# Telomere Structure Estimation

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# 1 Introduction to telomere biology

Telomeres consist of DNA tandem repeat sequences, their associated binding proteins, and a non-coding RNA transcript. They are located at the end of chromosomes and address two important problems in eukaryotes: the end-replication problem and the end-protection problem. A nice summary is provided in [1].

#### 1.1 Telomere composition and structure

In humans, the telomere tandem repeat sequence is TTAGGG. A telomere's size lies between 5 and 15 kb in humans. A key feature of the telomere end in all organisms is the 3' single-stranded G-rich overhang. In mammals, these overhangs are 30-500 nucleotides long [1].

Associated with the telomeric DNA are proteins of the shelterin complex. Shelterin is responsible for solving the end-protection problem, though it acts in a complicated manner.

The six known shelterin proteins are TRF1, TRF2, RAP1, TIN2, TPP1, and POT1. TRF1 and TRF2 bind to the duplex region of the DNA; POT1 coats the overhang with oligonucleotide/oligosaccharide binding folds. TIN2 has a bridging function since it interacts simultaneously with TRF1, TRF2, and TPP1.

In addition to the shelterin proteins, there are telomere-associated proteins that are recruited by shelterin to act as "accessory factors."

The long non-coding RNA might be important for telomere maintenance and function, though little is known about the underlying molecular mechanisms.

## 1.2 Telomeres and the end-replication problem

The telomerase enzyme solves the end-replication problem and helps counteract telomere erosion. The mechanisms underlying telomere length regulation in mammlian cells are not fully understood, though some key factors are known.

For example, overexpressing TRF1 in cancer-derived human cells leads to telomere shortening, whereas depleting telomeres of TRF1 results in elongation. Reducing levels of TRF2 and TIN2 leads to telomere extension, but also to heterochromatic marks that might offset the extension. There are also repressive histone marks on telomeric chromatin. Deletion of these marks correlates with large elongations of telomeres.

#### 1.3 Mammalian telomeres have nucleosomes

Mammalian telomeres are enriched with repressive histone marks, including H3K9me3 and H4K20me3 and the heterochromatin-specific factors chromobox homolog (CBX) 1, 3, and 5. Depleting these marks correlates with telomere elongation.

#### 1.4 Project goals

- 1. Establish a rigorous quantitative method for determining the size of chromatin structures from STORM data.
- 2. Use the method to determine the distribution of physical telomere sizes in Hela L(ong) and Hela S(hort) phenotypes.
- 3. Demonstrate how the genomic length of Hela telomeres correlates to physical size.
- 4. Determine how variation of shelterin composition at the Hela telomere affects the physical size.
- 5. Use the measured size distributions and polymer modeling to estimate the degree and nature of telomere packaging.

6. Correlate changes in telomere polymeric properties to histone marks acquired after knockdown of shelterin components.

# 2 Super-resolution imaging of telomeres

Telomeres in two phenotypes of Hela cells were imaged using the Biop's N-STORM microscope. DNA FISH was used to attach 18 bp labels to the telomeric repeats. The fluorescent label was Cy5. 3D imaging was achieved using a cylindrical lens in the Fourier plane of the microscope output, but the third dimension was not used in the end.

A single field of view consisted of about four cells on average and about twenty or thirty labeled telomeres. 10,000 images were taken for each field of view after bleaching Cy5 into its transient off-state. Single molecule localizations were extracted from the images using Nikon's NIS-Elements Ar software (possibly NIS-A N-STORM Analysis as well).

#### 2.1 Data processing of STORM

The localization data for each field of view was analyzed by grouping localizations into clusters using the DB-SCAN clustering algorithm. Before performing the full analyses, we found that a minimum number per cluster of 8 and neighborhood radius of 65 nm sufficed to successfully identify clusters and avoid artifacts and noise. In Fig.1, the number of identified clusters as a function of DB-SCAN clustering parameters is plotted. The plateau is the ideal area of the parameter space because neighboring telomeres are not grouped into one cluster and sparse clusters are still identified.

# 2.2 Assessing telomere size

The properties of STORM measurements prevent the accurate determination of any single telomere structure. The reason for this can be seen in Fig.2. A small portion of the DNA polymer is illustrated in A. However, one only has access to the data seen in B, which represents the individual localizations found during imaging. The localizations only reflect the true structure to a degree due to random labeling and finite localization precisions. An additional source of randomness comes from the fact that the telomere genomic length is not a single number, but is also random as well. The problem is essentially this: which of the many contours that can be traced through the localization point data is the real chromatin conformation? Even with extremely small localization precisions, random labeling of the polymer would still prevent answering this question exactly.

Because the real telomere conformation cannot be determined from STORM, we instead choose to analyze a statistical measure from the ensemble of telomeres, the **radius of gyration**,  $R_g$ . This has two advantages: 1) it satisfies our intuitive notion of size for a polymer, and 2) the mean  $R_g$  of the ensemble has an analytical solution from polymer physics, which we can use to constrain our models.

Note that we are not measuring **mean end-to-end distances** like in many earlier works, (c.f. [2]). In these works, specific genomic sequences could be labeled and the histogram of distances between two labeled sequences accurately determined. Because the telomeres are genomic repeats and because of stochasiticity of the labeling mentioned above, the radius of gyration is a more appropriate measure of size.  $R_g$  is also more amenable to assessing the telomere structure because the genomic distances are much smaller than in previous works (telomeres are at most 35 kb, whereas previous work explored genomic distances on the order of 100 kb.) Genomic and physical

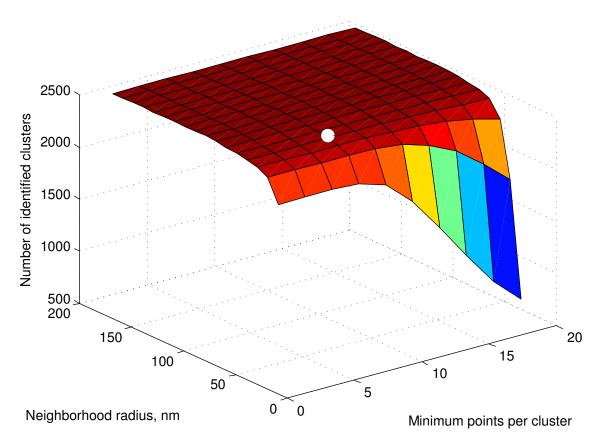


Figure 1: The number of identifiable clusters using DB-SCAN in the original Hela L dataset as a function of the minimum points per cluster and the neighborhood radius. The white circle marks k = 8 and  $\epsilon = 65$ , the values used in all datasets. The plateau indicates the optimum region because different telomeres are not grouped into the same cluster whereas spare clusters are still identified.

distances between molecules are too small to accurately determine their separation distributions at these scales.

We should also note that the localization precision of the measurements in the z-direction is about 50 nm. This is of the same order of magnitude as the telomere size. For this reason, we analyze the projected radii of gyration, i.e. the radius of gyration computed from the localizations' x- and y-coordinates only. For a polymer in thermodynamic equilibrium, the projected gyration radius is equal to the true, three-dimensional gyration radius up to a constant factor of  $\sqrt{3/2}$  [3].

# 3 Polymer modeling of telomeric chromatin

Polymer simulations are now becoming useful tools for modeling experimental data on chromatin and predicting other structural features of DNA [3, 4].

The difference between our work and the recent chromatin simulations of topologically associating domains (TAD's) is primarily one of scope: in [4], the authors model chromatin conformations on the 100 kb scale by allowing for independent interaction energies between any two sections of the fiber, essentially including the heterogeneity of the fiber as a central aspect of their model. This

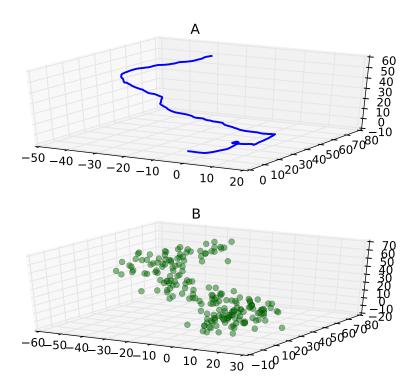


Figure 2: A) An illustration of a portion of telomeric DNA. B) Due to a finite localization precision and random labeling of the structure, the individual localizations only marginally reflect the underlying DNA structure.

is important for studying interactions between enhancer/promoter pairs, which should have more favorable interaction energies than with other sections of the fiber. However, the telomeres are composed of genomic repeats and are therefore much more homogeneous. Their smaller size also restricts the number of conformations available for them to adopt.

For these reasons, we modeled the polymer as simpler a homogeneous wormlike chain for which only two parameters are needed to describe the conformational ensemble, the packing ratio c and the persistence length  $\ell_p$ . The packing ratio is the number of base pairs per nanometer. To simulate a wormlike chain, we also needed the genomic length of telomeres which, when multiplied with c gives the fiber's contour length.

#### 3.1 Theory of the wormlike chain and a simulation approach

In the simplest WLC model, we treat the polymer as a semiflexible and homogeneous rod with a negligible thickness and a length  $L_c$ , known as the contour length. The flexibility of the rod is described by its persistence length  $\ell_p$ . Intuitively, the persistence length is the average length over which the polymer remains approximately straight. Polymers with a longer persistence length will be more rigid than shorter ones.

Mathematically, the persistence length is the characteristic length describing the exponential decay of the tanget-tanget correlation function [5],

$$\langle \mathbf{t}(s) \cdot \mathbf{t}(0) \rangle \sim \exp(s/\ell_p)$$
 (1)

where  $\mathbf{t}(s)$  is the unit vector tangent to the polymer at the one-dimensional coordinate s along the polymer. For distances s much greater than  $\ell_p$ , (1) states that there will be no correlation in the direction that the tangent vectors point.

The subunits that make up most polymers are very small molecules and are thus subject to agitation by the random collisions with other molecules in their environment. This is especially pertinent to polymers in aqueous environments, where collisions with the solvent molecules cause the polymer to change shape and conformation many times a second. According to Boltzmann's statistics, the probability that a semiflexible polymer in thermodynamic equilibrium will be found in one of any of its possible conformations at an instant in time is proportional to the Boltzmann factor

$$P(U) \sim \exp\left(-\frac{U}{k_B T}\right)$$
 (2)

where P(U) represents of the probability of observing a polymer conformation with associated internal energy U,  $k_B$  is Boltzmann's constant and T is the temperature of the system. The fact that it takes energy to bend the polymer into a particular conformation reflects the "semiflexible" qualities of the polymer.

The energy U required from the environment to achieve a given conformation can be determined by dividing the polymer into many short sections such that it can be represented as the summation of the energies of many small circular arcs. The energy required to bend a rod through an angle  $\theta$ with Young's modulus E, moment of inertia I is

$$U = \frac{EI}{2s}\theta^2 \tag{3}$$

The persistence length is related to the rod's mechanical properites by [5]

$$\ell_p = \frac{EI}{k_B T} \tag{4}$$

To simulate a wormlike chain, it suffices to create small linear segments with angles between the segments governed the probability distribution obtained by substituting Eq. 3 into Eq. 2 and using the relationship above for the persistence length. This will produce a Gaussian distribution, which is easily sampled on a computer. To produce a 3D chain, one must also include a random rotation of uniform probability between 0 and  $2\pi$  along the axis of one of the segments. Doing so effectively extends the simulation of 2D chains described in [3].

#### 3.2 Strategy for estimating telomere polymer properties

We want to estimate the packing density c and the persistence length  $\ell_p$  for the telomeres. This would allow us to quantitatively understand how shelterin modulates their physical structures.

We have the distributions of measured 3D gyration radii for about 1000 telomeres in each of many conditions, including wild type and shelterin knockdowns of different proteins.

Our strategy is to simulate a large number of polymers with input parameter-pair values  $(c, \ell_p)$  that we can choose. The genomic lengths of the telomeres are constrained by the mean genomic length and the minimum and maximum values, all three of which can be estimated from the Southern blot data. (Gaining more precise genomic length distribution information is not possible due to the nature of the Southern blot measurement). After simulating each polymer, it is "labeled" with flourophores whose positions have been randomly bumped to reflect the estimated localization imprecision of the measurement (see Fig. 2 above). The radius of gyration for the displaced fluorophores is measured, and a probability distribution of measured radii of gyration for all simulated and sampled polymers is constructed.

The simulated probability distribution for every parameter pair  $(c, \ell_p)$  is then compared to the measured STORM distribution using maximum likelihood reconstruction. The goal is to find which pair of polymer properties was most likely to have produced the measured dataset.

In other words, we're searching for the values of c and  $\ell_p$  that produce a probability distribution of gyration radii that best matches the measured distribution.

#### 3.3 Preliminary results for telomere maximum likelihood reconstruction

In Fig. 3, we plot the log-likelihood function for each pair of values for the packing density and persistence lengths. The redder the color, the higher the log-likelihood value, and the better the  $R_g$  distribution for that pair of parameter values matches the measured distribution.

There are a few things to note here: for one, contours of constant log-likelihood are very crooked. This is actually because of under-sampling the conformational distribution in the simulations and because a discrete number of parameter-pair values were simulated. 1000 conformations were simulated for each pair of parameter values in this plot. Prior simulations with 50,000 conformations produced much smoother contours. Also, in Fig. 4, we mark what pair of values were simulated. Once the full parameter space is mapped like this, we can explore the important regions by doing a finer sampling over a smaller region.

Importantly, the region near  $c=30\,bp/nm$  and  $\ell_p=50\,nm$  has a high likelihood value. This means that there should be a good overlap between the simulated  $R_g$  distributions and the measured STORM  $R_g$  distributions. To show this, we plot the measured and simulated distributions for  $\ell_p=50\,nm$  with  $c=30\,bp/nm$  and  $c=50\,bp/nm$  in Figs. 5 and Fig. 6, respectively. A much better agreement between the simulated model and the measured data is seen in the first figure.

These values allow us to make a comparison with known chromatin models. For example, the 10 nm fiber has a packing density of 15 bp/nm and persistence length of 30 nm. The 30 nm fiber has a packing density of 100 bp/nm and a persistence length of 200 nm. This preliminary data suggests that the telomere packing is actually much closer to the 10 nm fiber. Our initial estimates also show that Hela S cells have a lower packing density than Hela L, which may be due to the higher amounts of shelterin in Hela L. We do not yet have the simulation results yet to show this.

### 4 What's next?

### 4.1 Refine the parameter estimations

First, we need to run the simulations with a larger number of steps (about 100,000 compared to 1000). This will take several days, so Christmas break is an ideal time to do this. This will allow

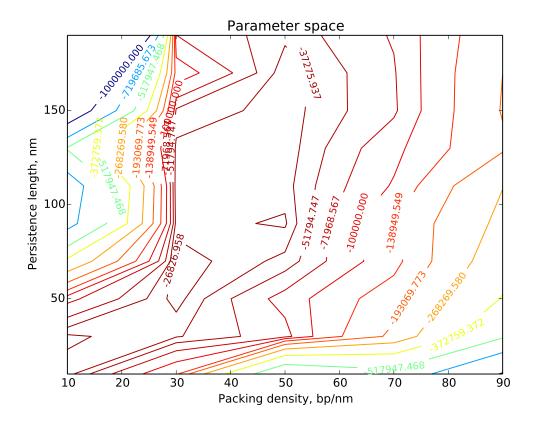


Figure 3: Log-likelihood that the simulated distribution of  $R_g$  values matches the measured one from STORM. Redder colors denote higher log-likelihoods. In particular, regions near c = 30 and  $\ell_p = 50$ , as well as c = 50 and  $\ell_p = 90$  match the measured data well.

us to better refine the parameter space plots and identify what regions need to be sampled with a finer grid.

We also need to determine the uncertainty in the parameter value estimates. Judging by the extreme differences between distributions when  $c=30\,bp/nm$  and  $c=50\,bp/nm$ , it looks like the uncertainty in the packing density is going to be small.  $\ell_p$  cannot be so easily constrained, though, in part because the telomeres are so small. The poor constraint on one parameter could be expected since c and  $\ell_p$  are actually only independent parameters when the polymer contour length is shorter than the persistence length. When the contour length is much greater than the persistence length, the only parameter governing polymer configuration is the ratio  $\frac{c}{\ell_p}$  [6]. At this point, the wormlike chain becomes the freely-jointed chain.

One way to address uncertainty in maximum likelihood estimates is to compute the local curvature of the parameter space at the point of the best estimate. This involves taking the second derivative of the parameter space and requires a finely sampled grid at the maximum. Due to the prohibitive computation times involved, we're delaying this uncertainty estimation for now and instead looking at the overlap between measured and simulated distributions to qualitatively judge the uncertainty in the parameters.

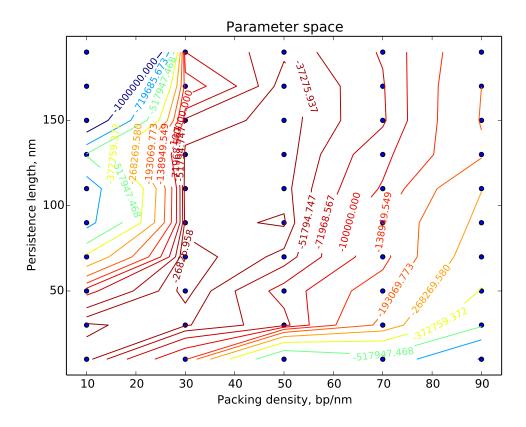


Figure 4: Same as above, but with a grid denoting what parameter-pair values were simulated.

#### 4.2 Improve simulation speeds

A portion of the simulation code can be rewritten to utilize parallel processing. On my 12-core machine, this should result in a 12-times reduction in the computation time. This isn't so important now and will likely be done after the break, since I can get a number of simulations to finish with the current code over Christmas.

#### 4.3 Compare parameter value estimations for Shelterin knockdowns

We've observed that the mean radius of gyration is always smaller when TRF2 is knocked down. There is also a larger number of heterochromatic marks on the telomeres in TRF2 knockdowns, which suggests that charge screening on the histones should result in telomere compaction. In Fig. 7, we show measured mean gyration radii for TRF1, TRF2, and TRF1/TRF2 double knockdowns, compared to the control pSuper. It also intersting to note that TRF1 knockdowns result in telomere decompaction, which might be seen in the figure as well, though the results are less obvious and the cause less clear [1].

We would like to quantitatively assess exactly how much more compact the TRF2 knockdowns are than the controls, which is reflected in the value for the packing density. At this point, though, it's unclear whether we can say that there is a statistically significant difference in packing densities

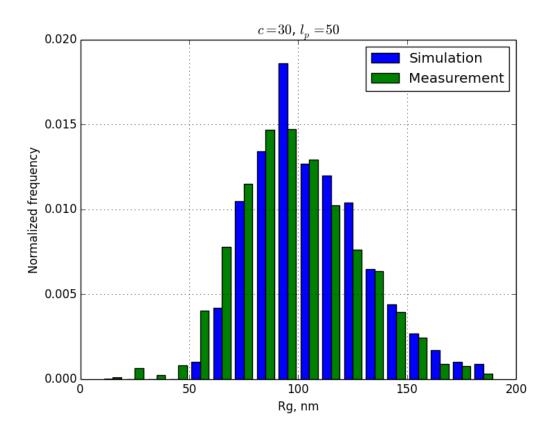


Figure 5: Measured and simulated  $R_g$  distributions for  $c = 30 \, bp/nm$  and  $\ell_p = 50 nm$ .

since the change in mean size is rather small.

If we are however capable of determining the change in packing density, this will be the first physical measurement (to my knowledge) of small scale compaction in chromatin due to heterochromatic marks that are caused by charge screening on the histones.

## 5 Additional features not included in the model

#### 5.1 Excluded volume interactions

The polymer/STORM model I've developed does not take into account excluded volume interactions, though these have been suggested to be important for some chromatin fiber interactions.

The reason for this is that excluded volume interactions would require an additional parameter in the model, which is the radius of the excluded volume around a polymer segment. However, I know of no constraints on this value that we can use, except for maybe the known diameter of the 10 nm and 30 nm fibers. This is a poor constraint because it only reflects hard-core electrostatic interactions, not the the longer-range interactions that would more likely determine the polymer conformation.

I also have good reason to think that excluded volume interactions are not going to matter in telomeres anyway. We're finding a minimum persistance length consistent with the data of about 50

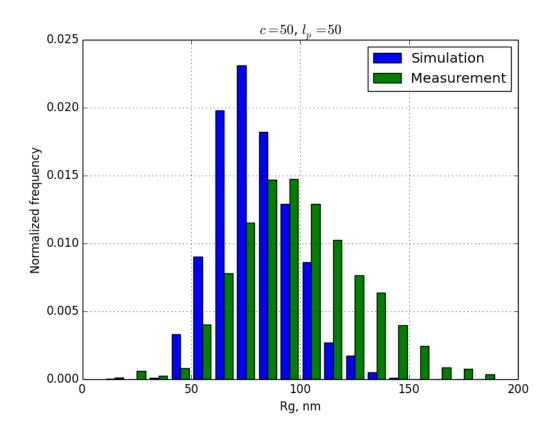


Figure 6: Measured and simulated  $R_g$  distributions for  $c = 50 \, bp/nm$  and  $\ell_p = 50 nm$ .

nm. Since the telomeres are only about 200 nm in diameter, the chromatin fiber probably doesn't bend over onto itself in this condition because it's too stiff, and thus there simply aren't enough conformations in the ensemble where the fiber overlaps itself.

## 5.2 T-loops

We are not including the T-loop in the polymer model. The T-loop is a structure where the single strand overhang folds back and inserts itself into the double-stranded portion of the telomere. I have been told that these have only been observed in vitro (c.f. the work from Prof. Zhuang's lab [7], where they isolated telomeres and spread them out on a surface before imaging). Additionally, our biology colleagues seem to think the T-loop is some sort of fascination of other groups and that their existence is in vivo is questionable.

I think it might be challenging to address this point. Our results might in fact conflict with [7] since we see a shrinking telomere when TRF2 is knocked down. They see in vitro that the T-loop is not formed when TRF2 is knocked down, and I suspect (though on very limited grounds) that T-loop removal in vivo would cause an increase in telomere size. This point really needs to be thought about before we submit, but I haven't received much help on this point.

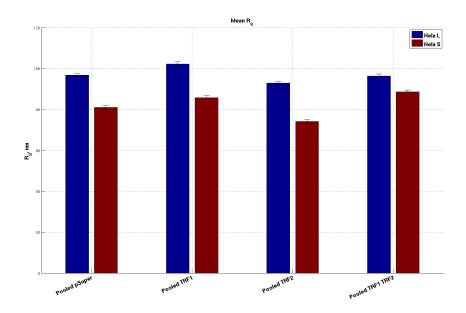


Figure 7: Mean gyration radii for the TRF1, TRF2, and TRF1/TRF2 knockdown experiments. The control is pSuper. Error bars are standard errors of the mean.

### 5.3 G-quadruplex

There is a higher-order structure in the telomere called a G-quadruplex [8]. It is formed by interactions between guanines in double stranded DNA. My limited knowledge on the subject is that the 4-guanine interaction is weak, that G-quadruplex existence is questionable by some in vivo, and that the relatively harsh conditions of FISH preparation would essentially remove these formations during labeling, if they did indeed exist.

Still, we also have to address this point, like the T-loops. One way is to suggest that the estimated packing density c effectively includes contributions from the G-quadruplexes. After all, we measure a packing density that's higher than the 10 nm fiber; this increase could reflect more compaction due to the formation of G-quadruplexes.

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