular description of segregation.

**SUPPLEMENTARY DATA**

**Supplementary Data** are available at NAR Online.

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**CONFLICT OF INTEREST STATEMENT**

None declared.

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