

Structure determination of genomes and genomic domains by satisfaction of spatial restraints

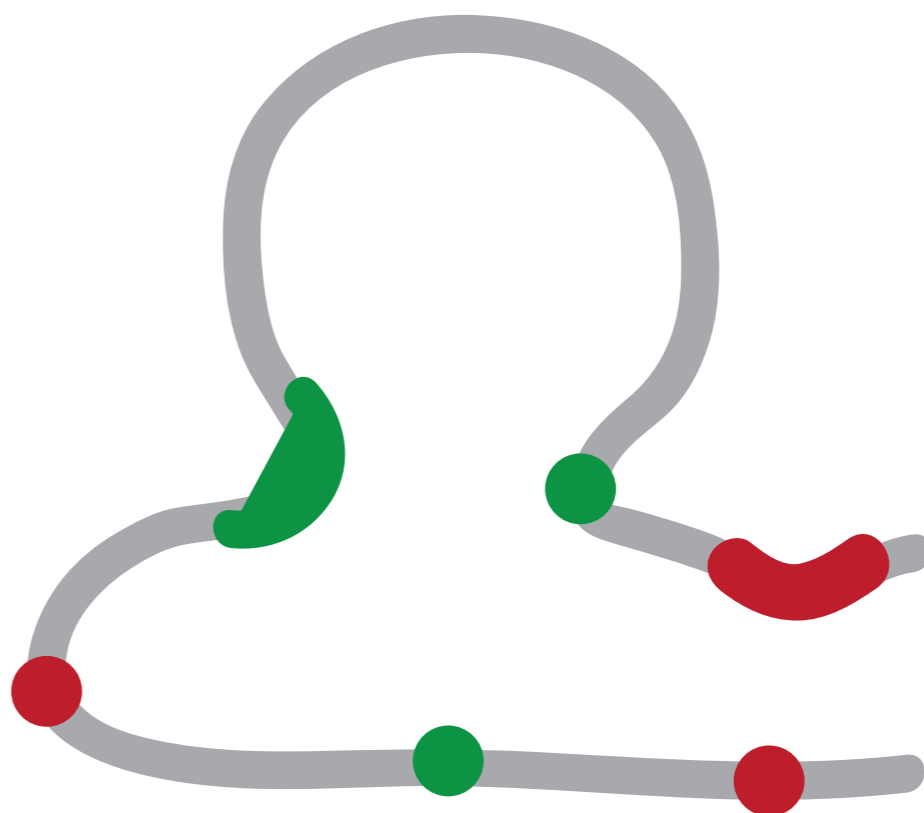
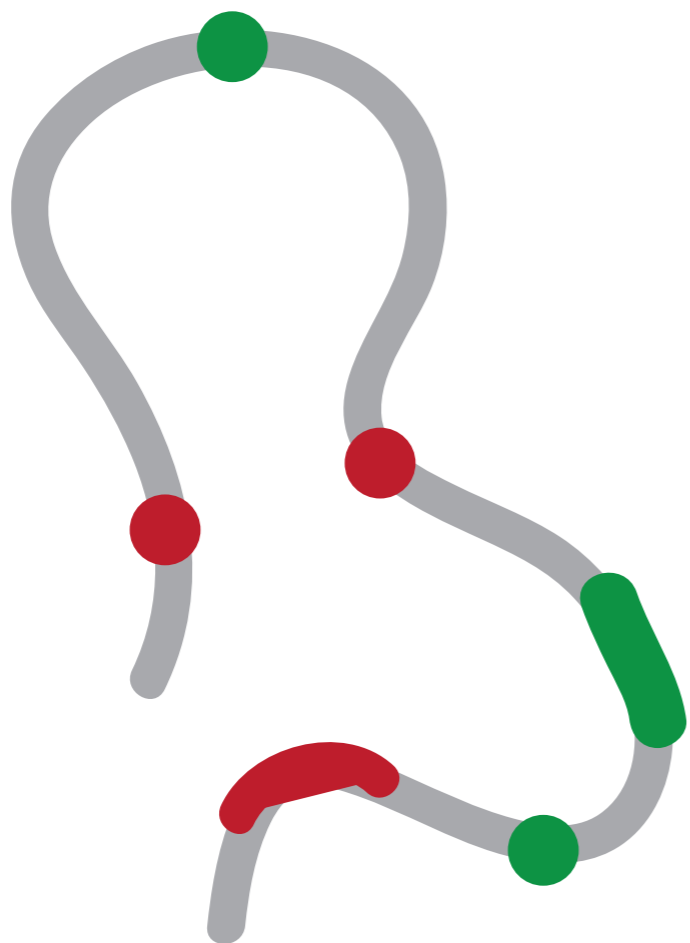
Marc A. Martí-Renom

Structural Genomics Group (ICREA, CNAG-CRG)

<http://marciuslab.org>
<http://3DGenomes.org>
<http://cnag.crg.eu>

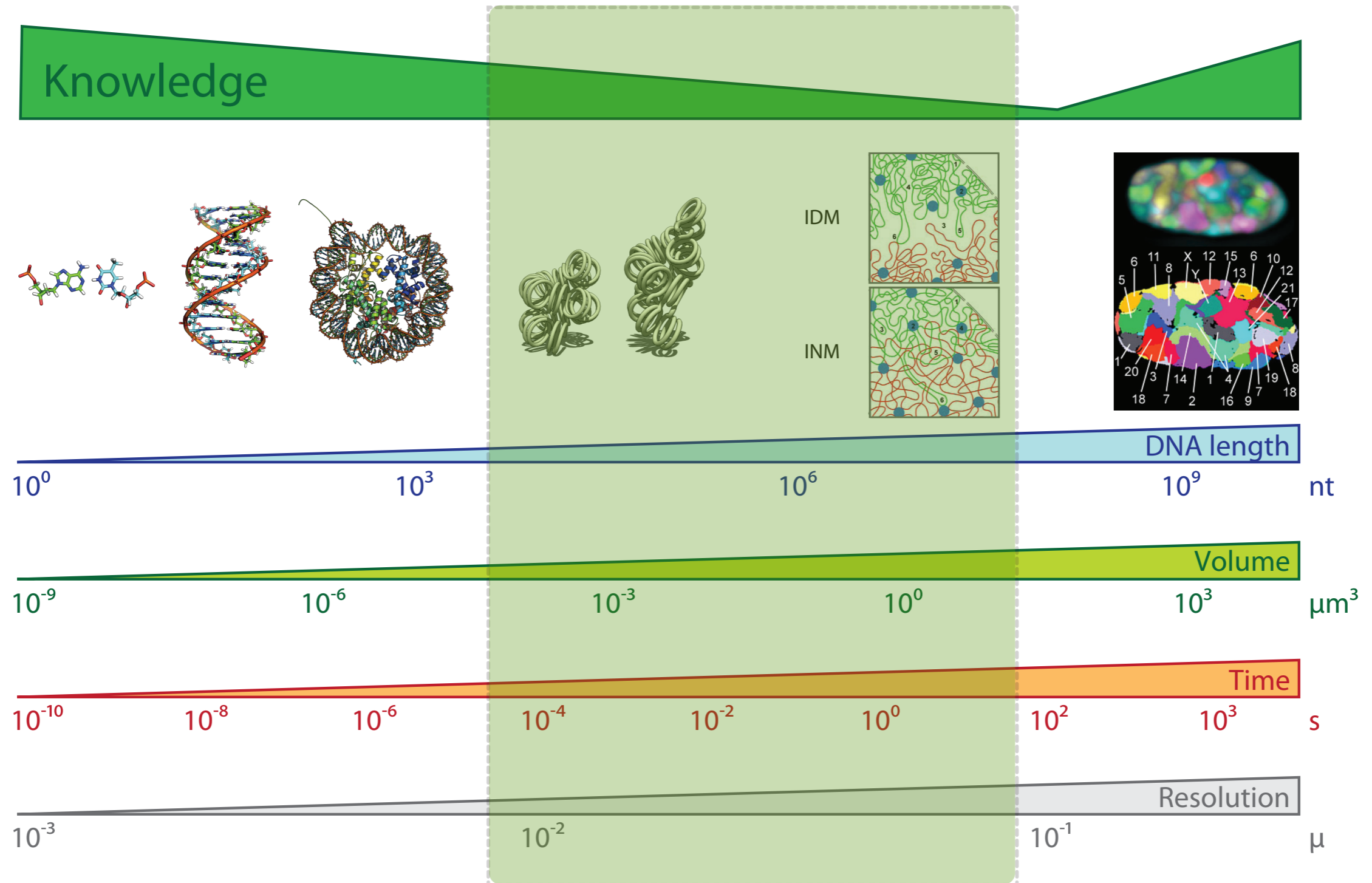
cnag CRG[®]  ICREA





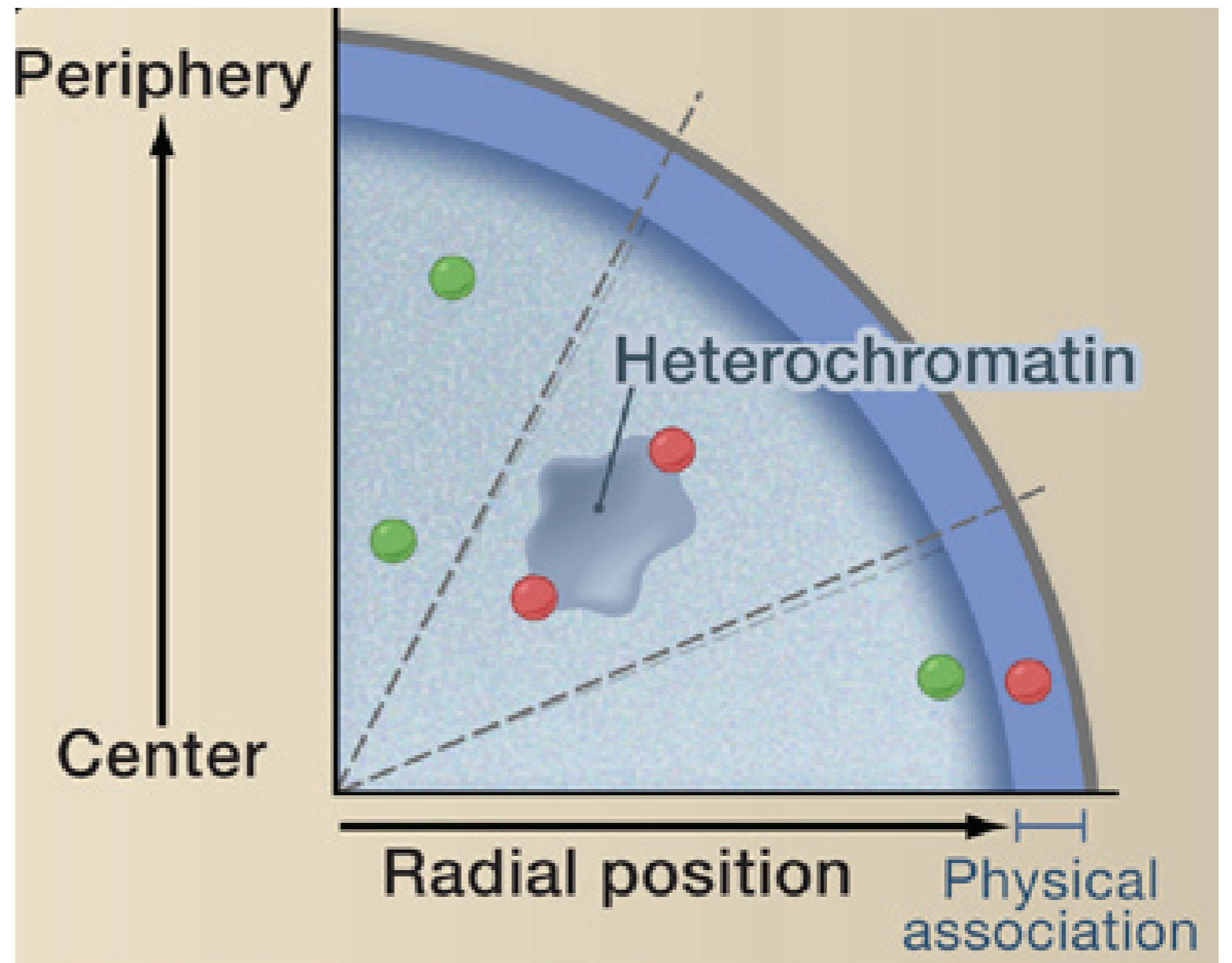
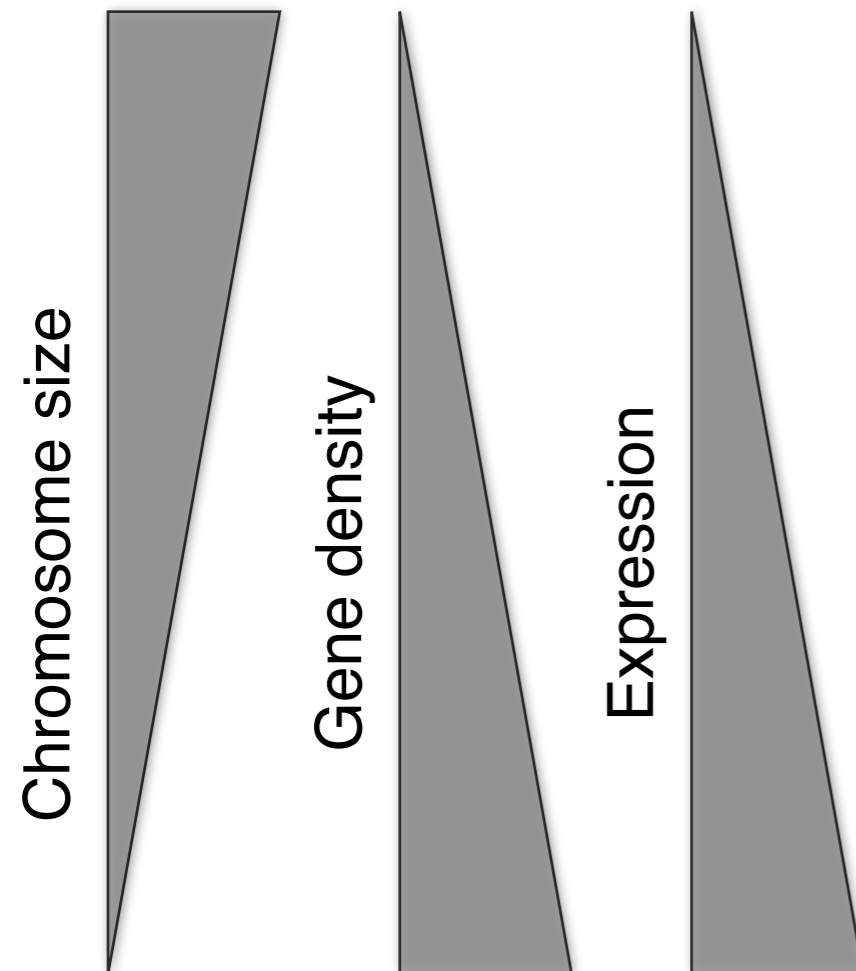
Resolution Gap

Marti-Renom, M. A. & Mirny, L. A. PLoS Comput Biol 7, e1002125 (2011)



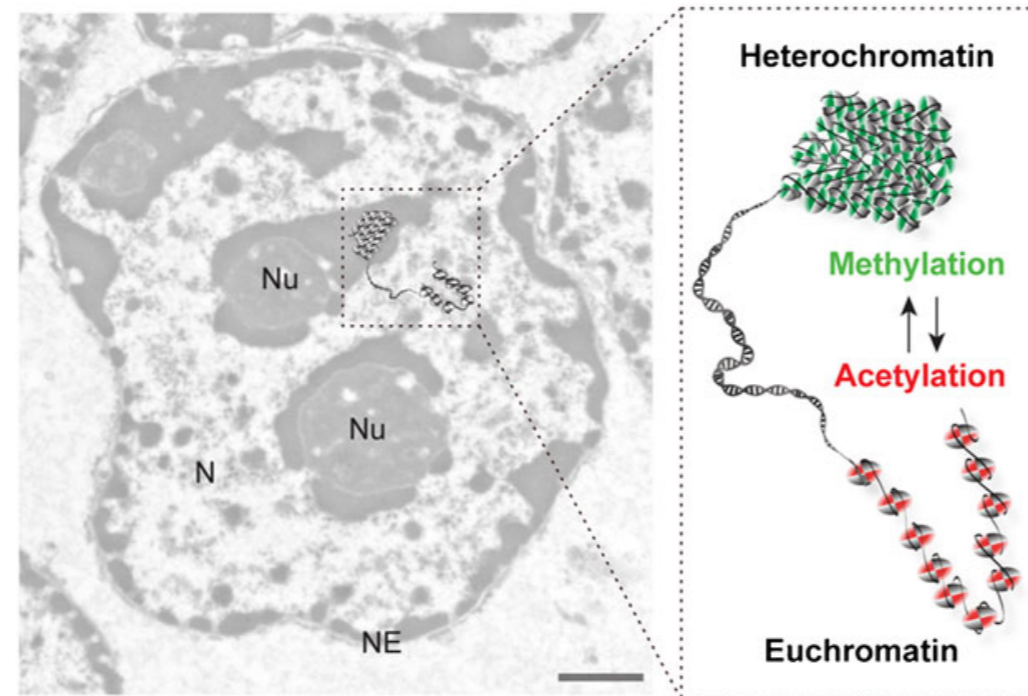
Level I: Radial genome organization

Takizawa, T., Meaburn, K. J. & Misteli, T. The meaning of gene positioning. *Cell* 135, 9–13 (2008).



Level II: Euchromatin vs heterochromatin

Electron microscopy



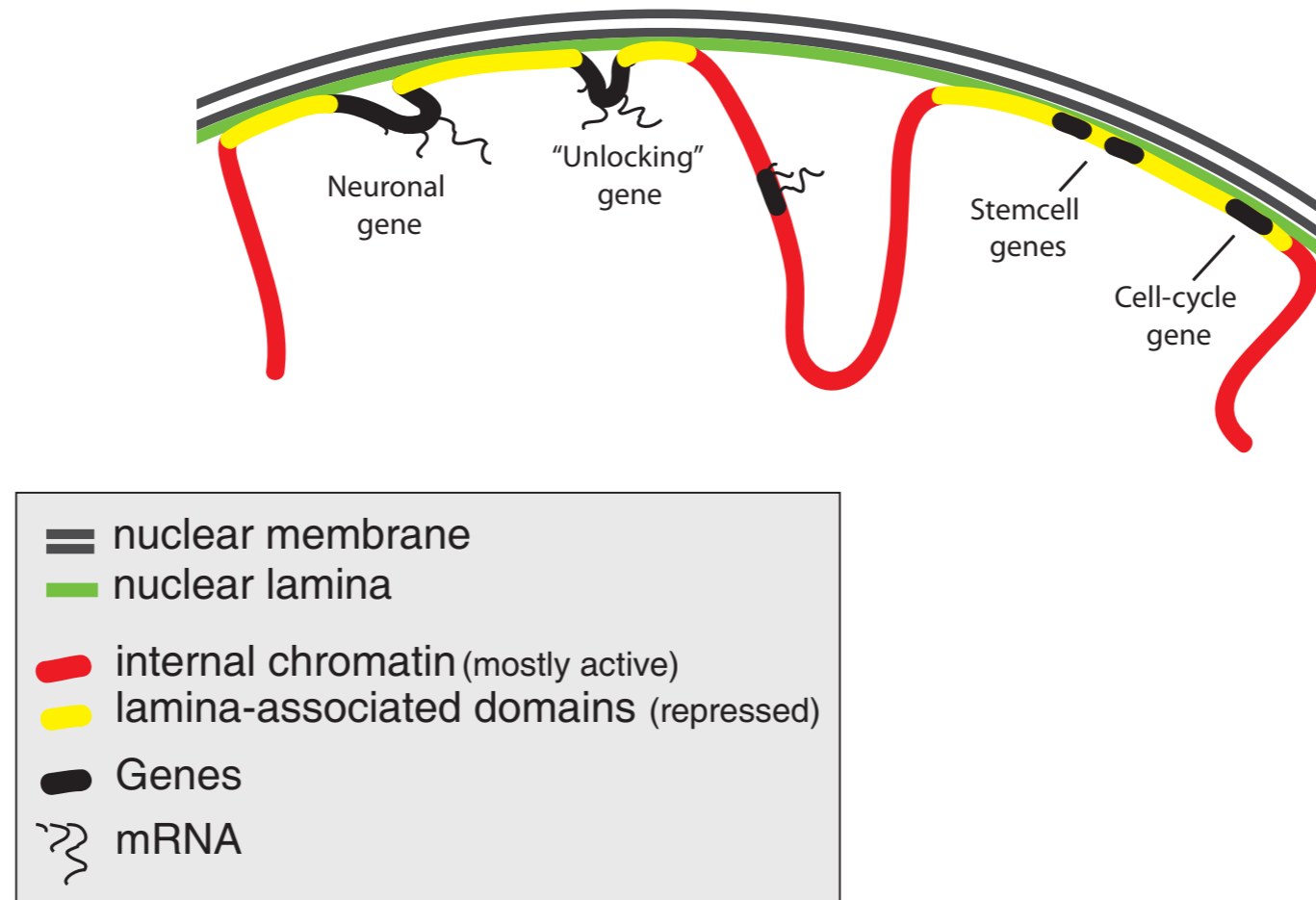
Euchromatin:

chromatin that is located away from the nuclear lamina, is generally less densely packed, and contains actively transcribed genes

Heterochromatin:

chromatin that is near the nuclear lamina, tightly condensed, and transcriptionally silent

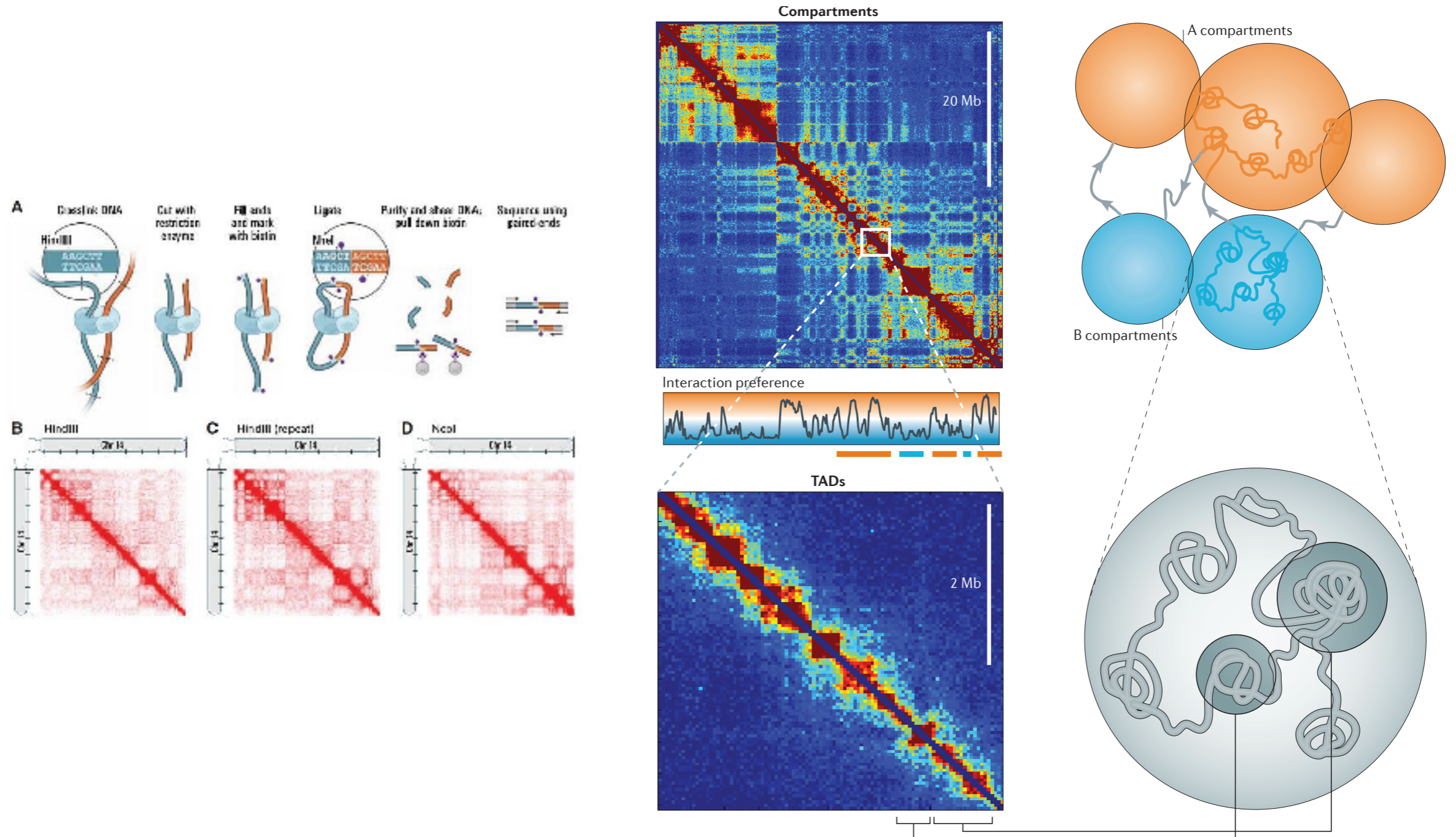
Level III: Lamina-genome interactions



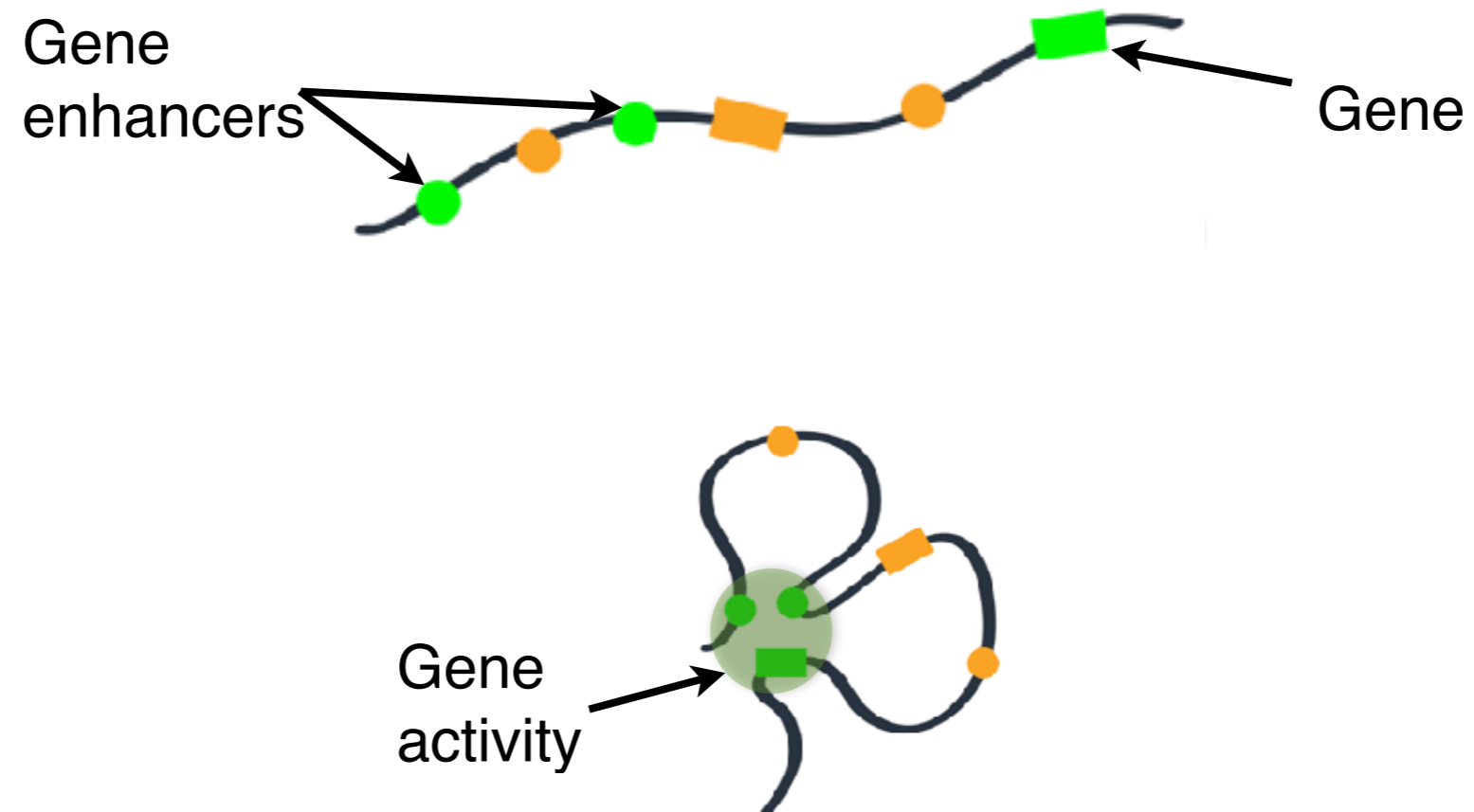
Most genes in Lamina Associated Domains are transcriptionally silent, suggesting that **lamina-genome interactions** are widely involved in the control of **gene expression**

Level IV: Higher-order organization

Dekker, J., Marti-Renom, M. A. & Mirny, L. A. Nat Rev Genet 14, 390–403 (2013).



Level V: Chromatin loops



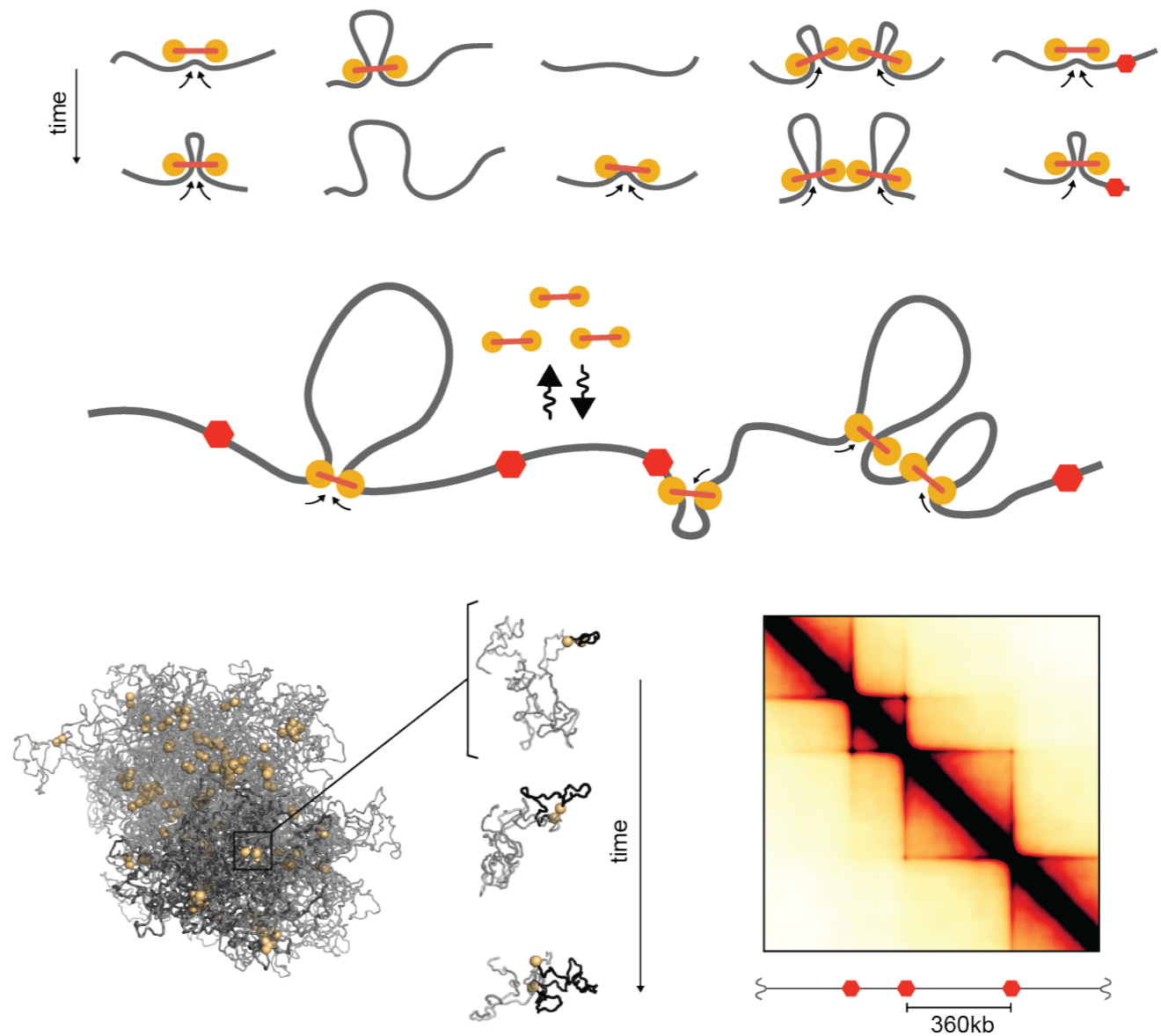
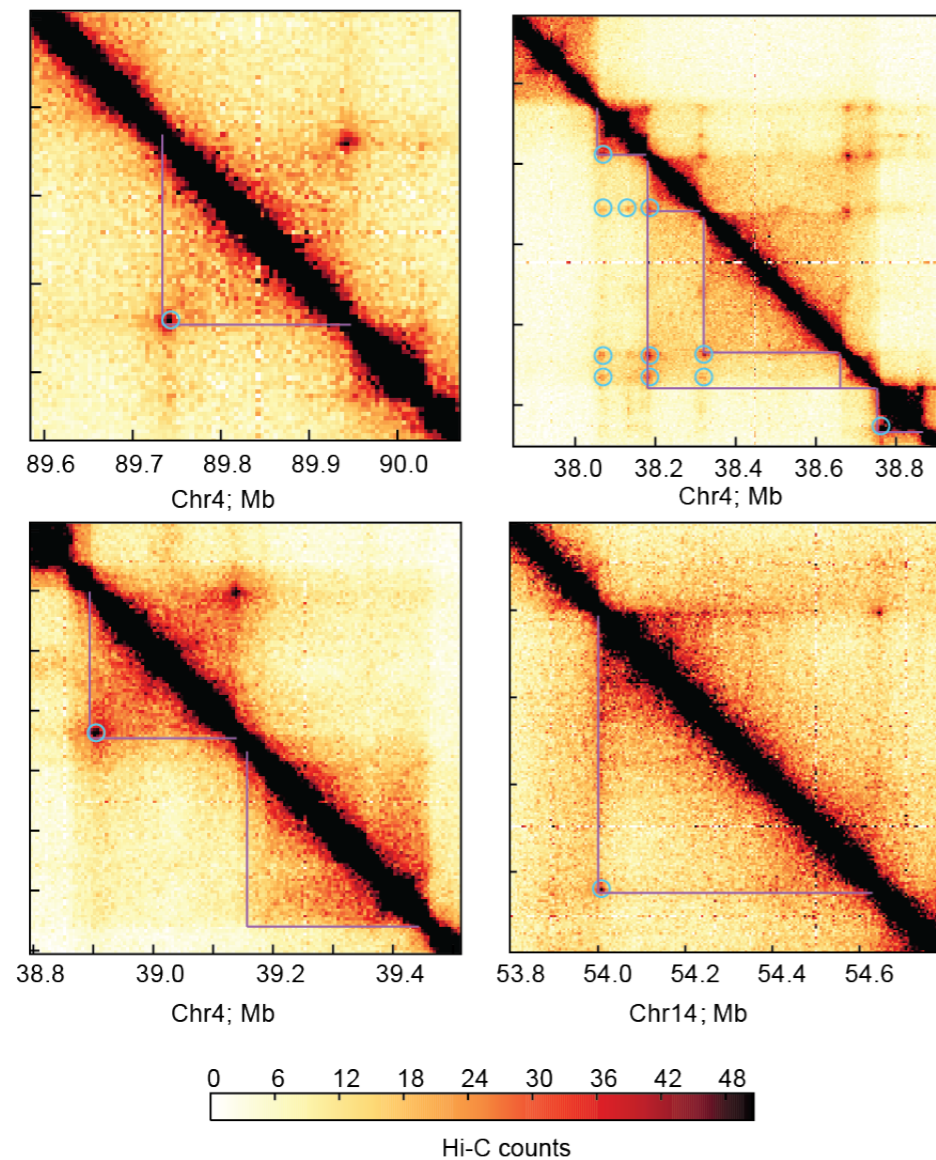
Loops bring distal genomic regions in close proximity to one another

This in turn can have profound effects on gene transcription

Enhancers can be thousands of kilobases away from their target genes in any direction (or even on a separate chromosome)

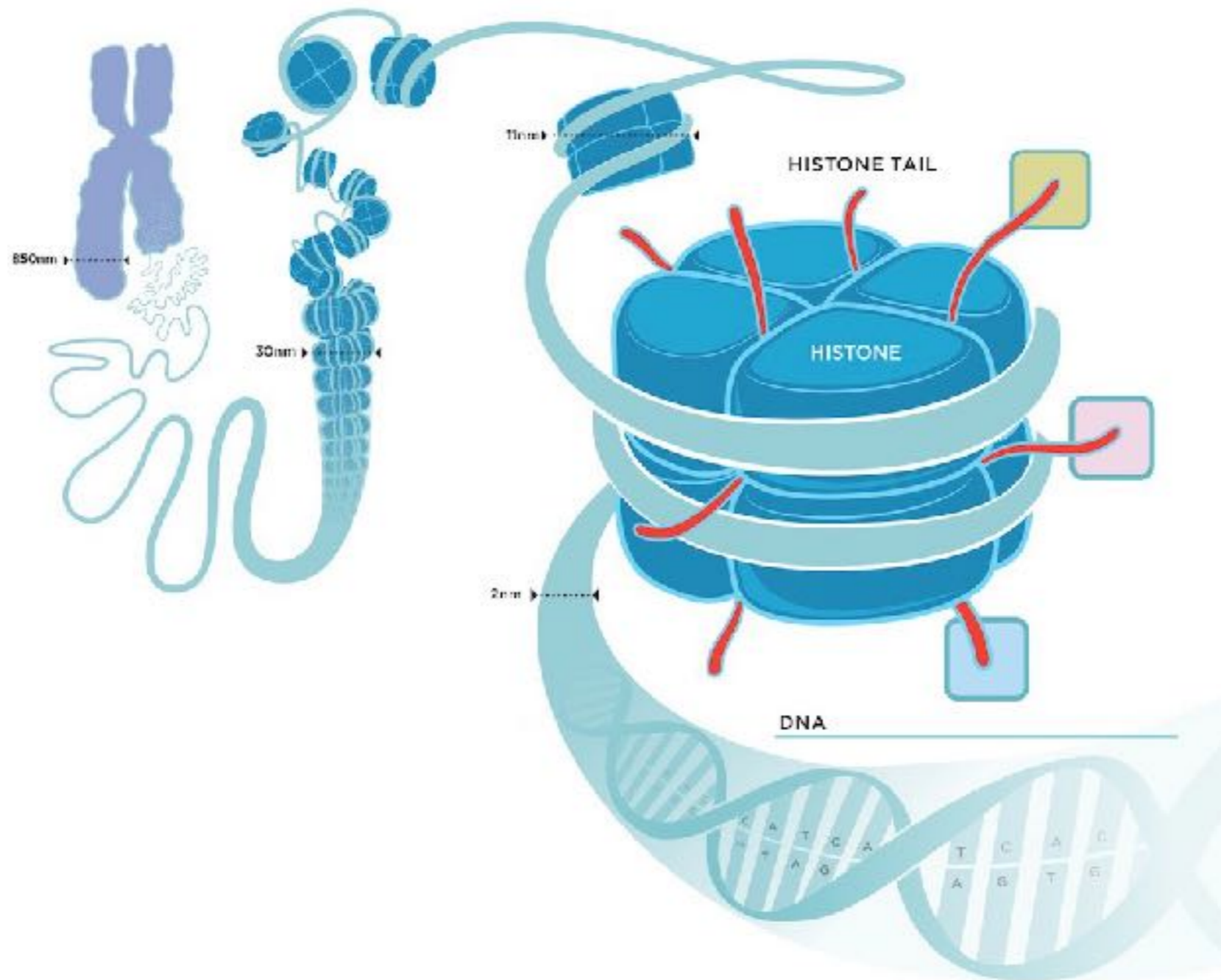
Level V: Loop-extrusion as a driving force

Fudenberg, G., Imakaev, M., Lu, C., Goloborodko, A., Abdennur, N., & Mirny, L. A. (2015).
Formation of Chromosomal Domains by Loop Extrusion. bioRxiv.



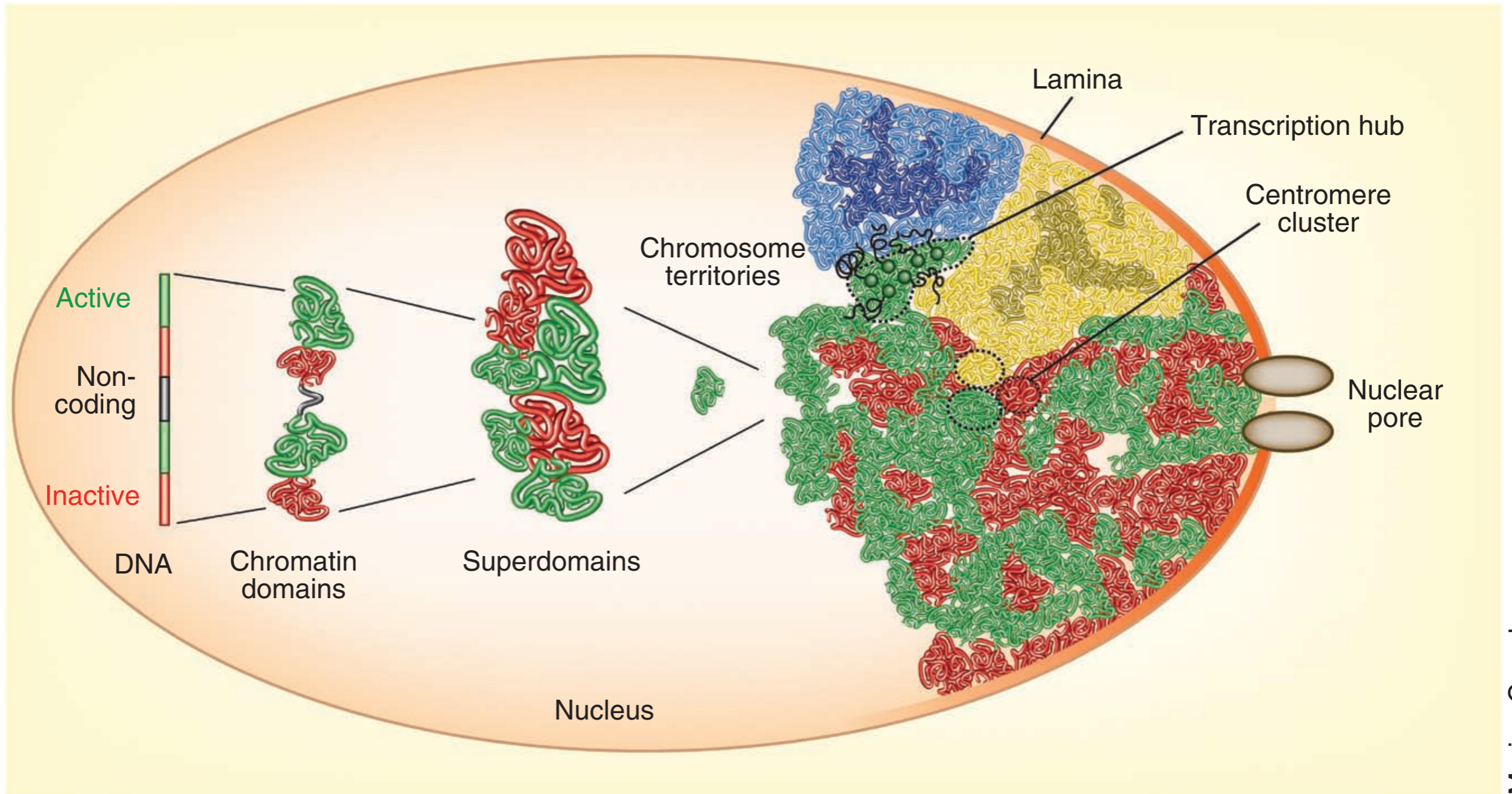
Level VI: Nucleosome

Chromosome **Chromatin fibre** **Nucleosome**



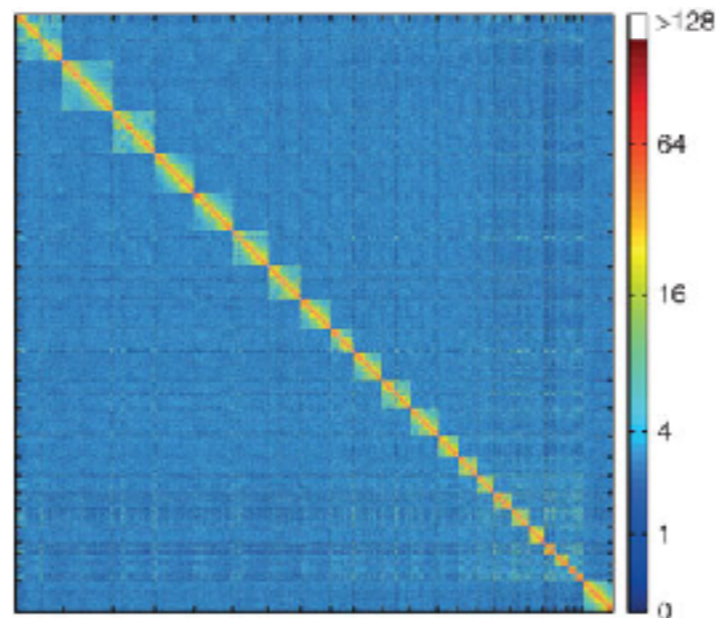
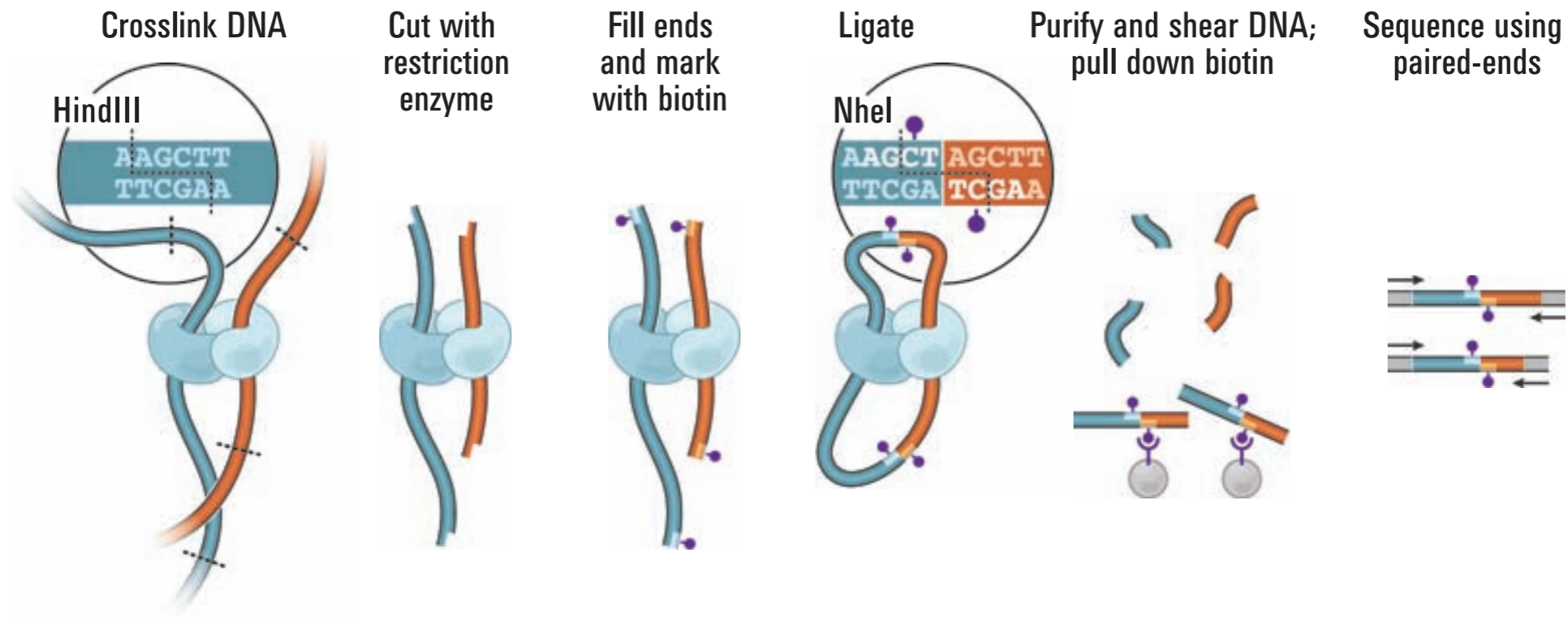
Complex genome organization

Cavalli, G. & Misteli, T. Functional implications of genome topology. *Nat Struct Mol Biol* 20, 290–299 (2013).

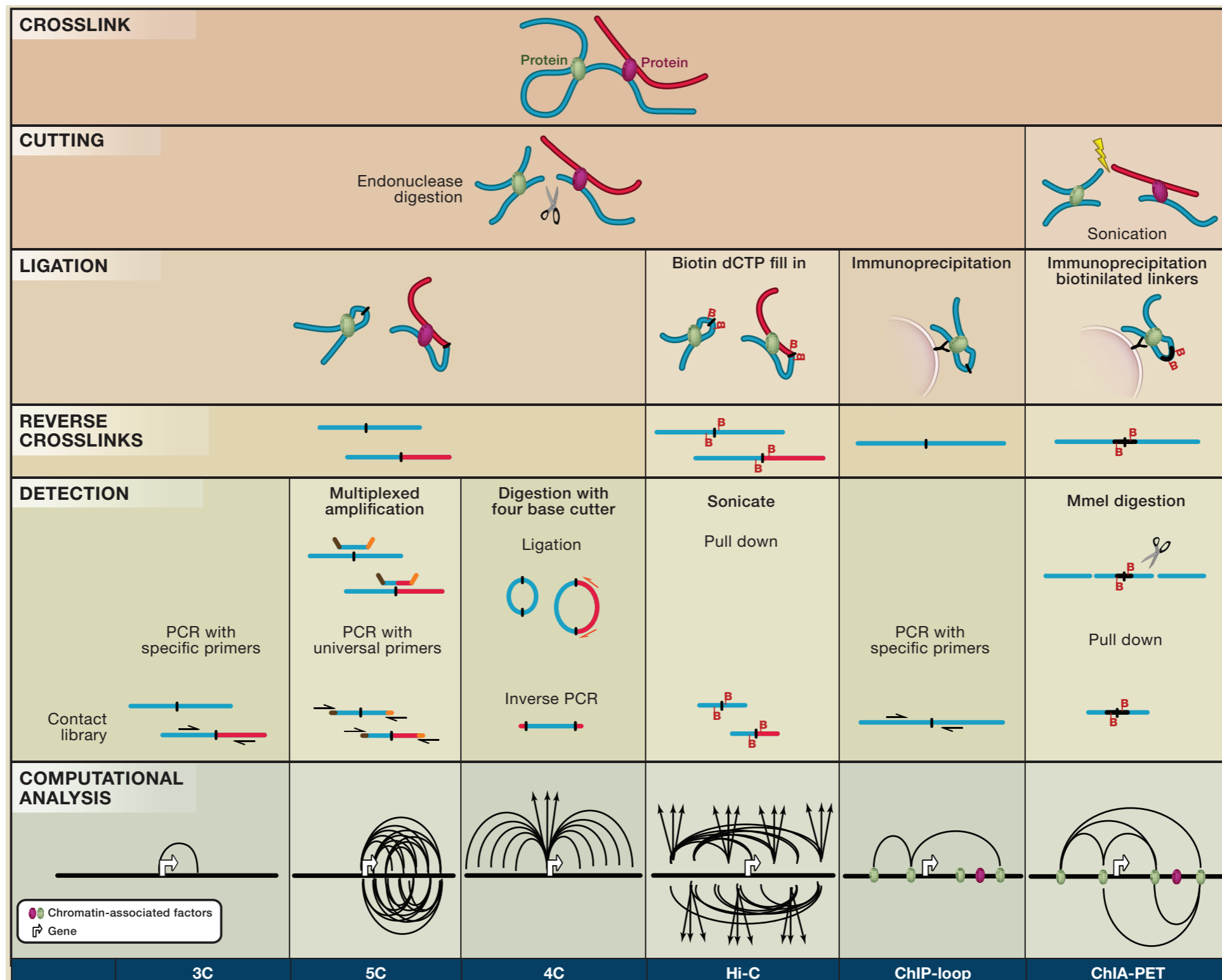


Chromosome Conformation Capture

Dekker, J., Rippe, K., Dekker, M., & Kleckner, N. (2002). *Science*, 295(5558), 1306–1311.
Lieberman-Aiden, E., et al. (2009). *Science*, 326(5950), 289–293.

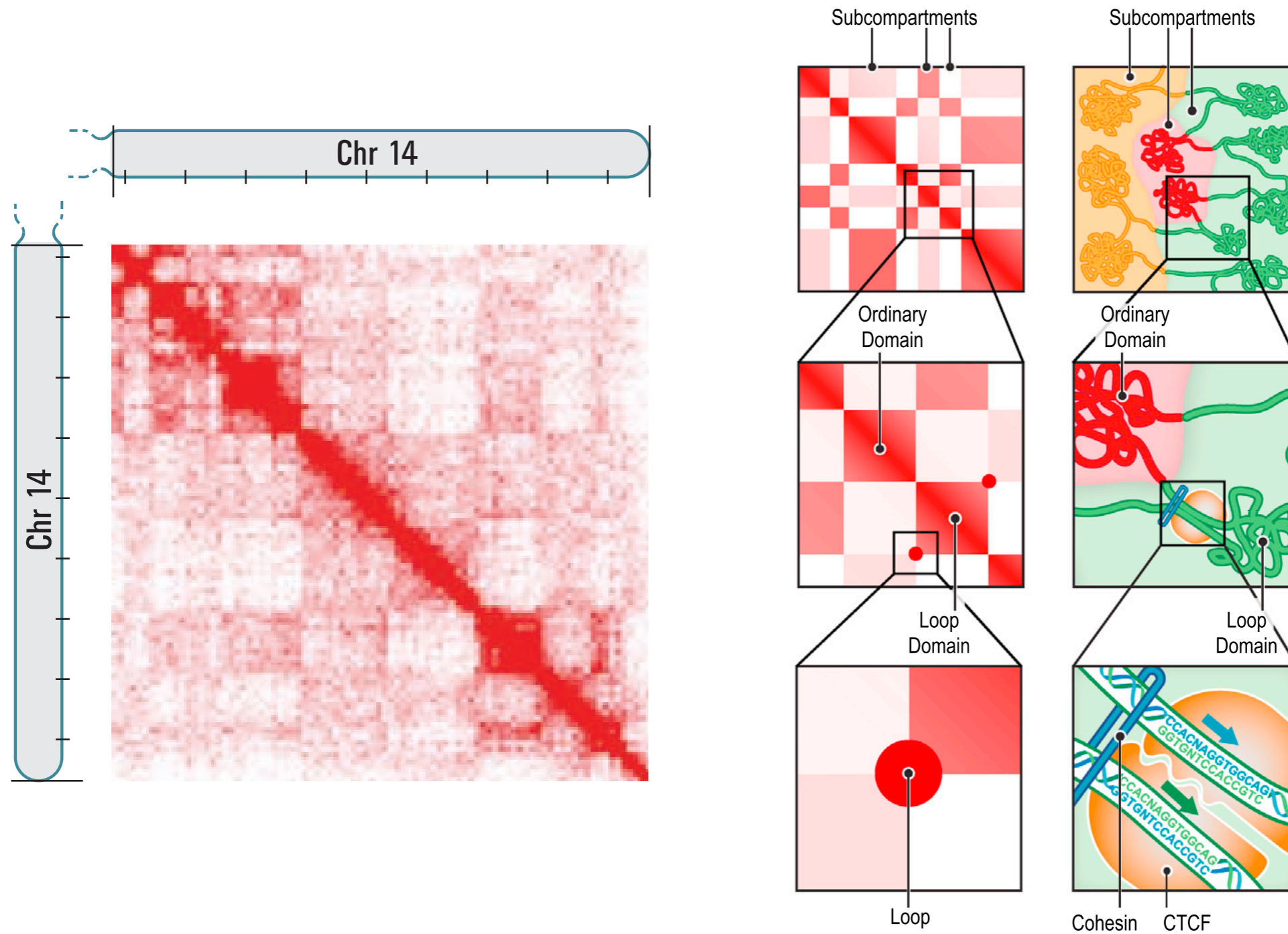


Chromosome Conformation Capture



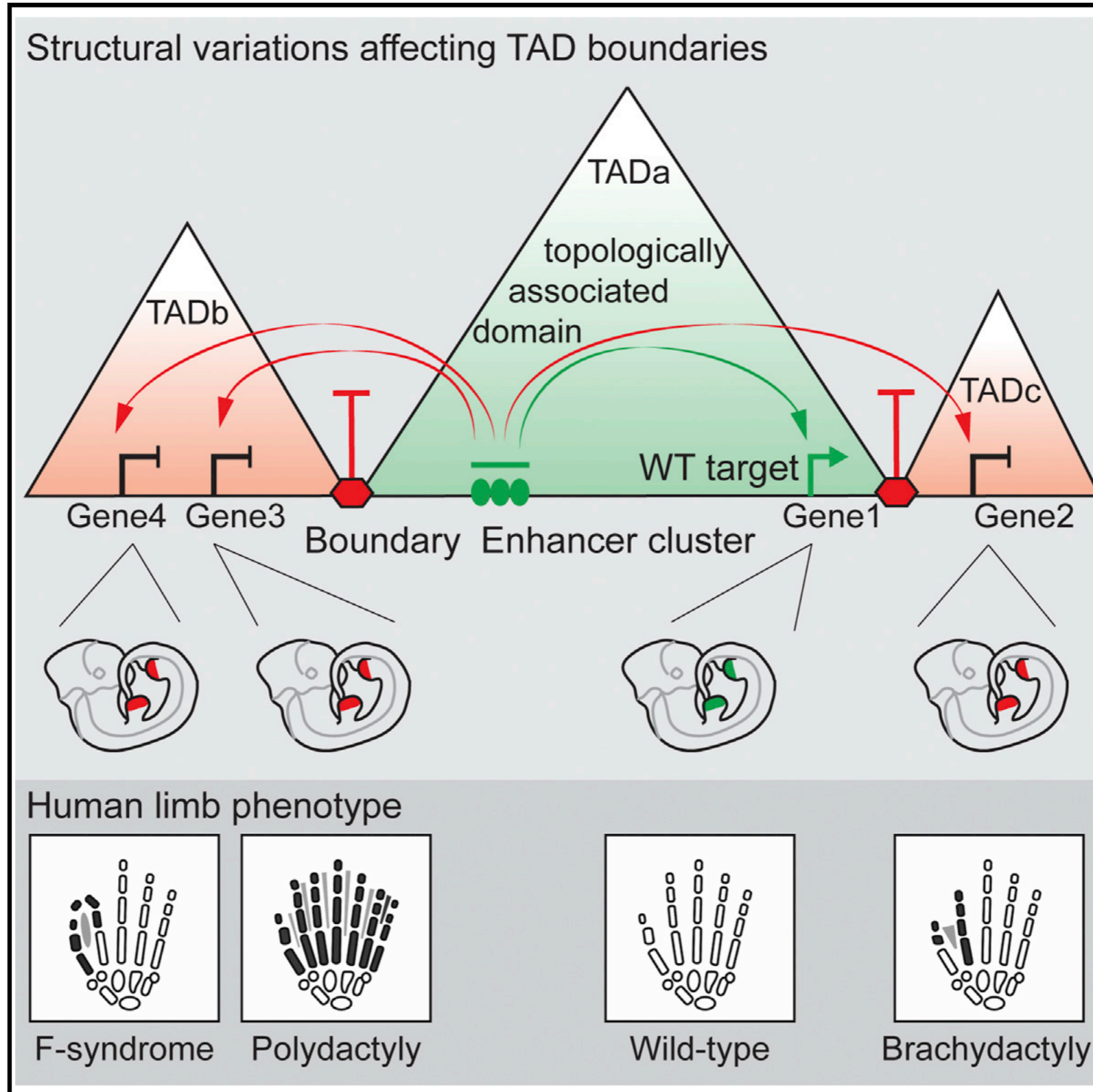
Hakim, O., & Misteli, T. (2012). SnapShot: Chromosome Confirmation Capture. Cell, 148(5), 1068–1068.e2.

Hierarchical genome organisation



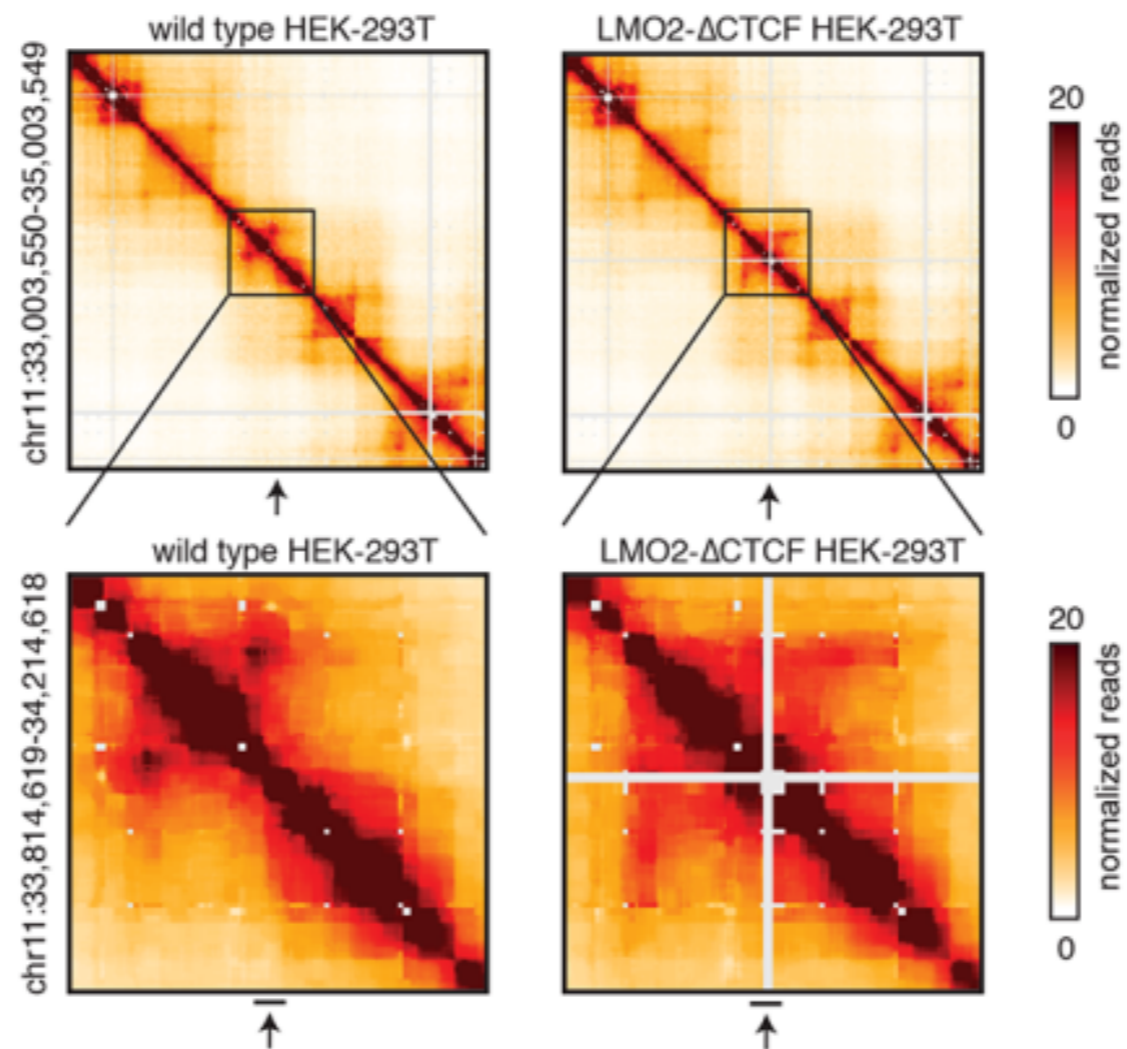
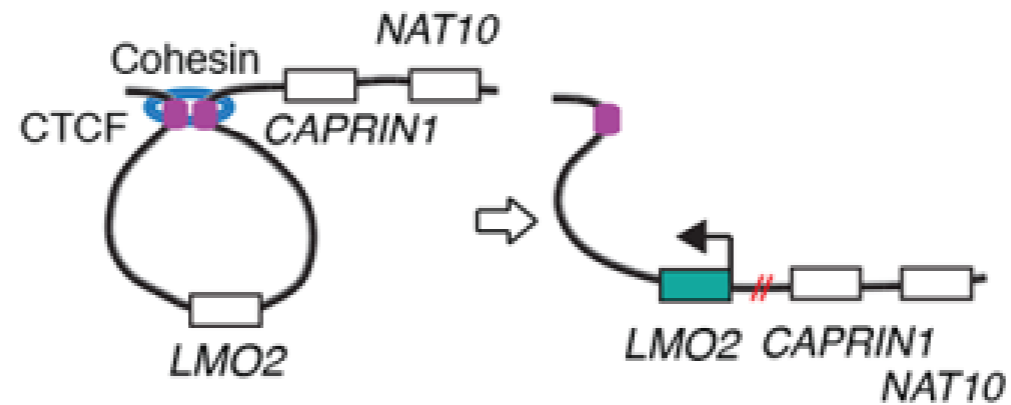
TADs are functional units

Lupiáñez, et al. (2015). Cell, 1–15.



TADs are functional units

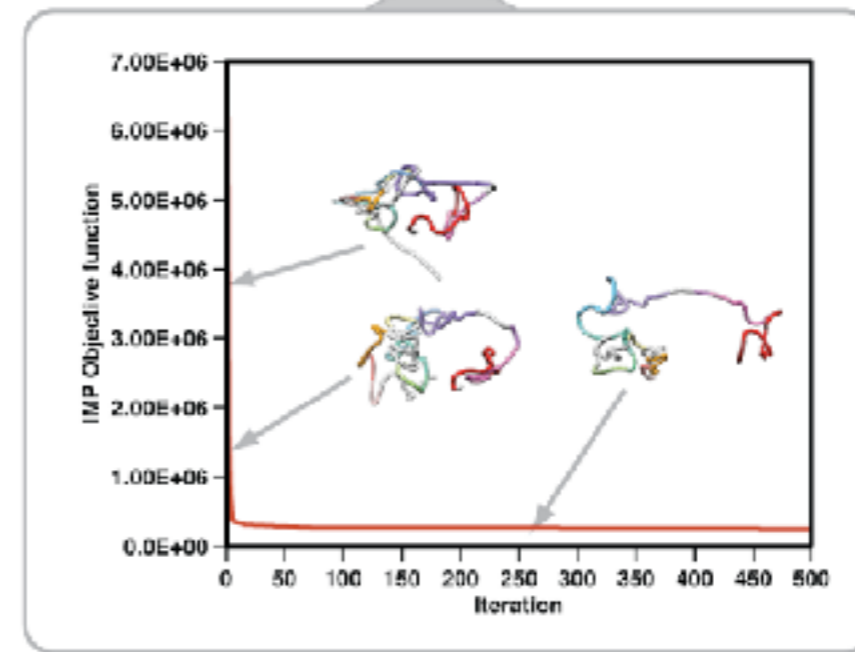
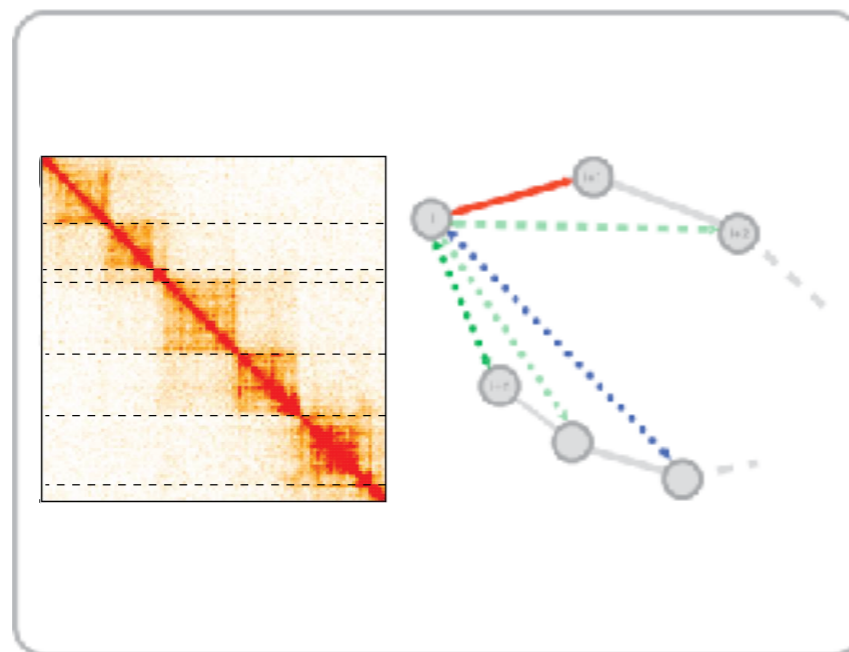
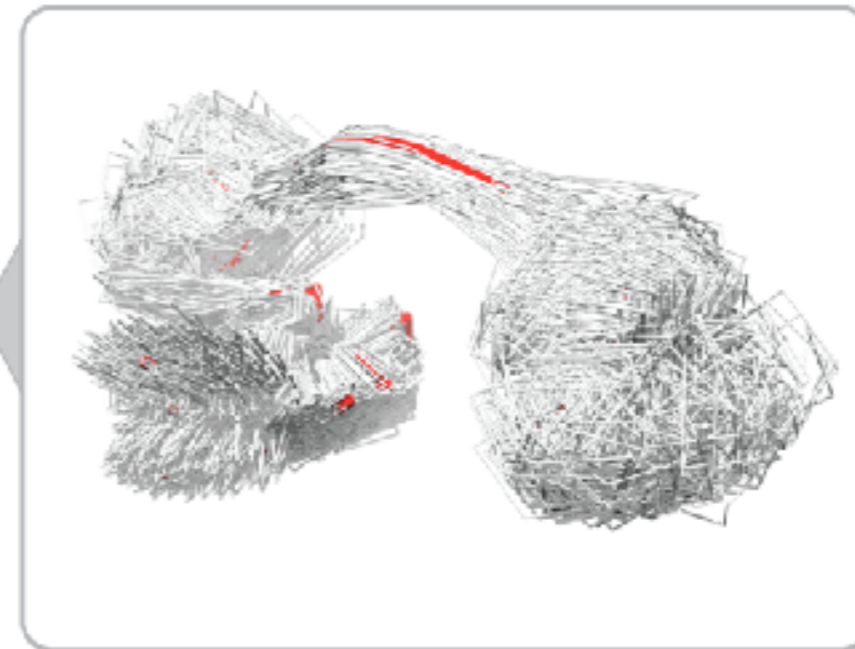
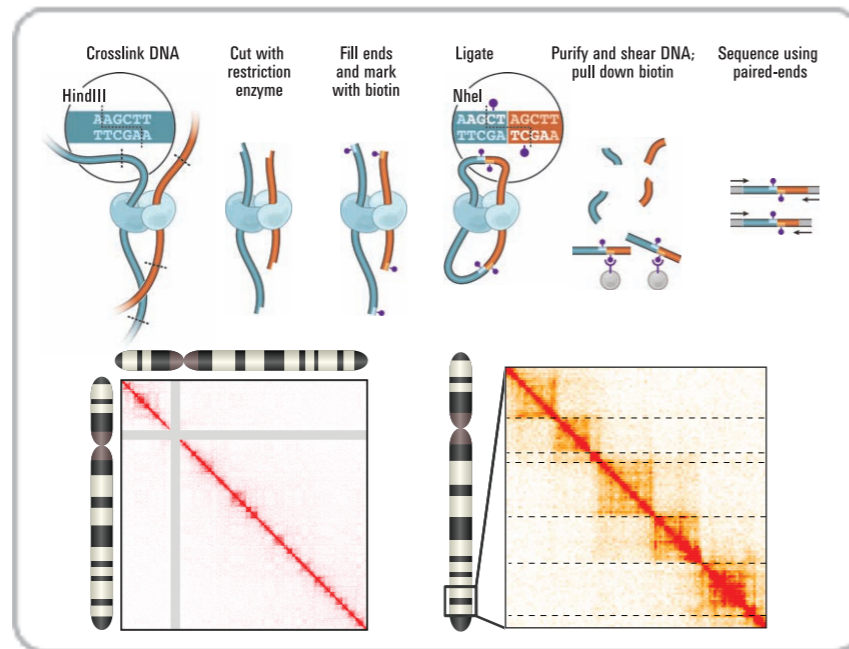
Hnisz, D., et al. (2016). Science, on line



Hybrid Method

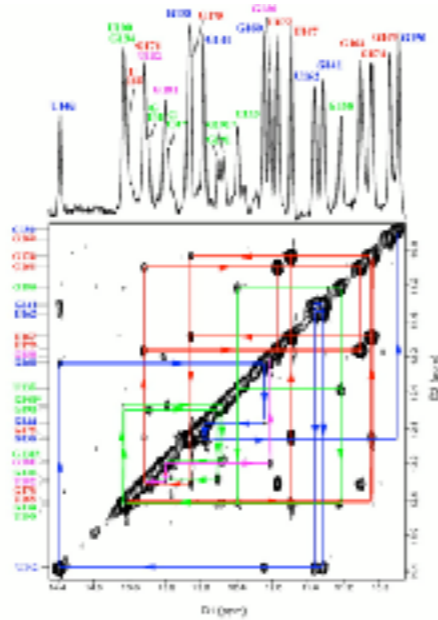
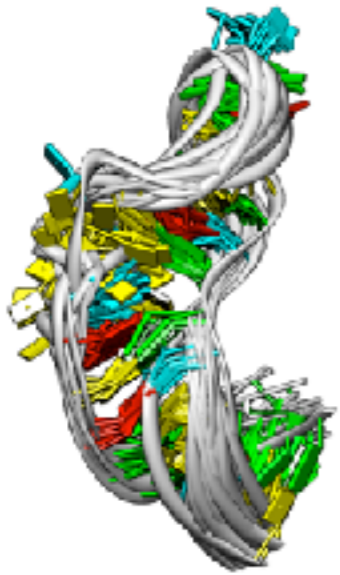
Baù, D. & Marti-Renom, M. A. *Methods* 58, 300–306 (2012).

Experiments

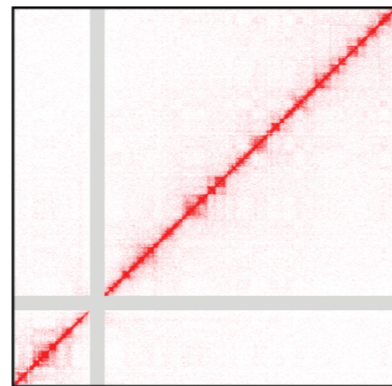
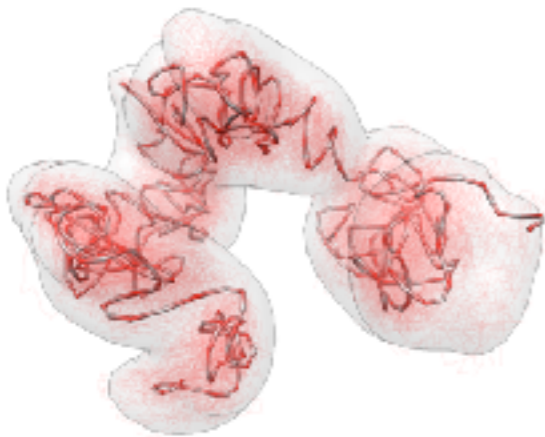


Computation

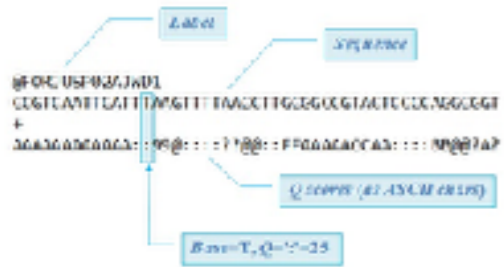
Structure determination using Hi-C data



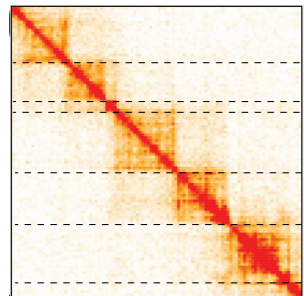
Biomolecular structure determination
2D-NOESY data



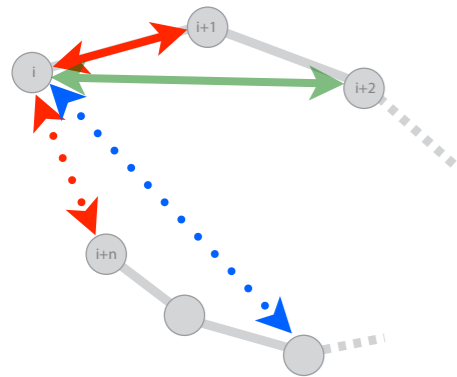
Chromosome structure determination
3C-based data



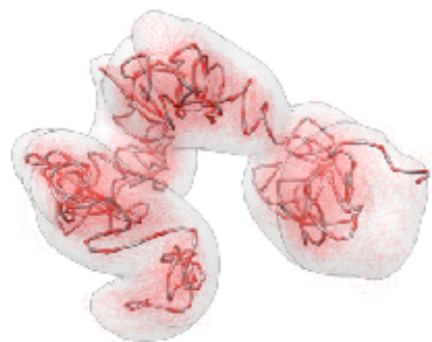
FastQ files to Maps



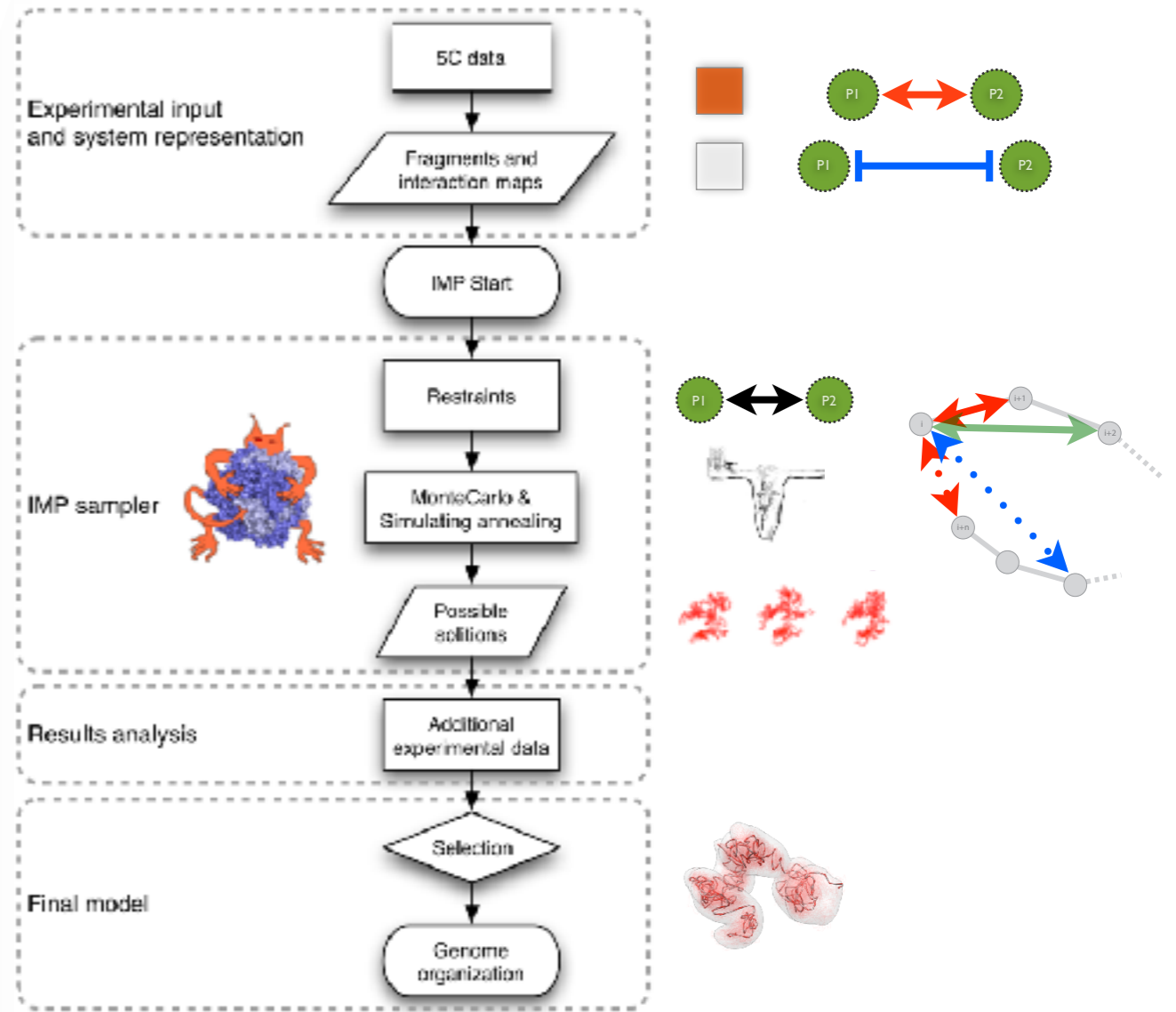
Map analysis



Model building



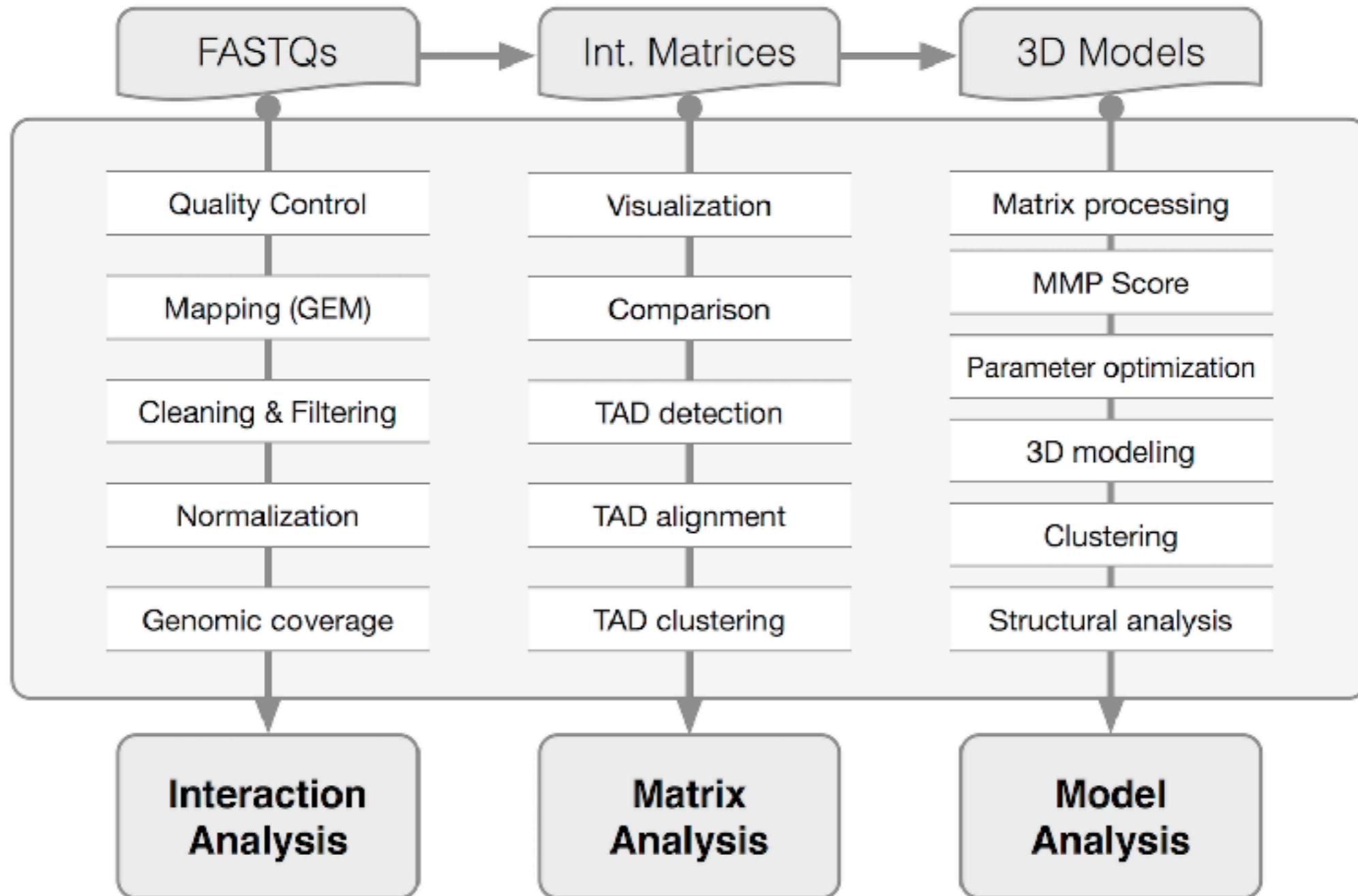
Model analysis





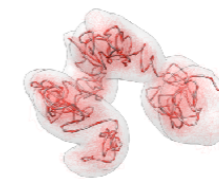
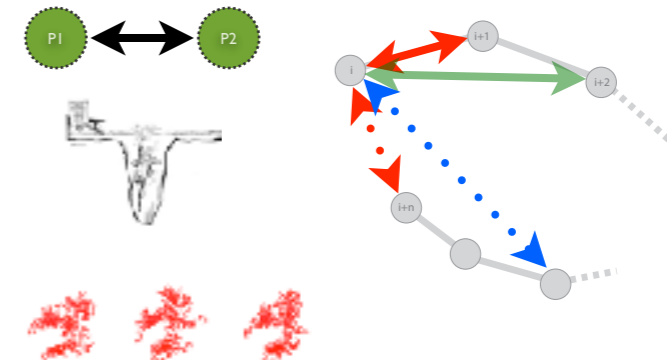
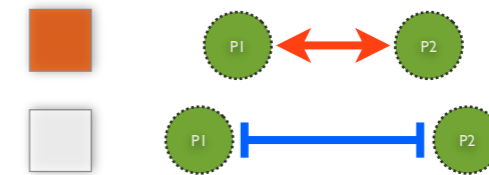
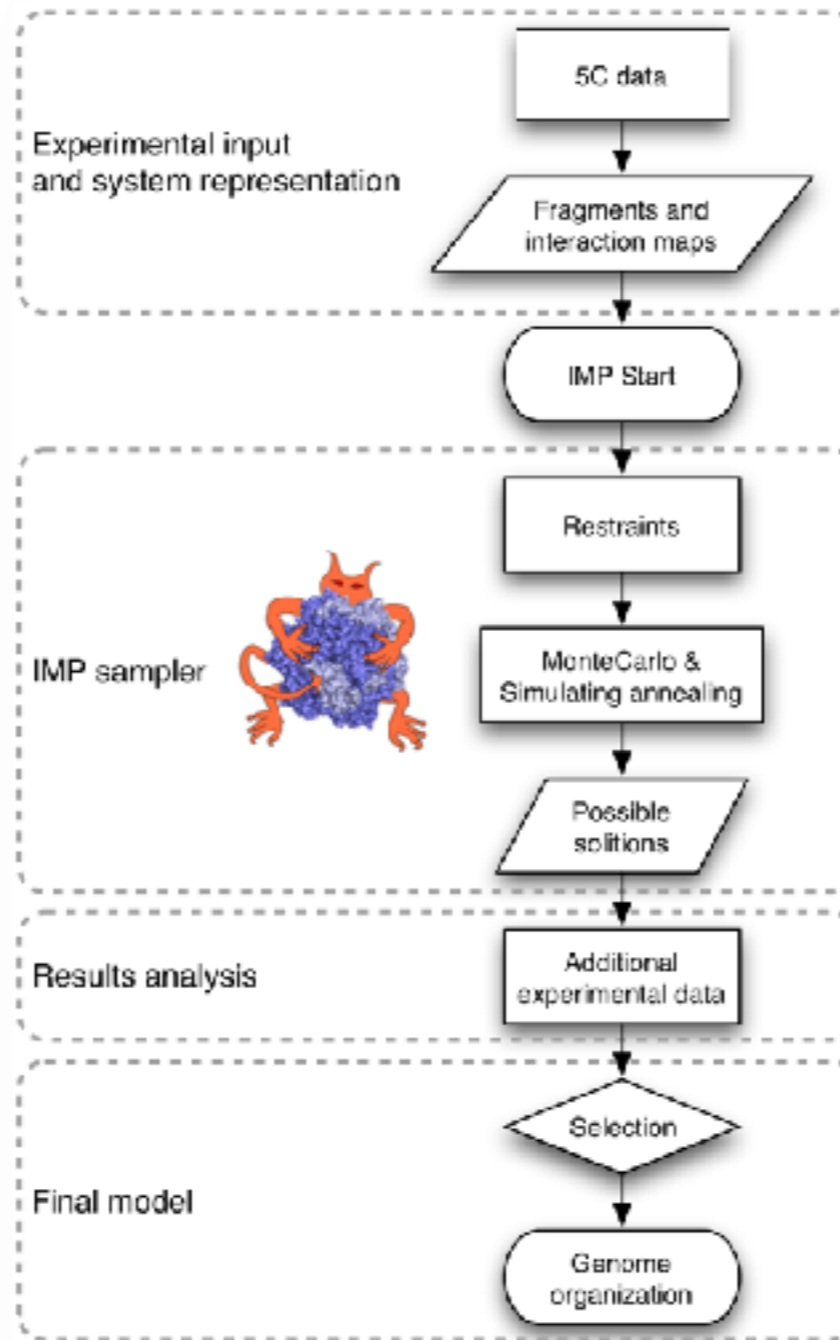
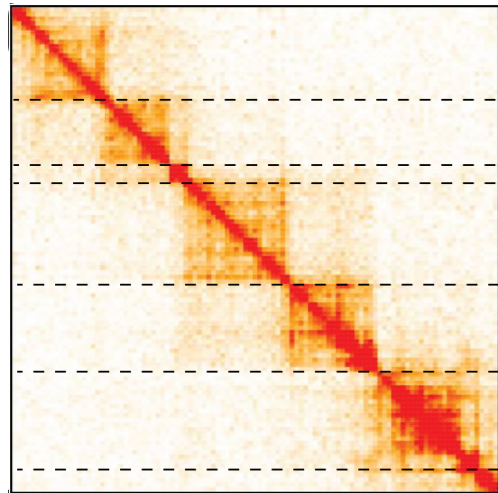
TADbit

Serra, Baù, et al. (2017). PLOS CompBio

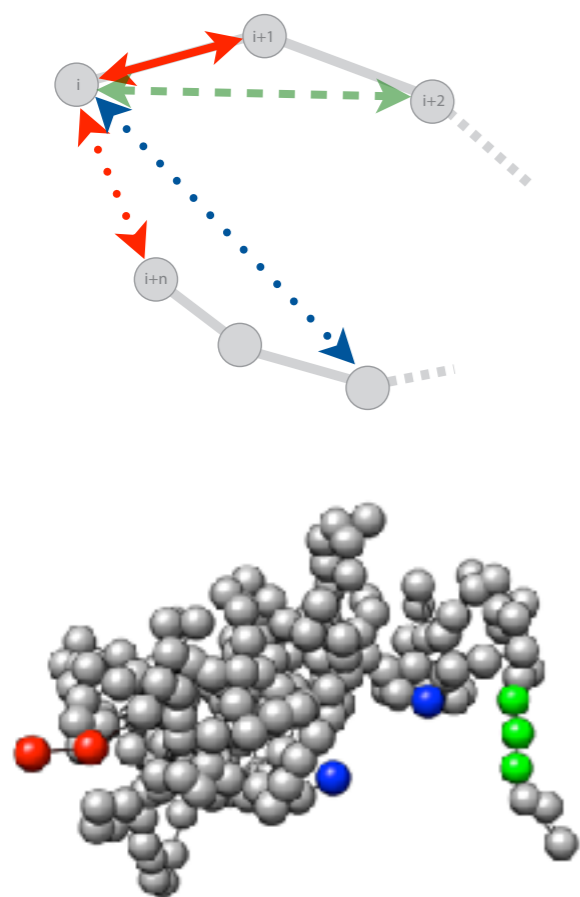


TADbit

<http://3DGenomes.org>
<http://www.integrativemodeling.org>



Model representation and scoring

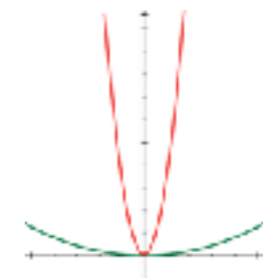


$d = d_0$



Harmonic

$$H_{i,j} = k(d_{i,j} - d_{i,j}^0)^2$$

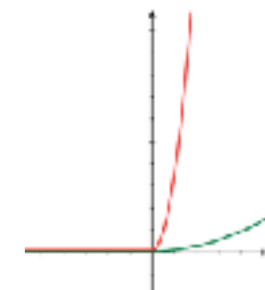


$d < d_0$



Harmonic Upper Bound

$$\begin{cases} \text{if } d_{i,j} \geq d_{i,j}^0; & ubH_{i,j} = k(d_{i,j} - d_{i,j}^0)^2 \\ \text{if } d_{i,j} < d_{i,j}^0; & ubH_{i,j} = 0 \end{cases}$$

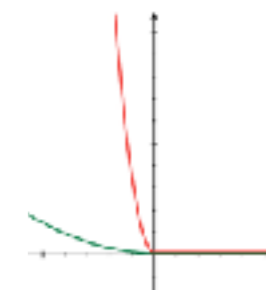


$d > d_0$

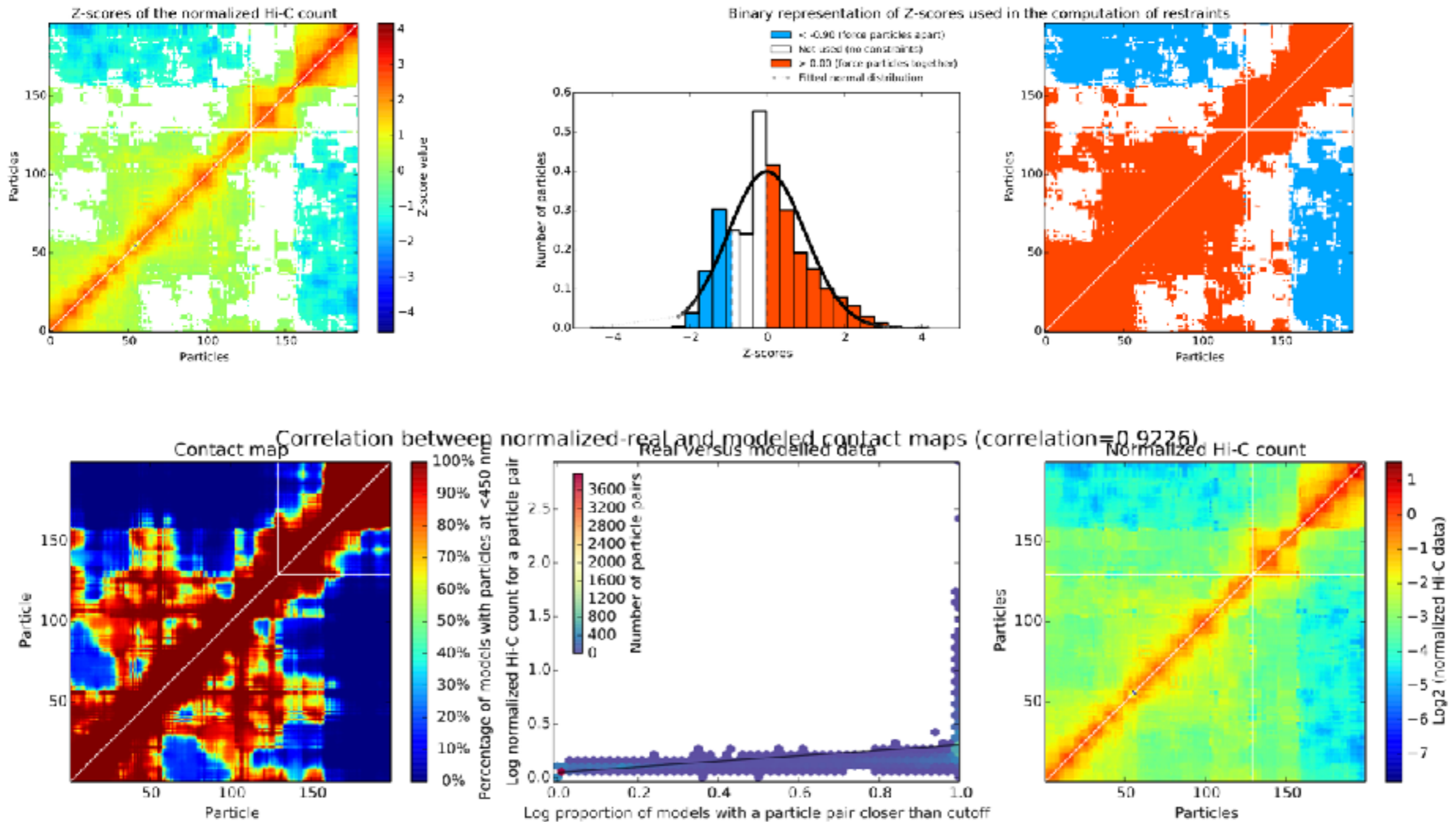


Harmonic Lower Bound

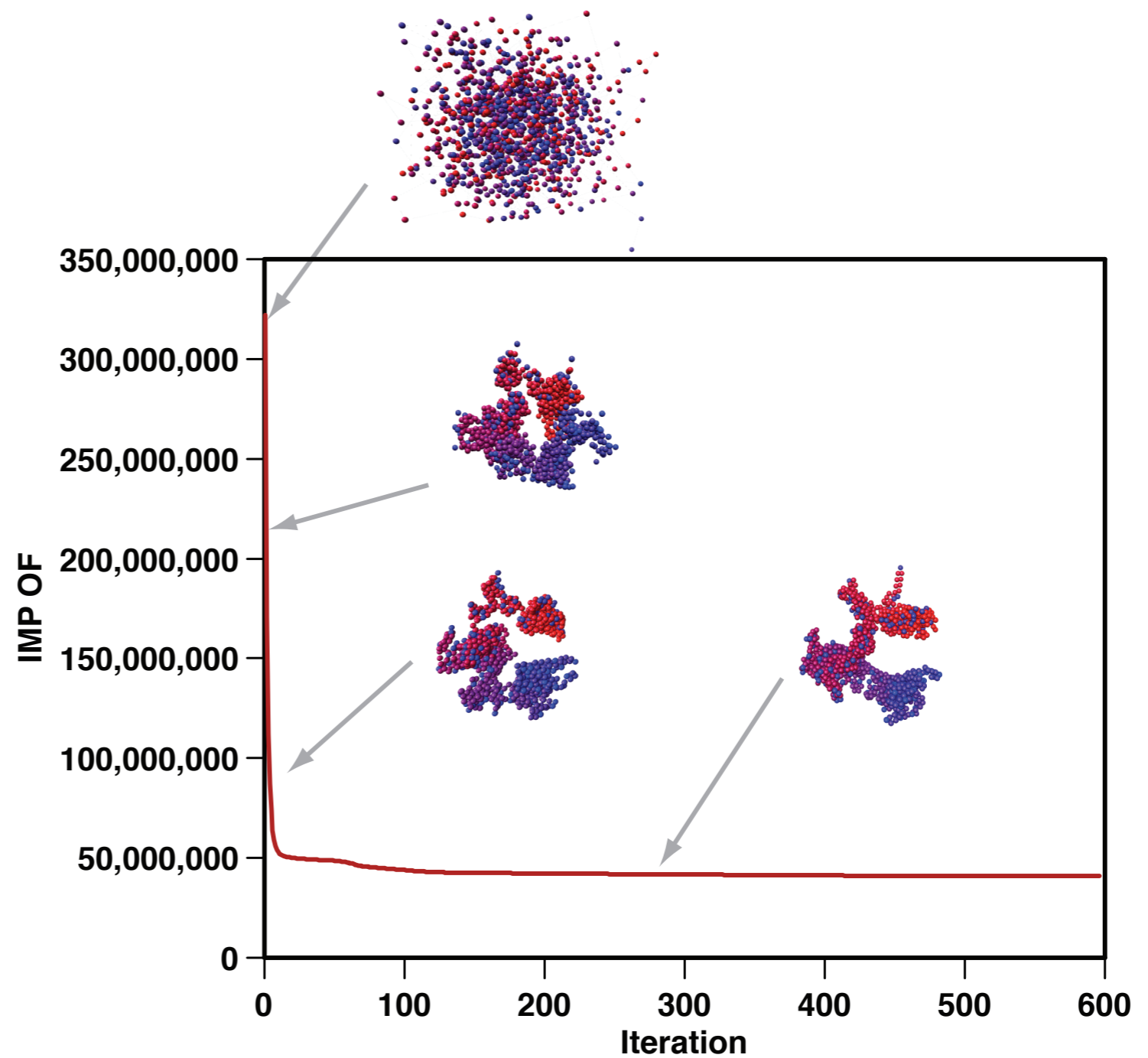
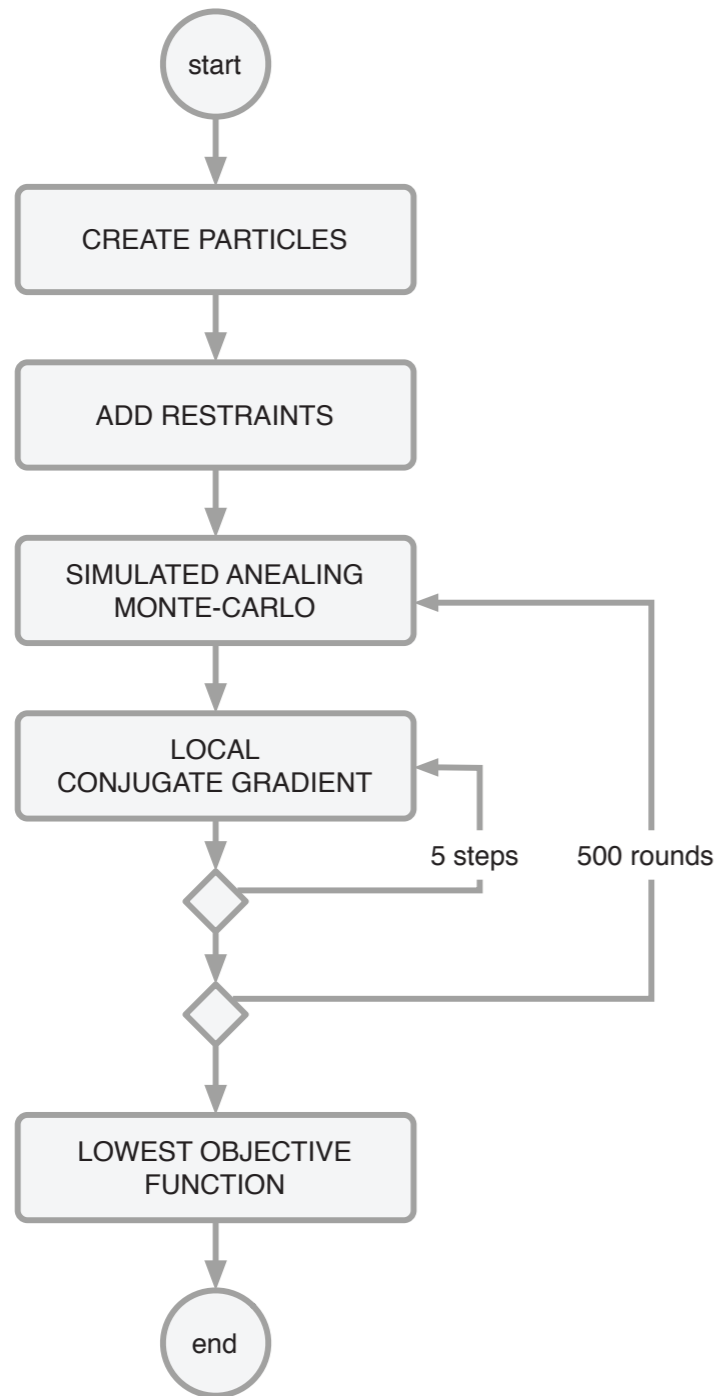
$$\begin{cases} \text{if } d_{i,j} \leq d_{i,j}^0; & lbH_{i,j} = k(d_{i,j} - d_{i,j}^0)^2 \\ \text{if } d_{i,j} > d_{i,j}^0; & lbH_{i,j} = 0 \end{cases}$$



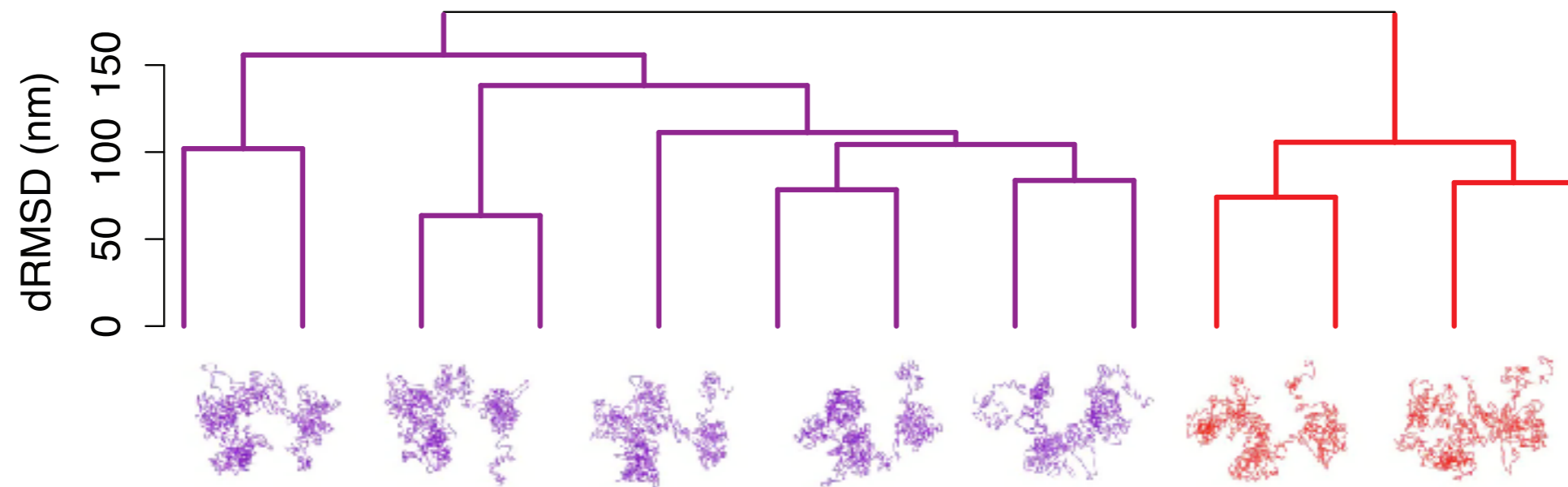
Parameter optimization



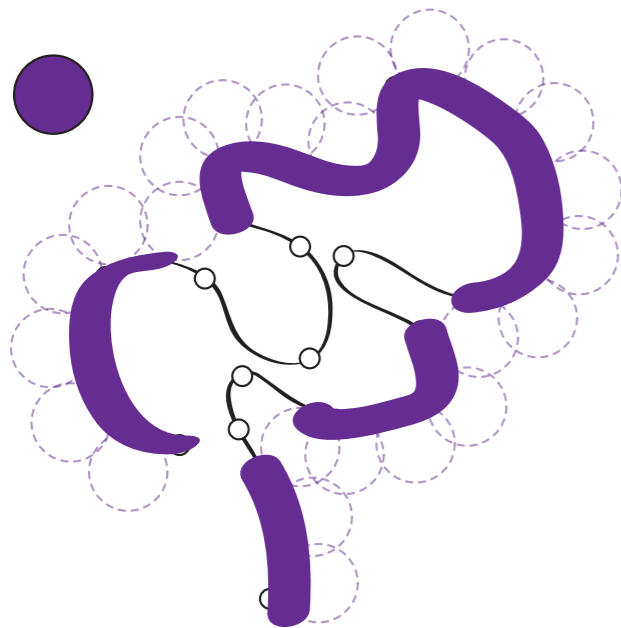
Optimization of the scoring function



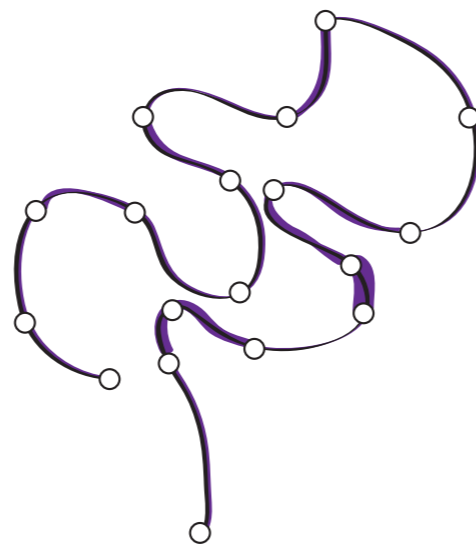
Model analysis: clustering and structural features



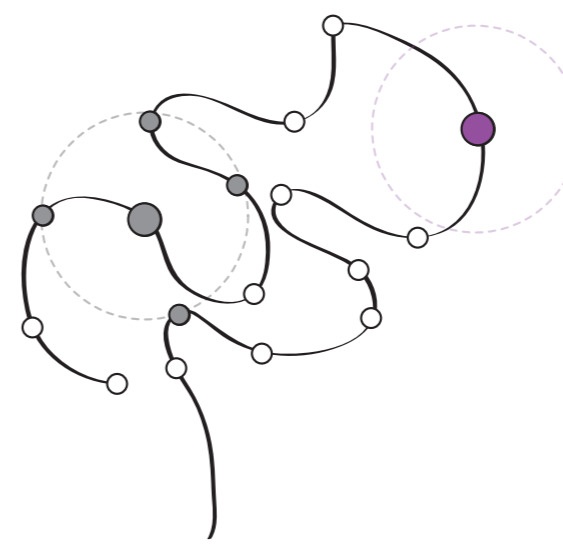
Accessibility (%)



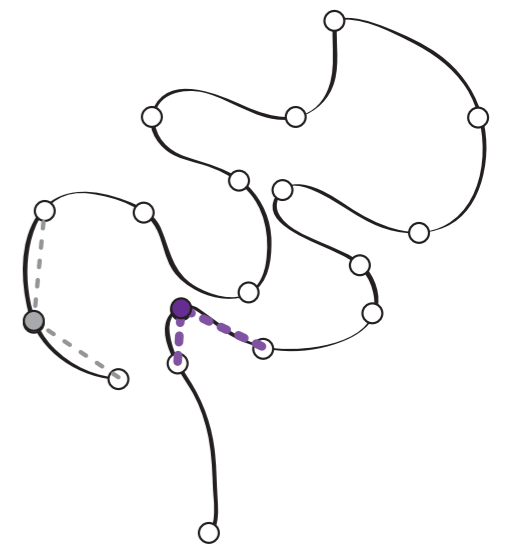
Density (bp/nm)

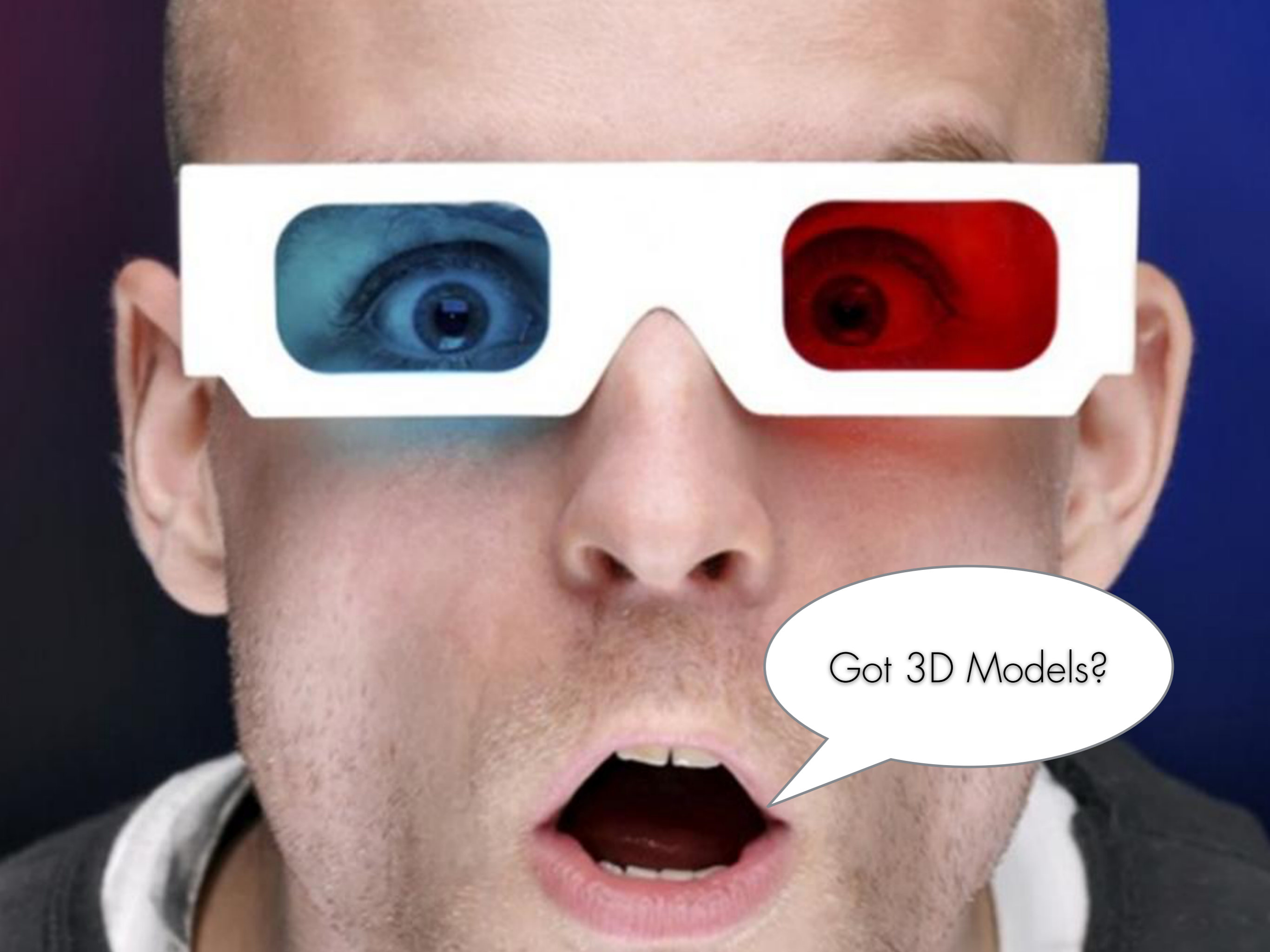


Interactions



Angle

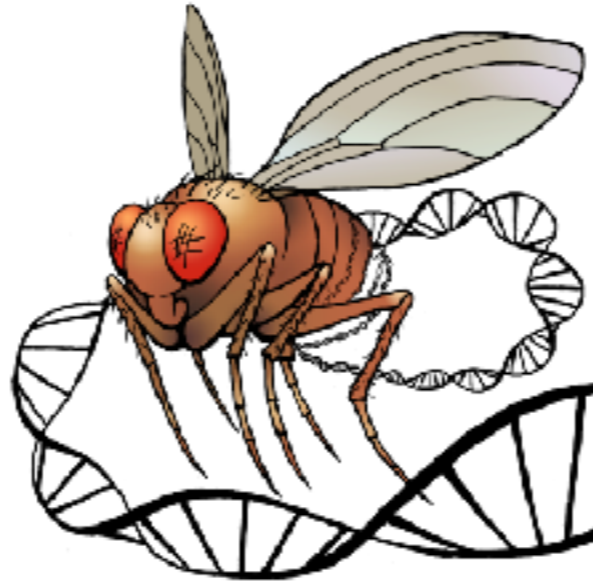




Got 3D Models?

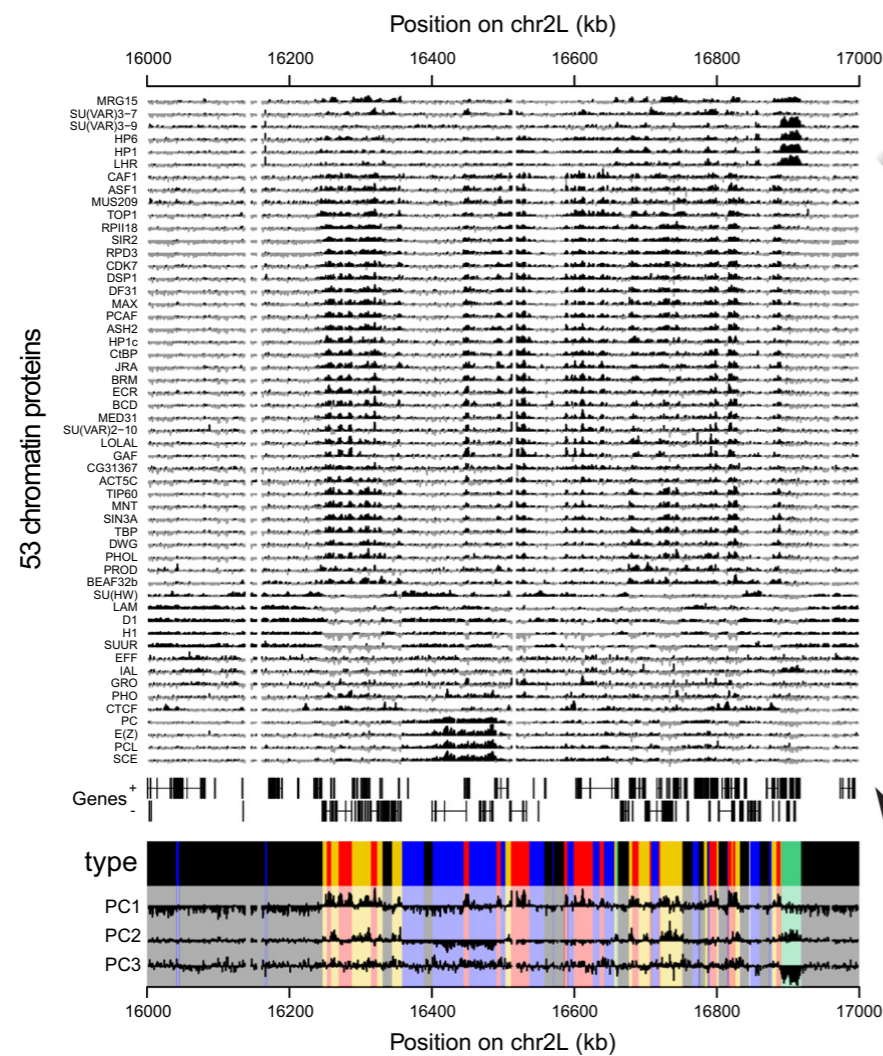
Structuring the **COLORs** of chromatin

Serra, Baù et al. (2017) PLOS CompBio.

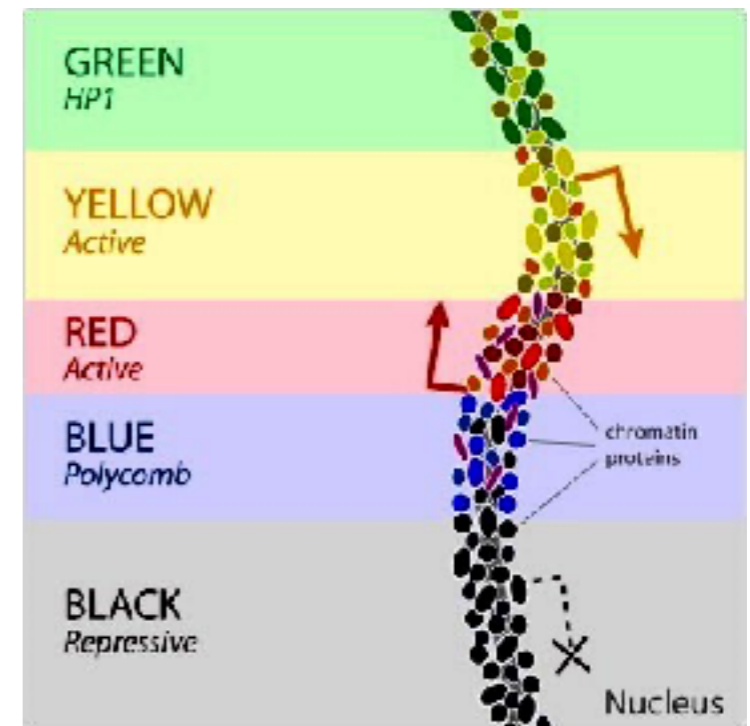
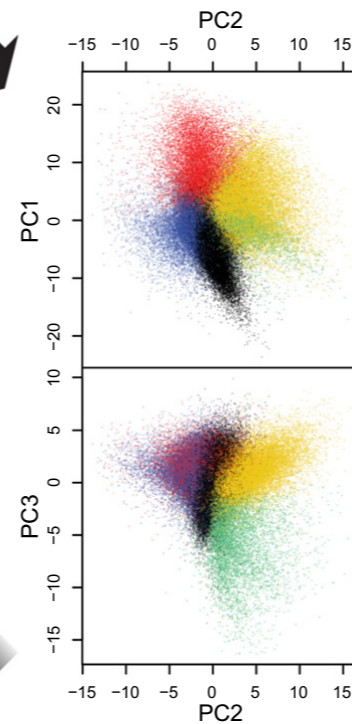


Fly Chromatin **COLORs**

Filion et al. (2010). Cell, 143(2), 212–224.



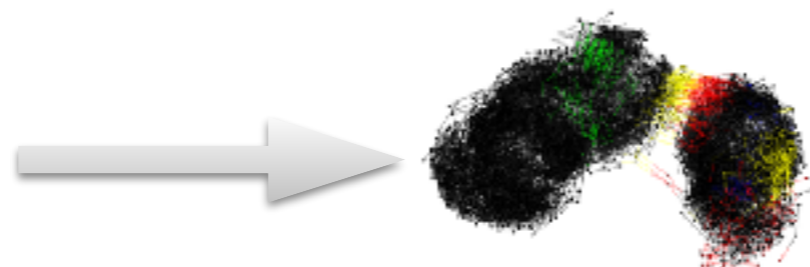
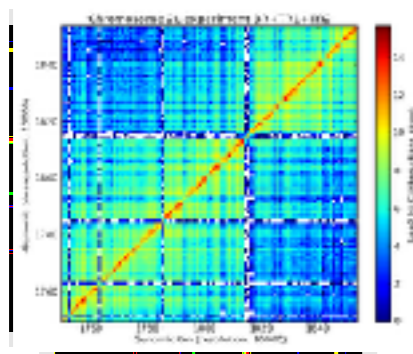
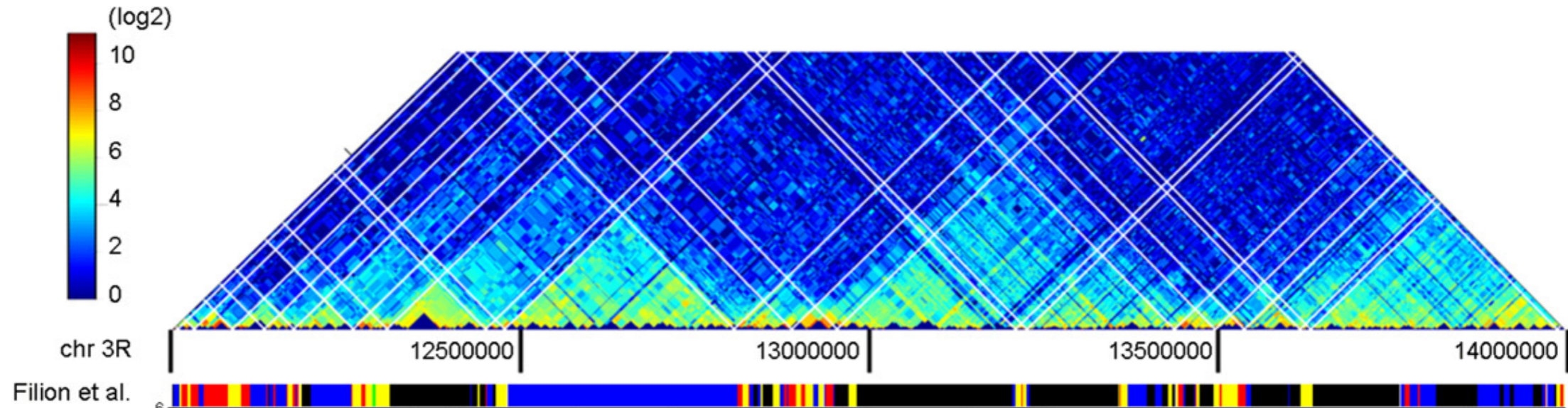
Principal component analysis



Hidden Markov model

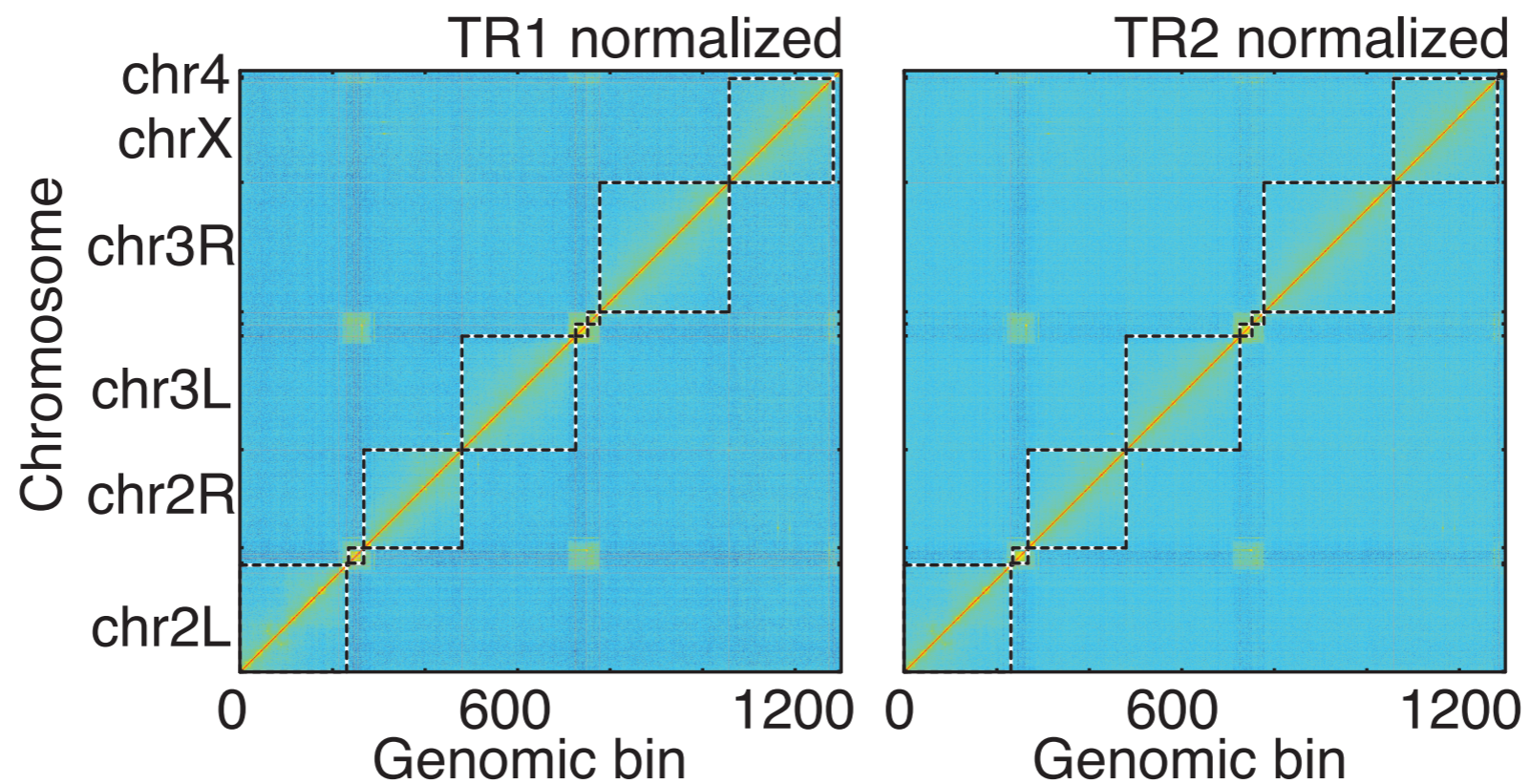
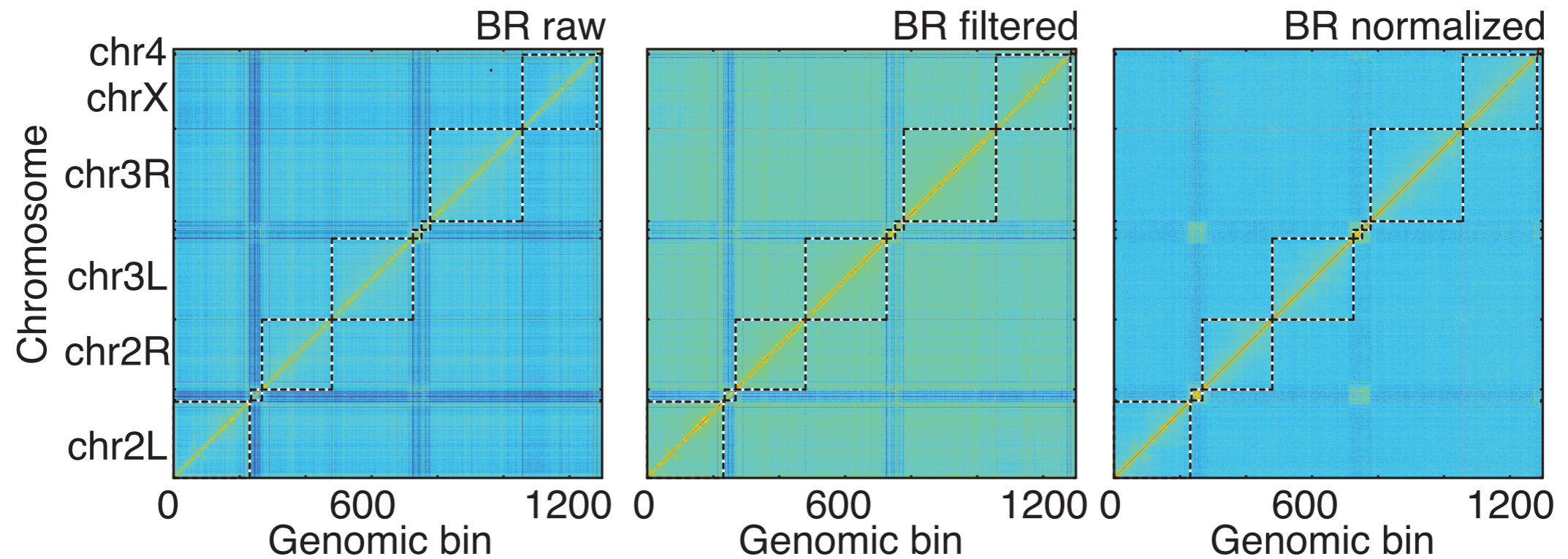
Fly Chromatin **COLORs**

Hou et al. (2012). *Molecular Cell*, 48(3), 471–484.

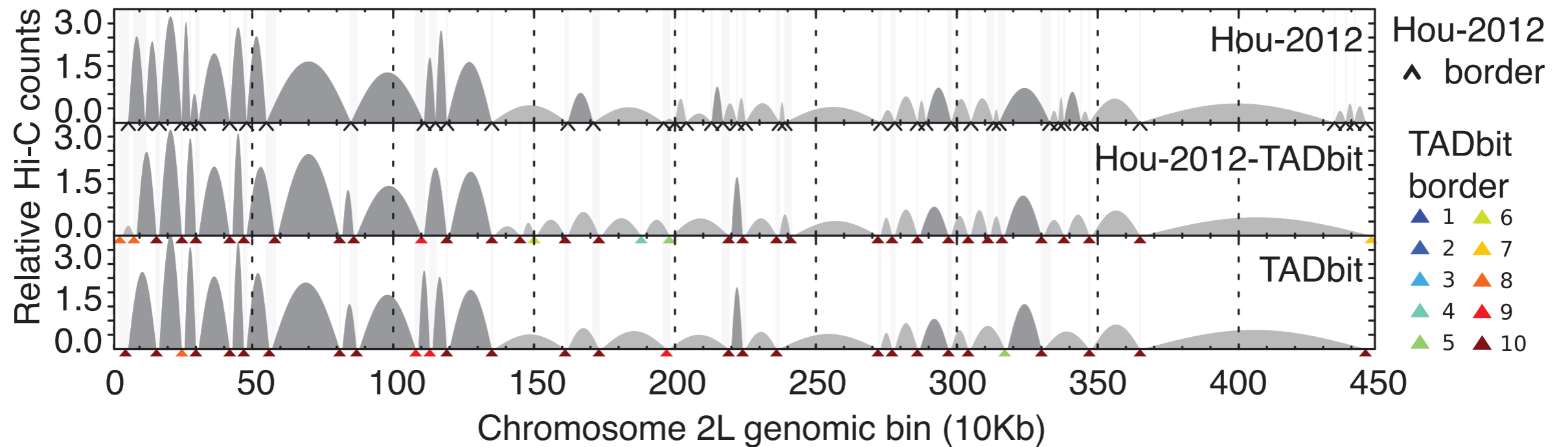
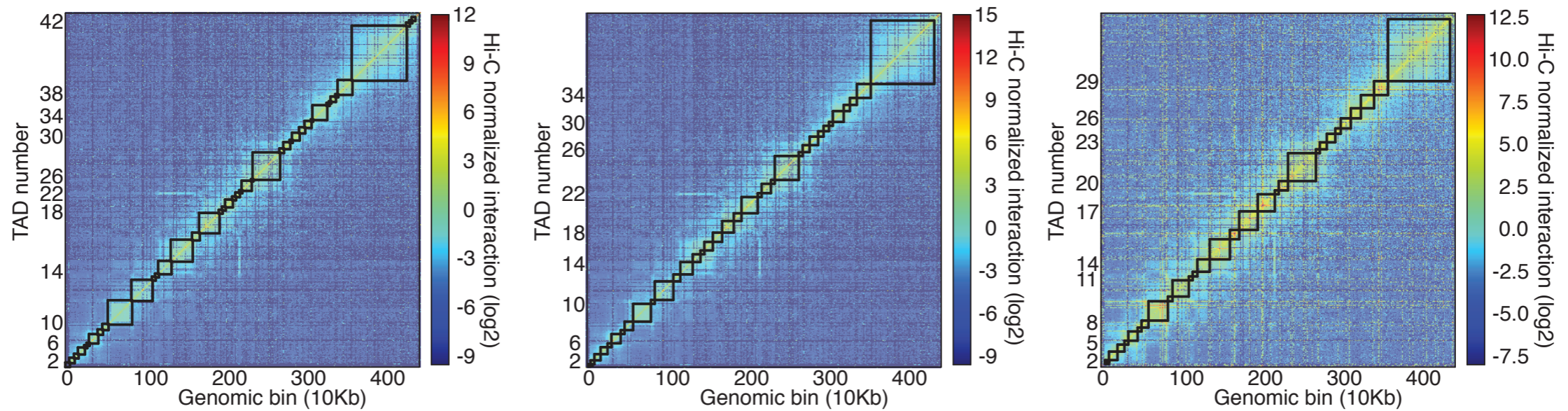


~200 regions of ~5Mb each
2Kb resolution

Mapping · Filtering · Normalizing



TAD detection · comparison



Structural properties

50 1Mb regions. 10 enriched for each color.

RED dense region
3R:18920000-19920000

22% 17% 0% 11% 45% 6%

YELLOW dense region
X:15590000-16600000

0% 48% 4% 20% 26% 3%

GREEN dense region
2R:510000-1530000

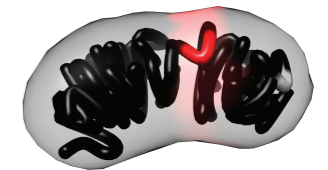
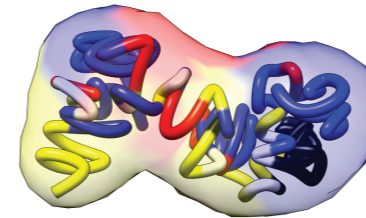
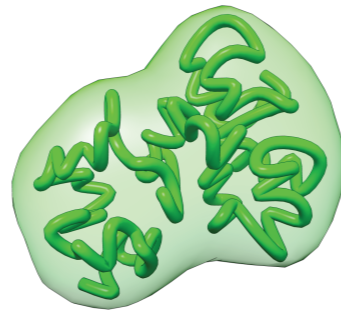
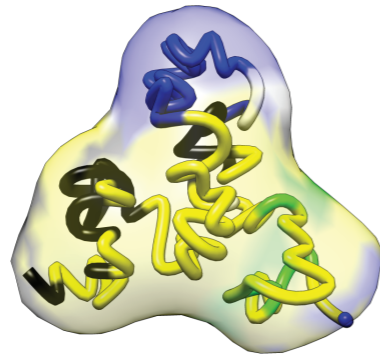
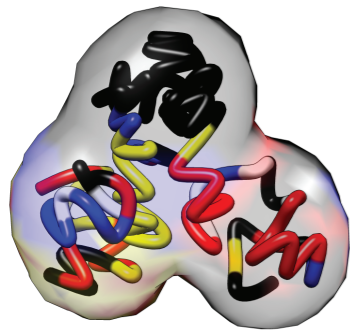
0% 0% 100% 0% 0% 0%

BLUE dense region
3L:210000-1230000

11% 17% 0% 52% 13% 0%

BLACK dense region
2L:17500000-18530000

1% 0% 0% 0% 98% 1%

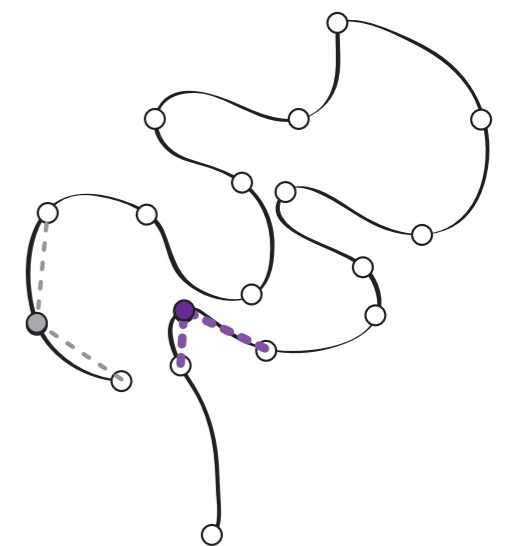
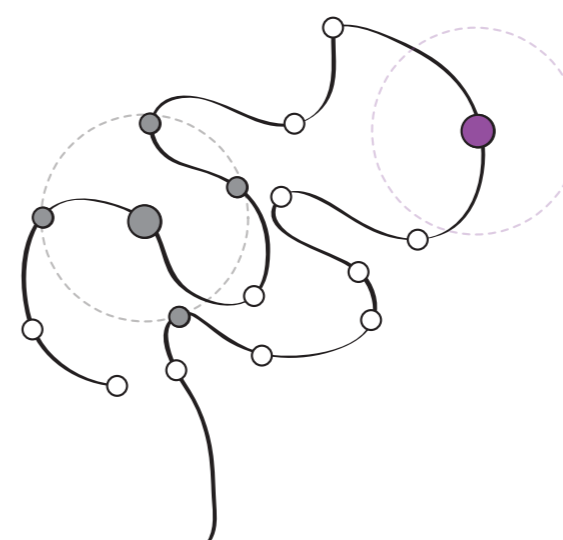
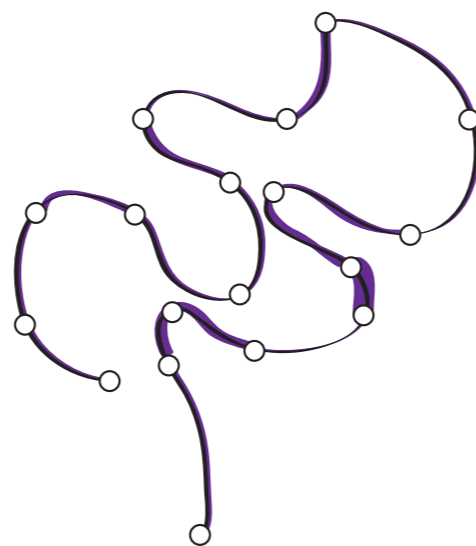
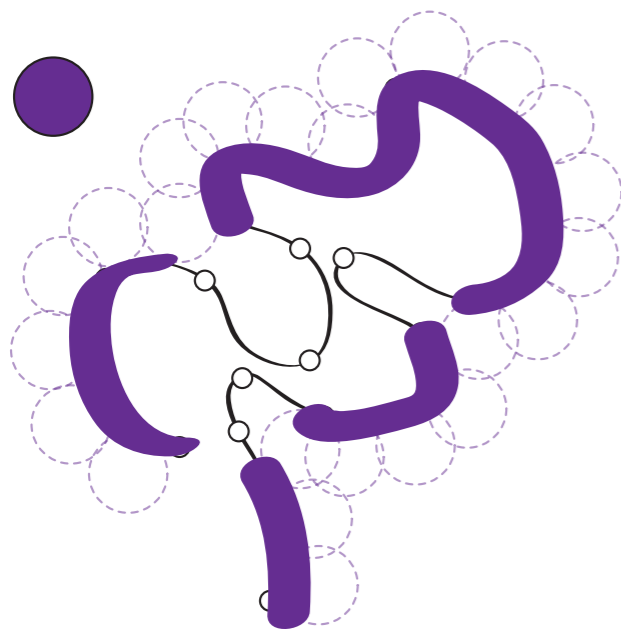


Accessibility (%)

Density (bp/nm)

Interactions

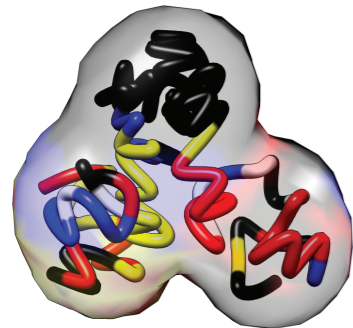
Angle



Structural **COLORs**

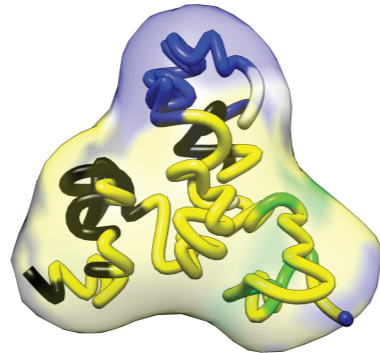
RED dense region
3R:18920000-19920000

22% 17% 0% 11% 45% 6%



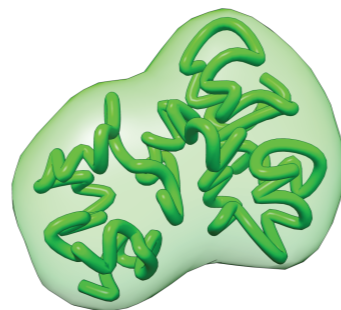
YELLOW dense region
X:15590000-16600000

0% 48% 4% 20% 26% 3%



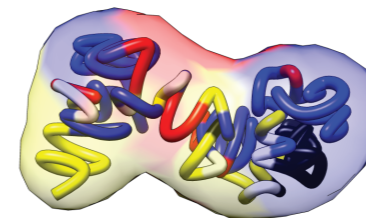
GREEN dense region
2R:510000-1530000

0% 0% 100% 0% 0% 0%



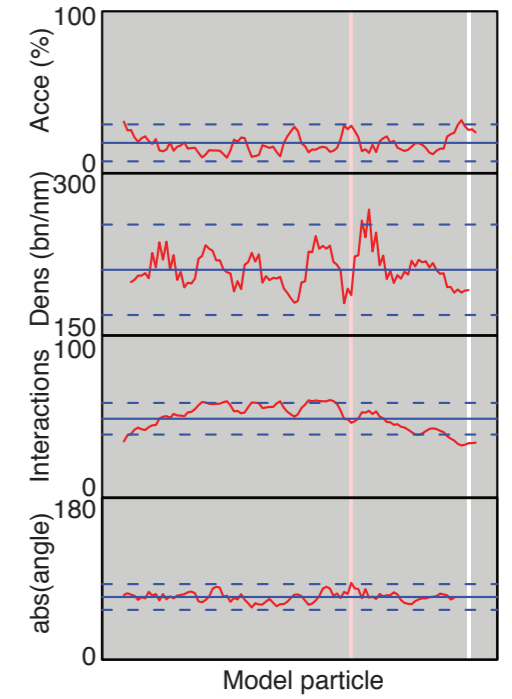
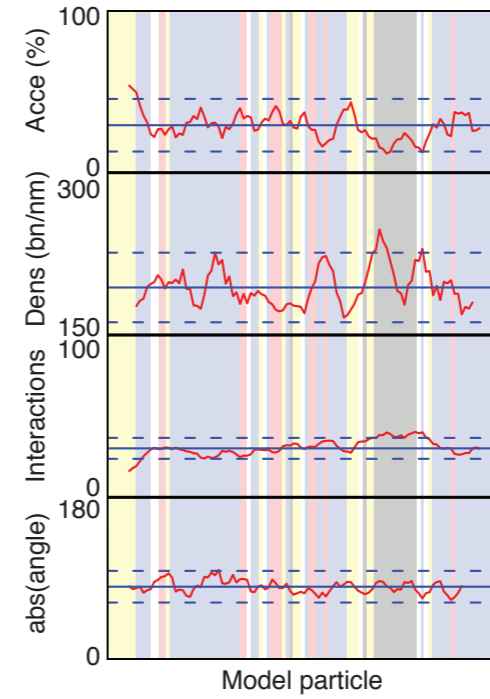
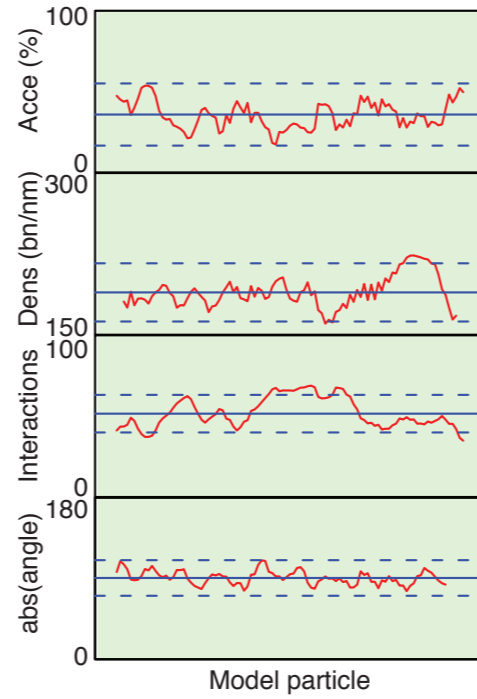
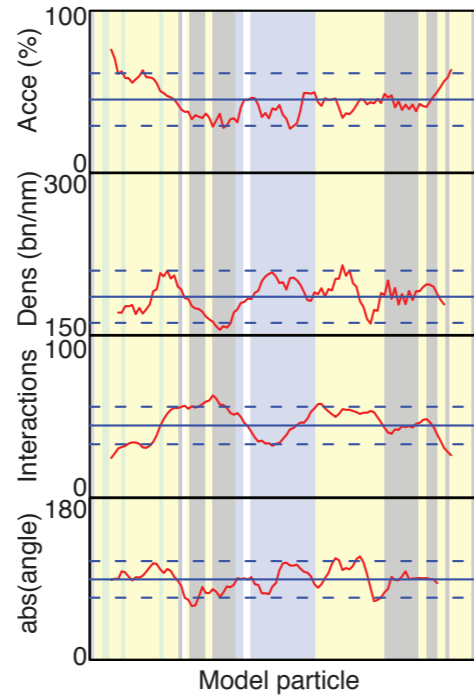
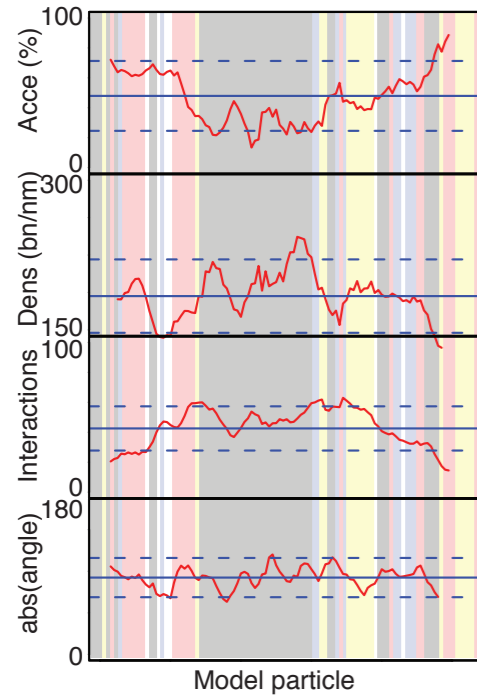
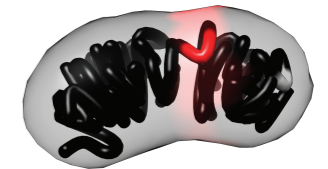
BLUE dense region
3L:210000-1230000

11% 17% 0% 52% 13% 0%

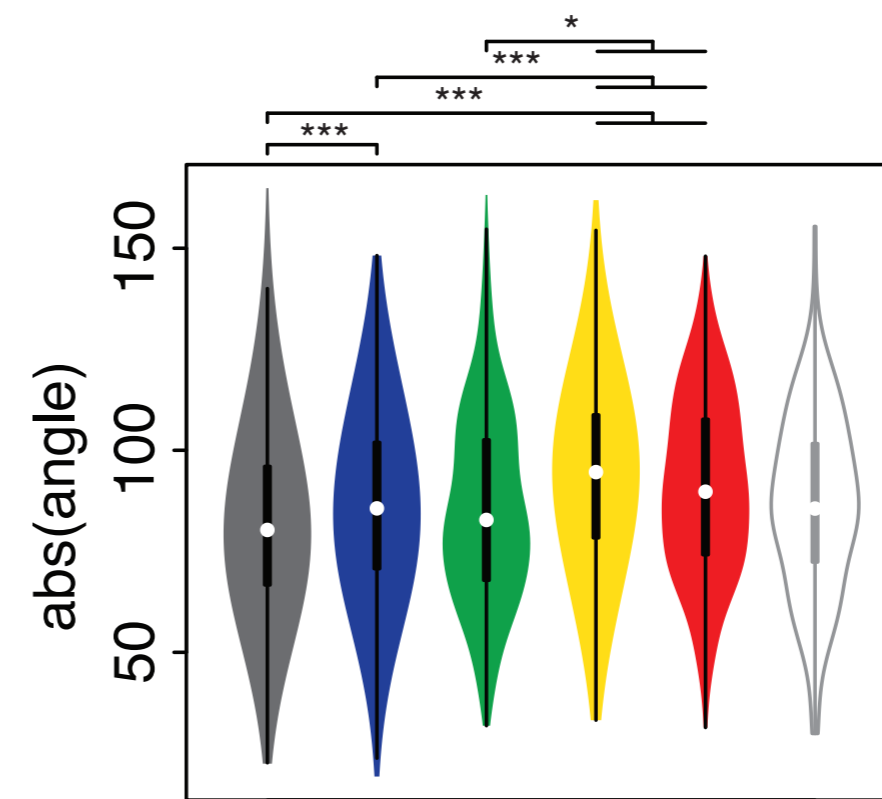
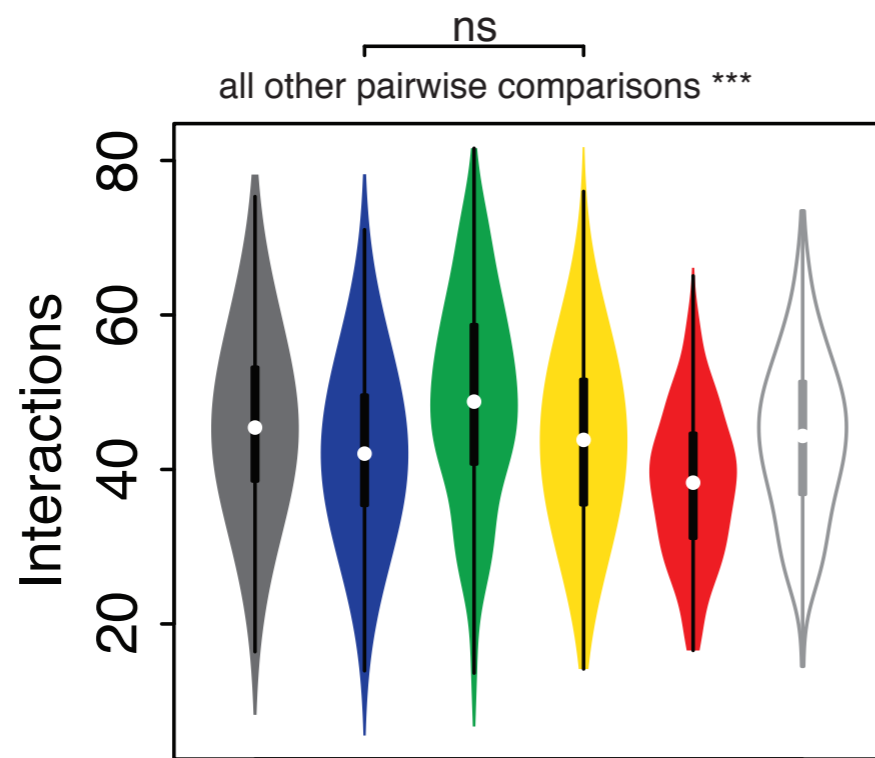
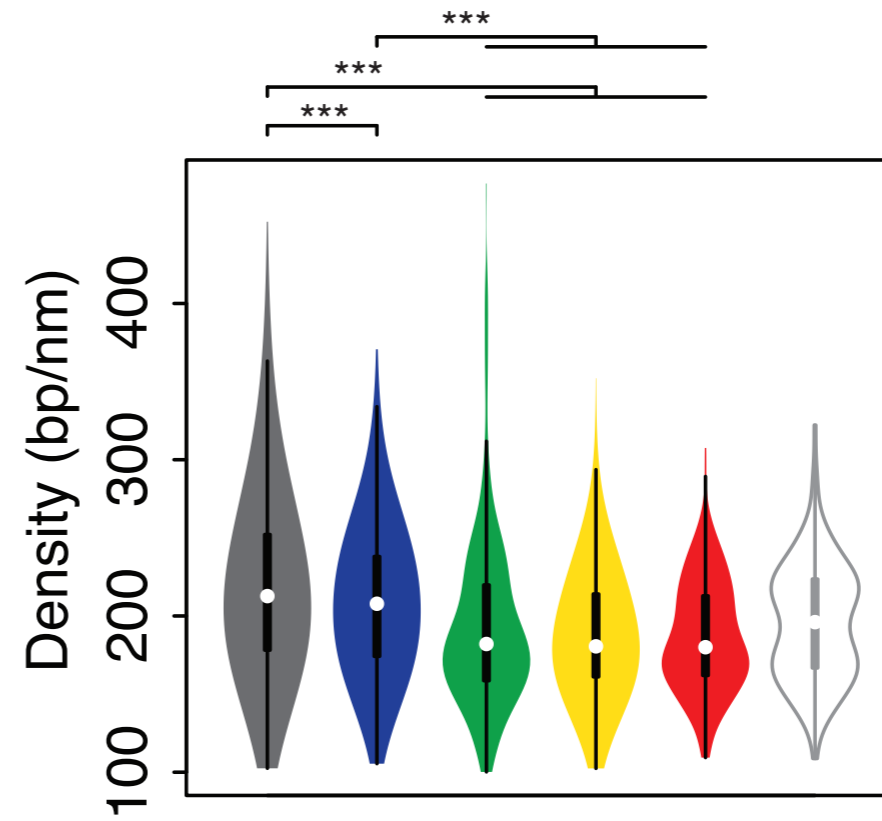
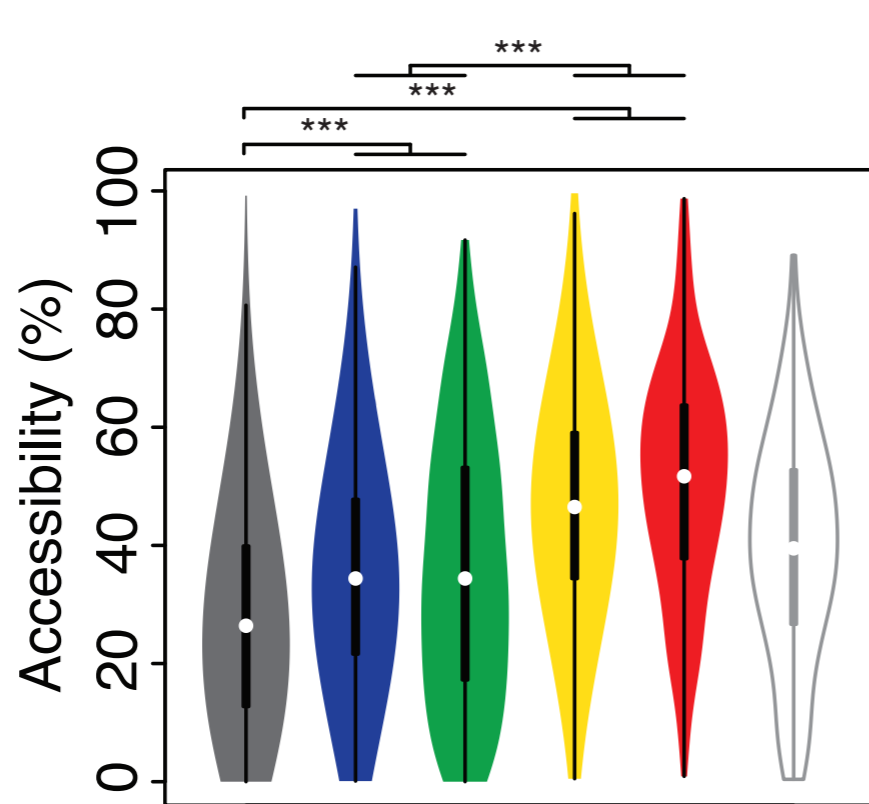


BLACK dense region
2L:17500000-18530000

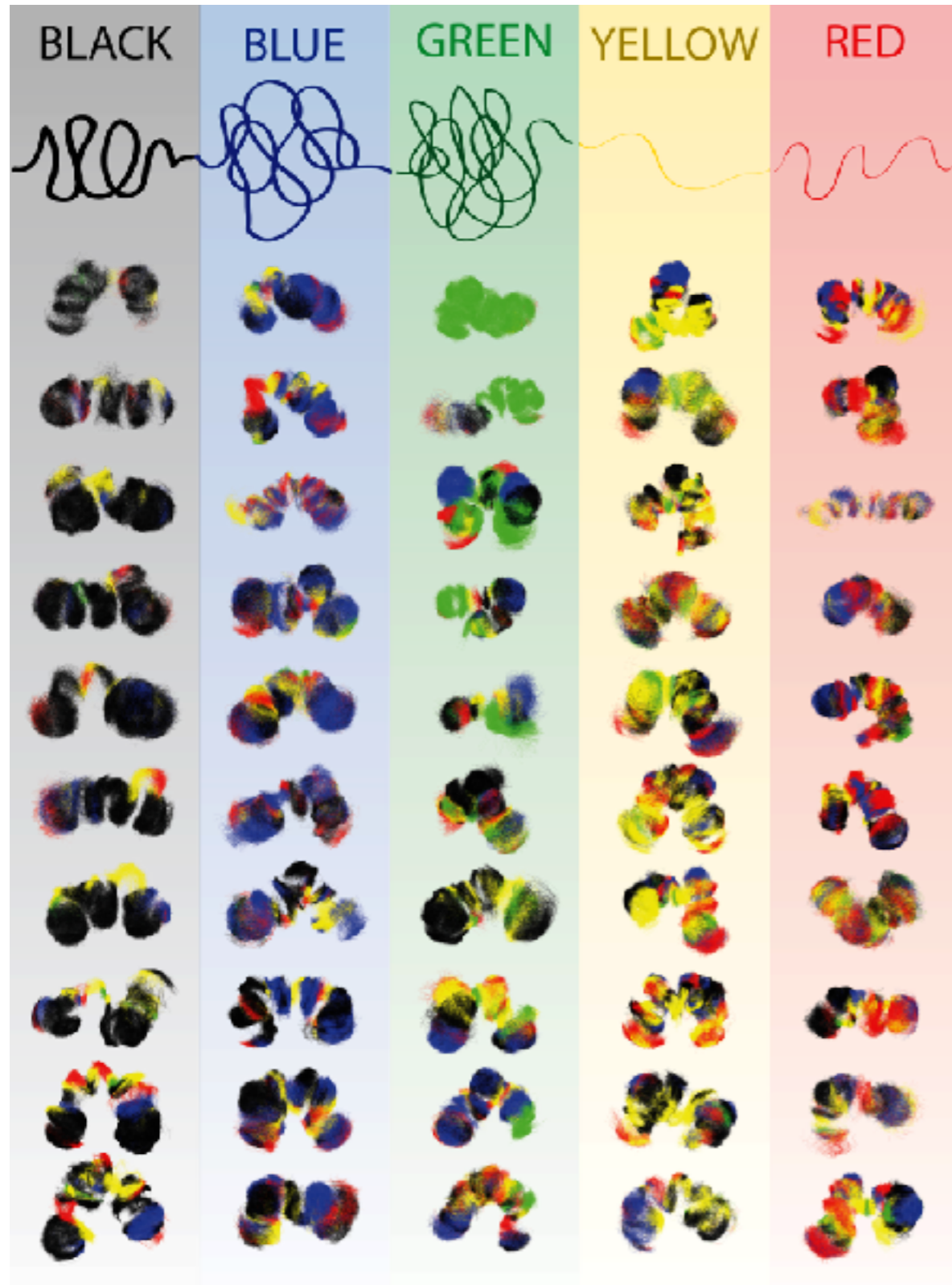
1% 0% 0% 0% 98% 1%



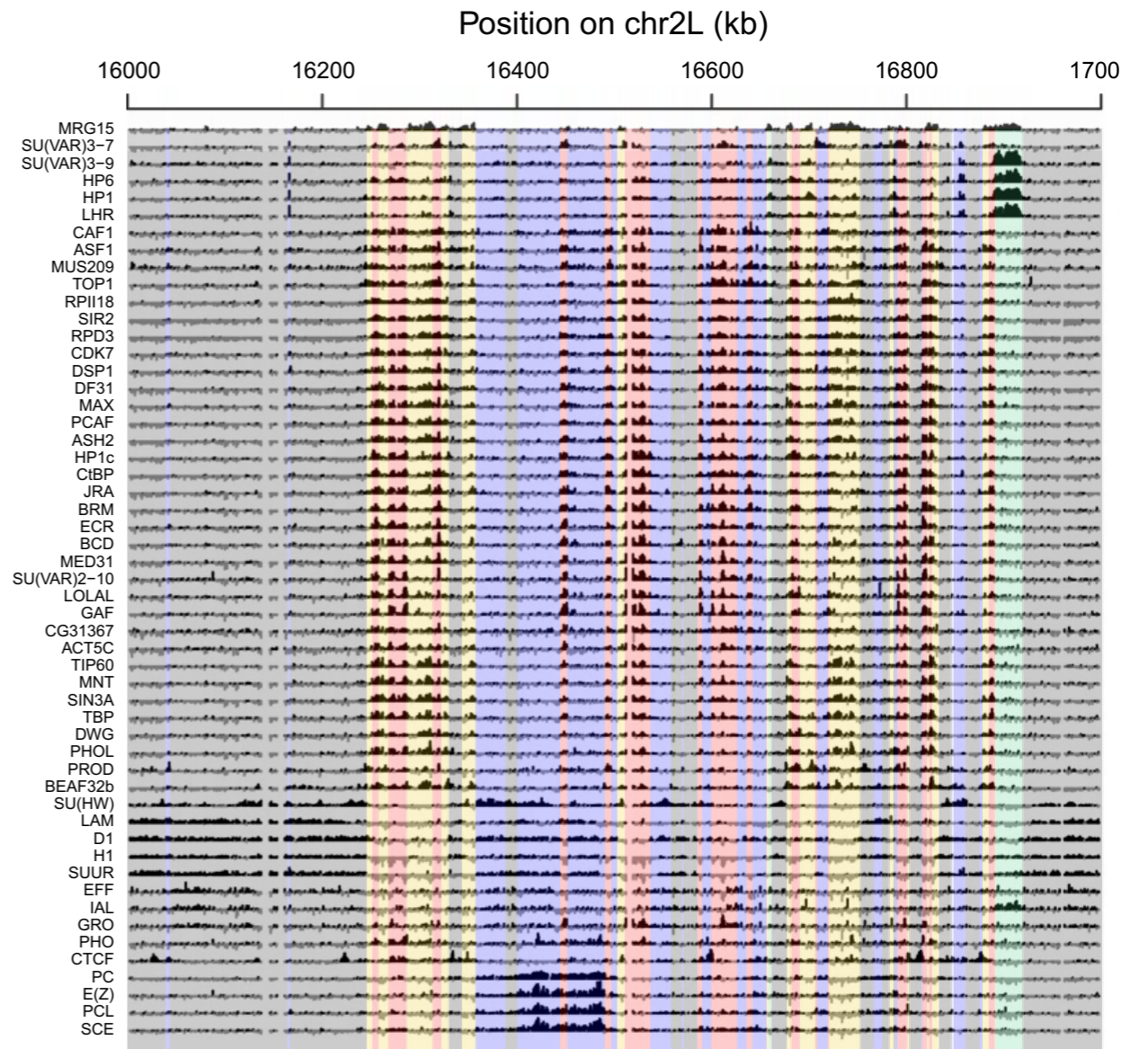
Structural **COLORs**

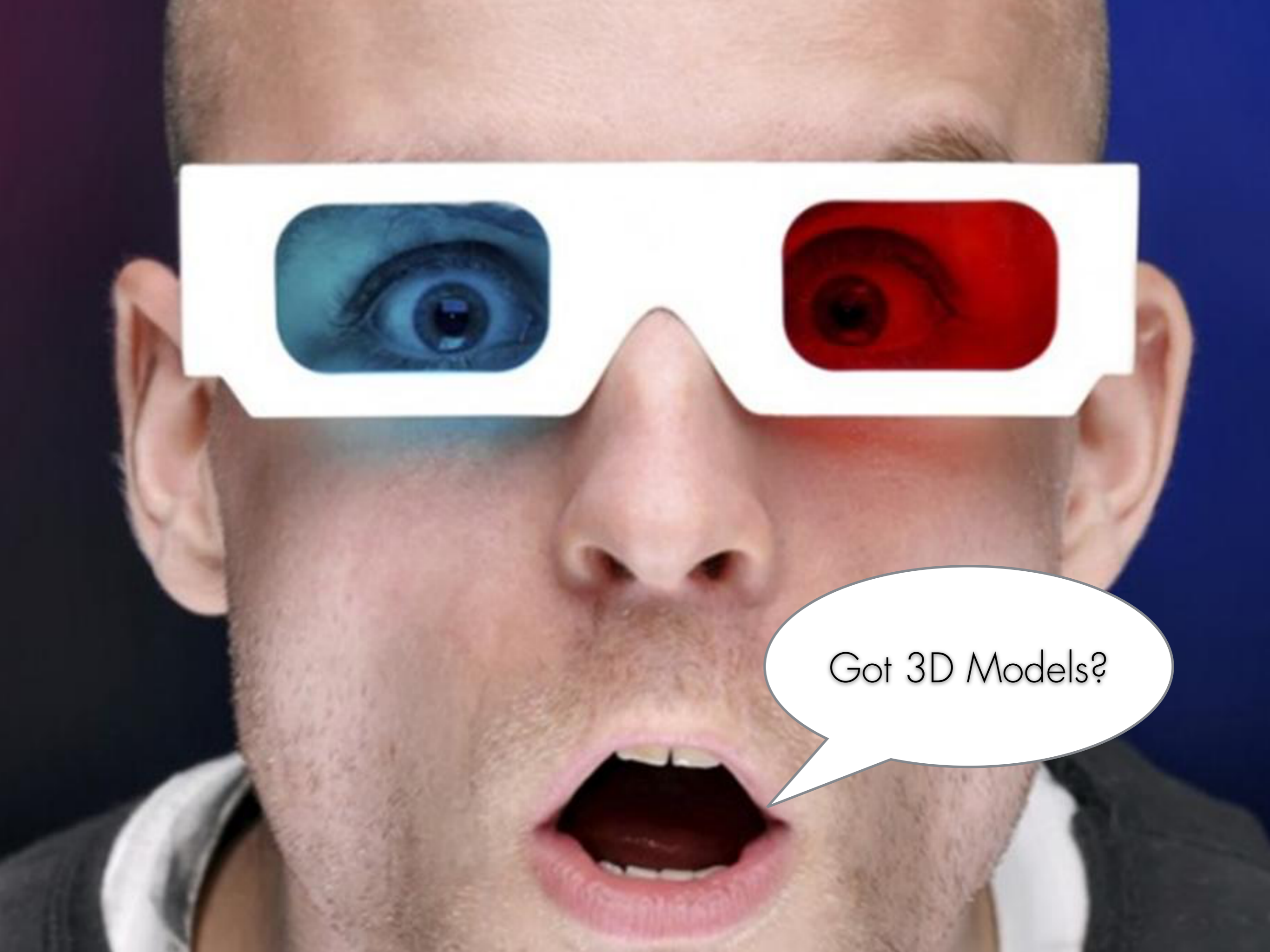


Structural **CO**LO**R**s



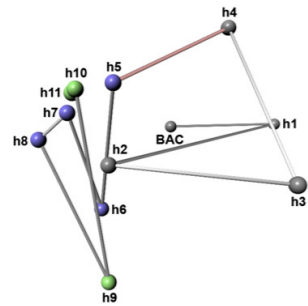
53 chromatin proteins



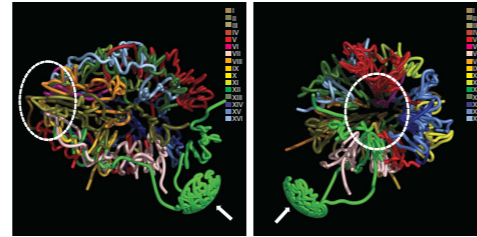


Got 3D Models?

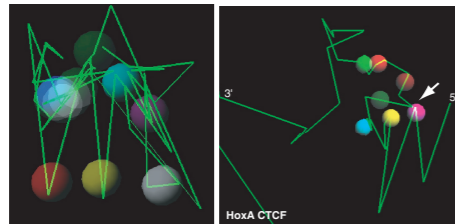
Are the models correct?



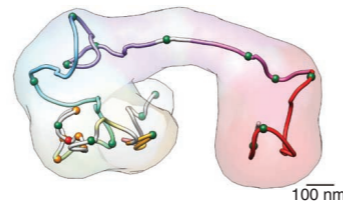
Jhunjhunwala (2008) Cell



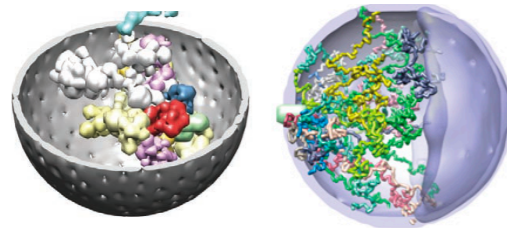
Duan (2010) Nature



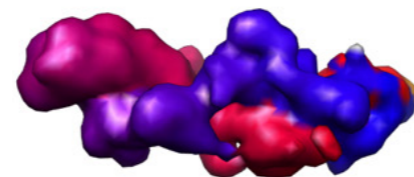
Fraser (2009) Genome Biology
Ferraiuolo (2010) Nucleic Acids Research



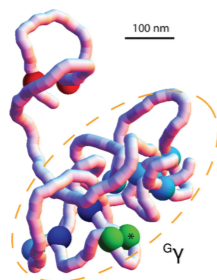
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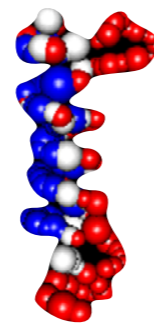
Kalhor (2011) Nature Biotechnology
Tjong (2012) Genome Research



Umbarger (2011) Molecular Cell



Junier (2012) Nucleic Acids Research



Hu (2013) PLoS Computational Biology

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Nucleic Acids Research, 2015, 1
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Assessing the limits of restraint-based 3D modeling of genomes and genomic domains

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ABSTRACT

Restraint-based modeling of genomes has been recently explored with the advent of Chromosome Conformation Capture (3C-based) experiments. We previously developed a reconstruction method to resolve the 3D architecture of both prokaryotic and eukaryotic genomes using 3C-based data. These models were congruent with fluorescent imaging validation. However, the limits of such methods have not systematically been assessed. Here we propose the first evaluation of a mean-field restraint-based reconstruction of genomes by considering diverse chromosome architectures and different levels of data noise and structural variability. The results show that: first, current scoring functions for 3D reconstruction correlate with the accuracy of the models; second, reconstructed models are robust to noise but sensitive to structural variability; third, the local structure organization of genomes, such as Topologically Associating Domains, results in more accurate models; fourth, to a certain extent, the models capture the intrinsic structural variability in the input matrices and fifth, the accuracy of the models can be *a priori* predicted by analyzing the properties of the interaction matrices. In summary, our work provides a systematic analysis of the limitations of a mean-field restraint-based method, which could be taken into consideration in further development of methods as well as their applications.

INTRODUCTION

Recent studies of the three-dimensional (3D) conformation of genomes are revealing insights into the organization and the regulation of biological processes, such as gene

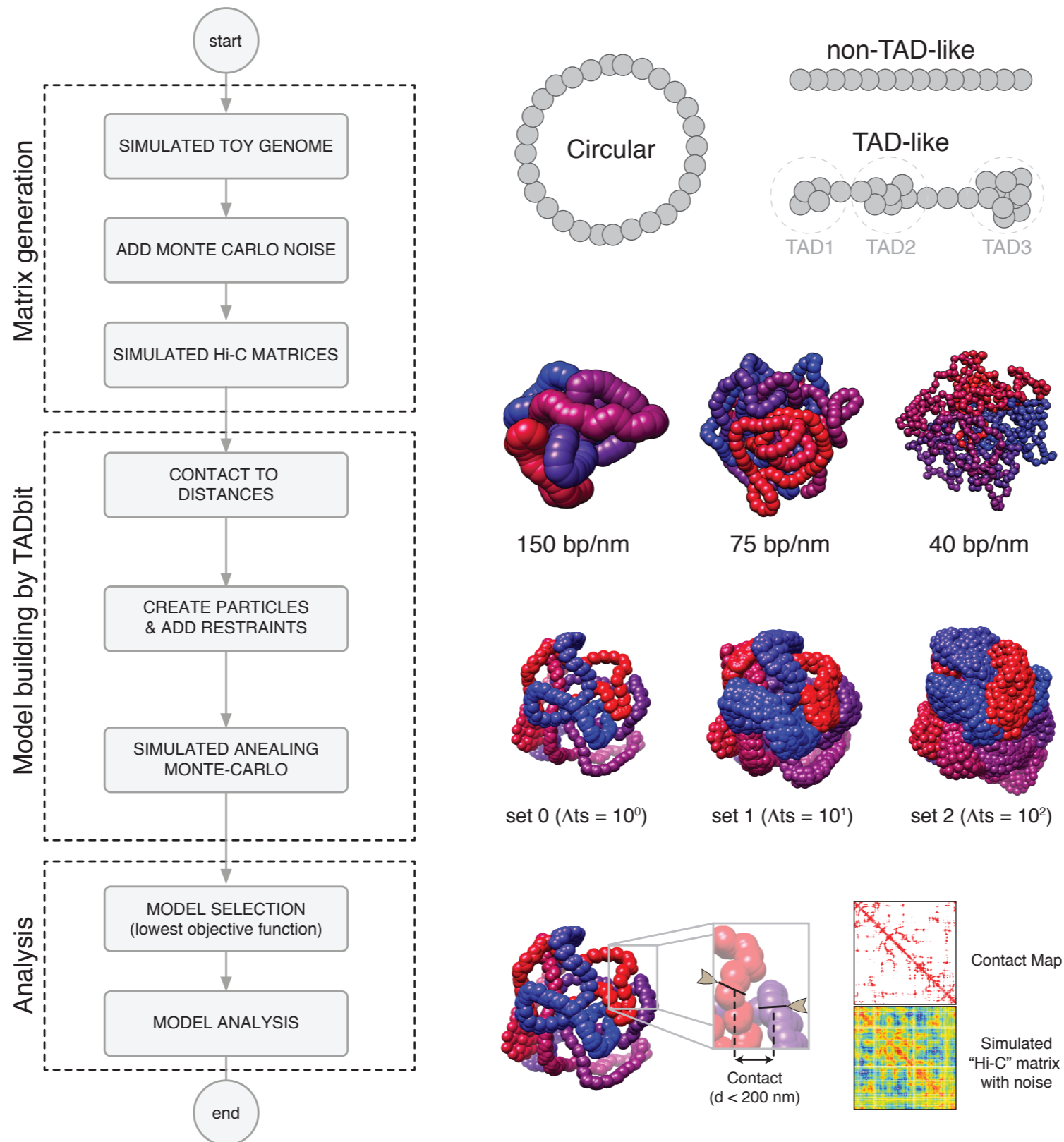
expression regulation and replication (1–6). The advent of the so-called Chromosome Conformation Capture (3C) assays (7), which allowed identifying chromatin-looping interactions between pairs of loci, helped deciphering some of the key elements organizing the genomes. High-throughput derivations of genome-wide 3C-based assays were established with Hi-C technologies (8) for an unbiased identification of chromatin interactions. The resulting genome interaction matrices from Hi-C experiments have been extensively used for computationally analyzing the organization of genomes and genomic domains (5). In particular, a significant number of new approaches for modeling the 3D organization of genomes have recently flourished (9–14). The main goal of such approaches is to provide an accurate 3D representation of the bi-dimensional interaction matrices, which can then be more easily explored to extract biological insights. One type of methods for building 3D models from interaction matrices relies on the existence of a limited number of conformational states in the cell. Such methods are regarded as mean-field approaches and are able to capture, to a certain degree, the structural variability around these mean structures (15).

We recently developed a mean-field method for modeling 3D structures of genomes and genomic domains based on 3C interaction data (9). Our approach, called TADbit, was developed around the Integrative Modeling Platform (IMP, <http://integrativemodeling.org>), a general framework for restraint-based modeling of 3D bio-molecular structures (16). Briefly, our method uses chromatin interaction frequencies derived from experiments as a proxy of spatial proximity between the ligation products of the 3C libraries. Two fragments of DNA that interact with high frequency are dynamically placed close in space in our models while two fragments that do not interact as often will be kept apart. Our method has been successfully applied to model the structures of genomes and genomic domains in eukaryote and prokaryote organisms (17–19). In all of our studies, the final models were partially validated by assessing their

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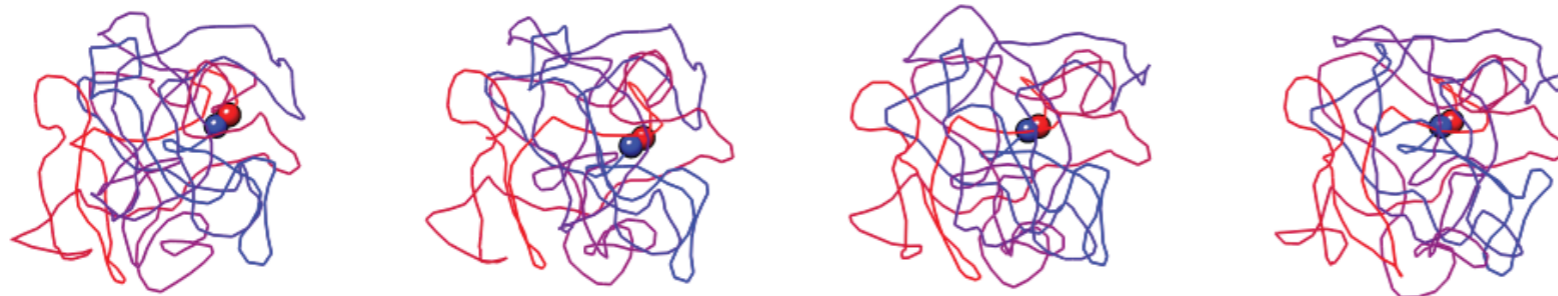
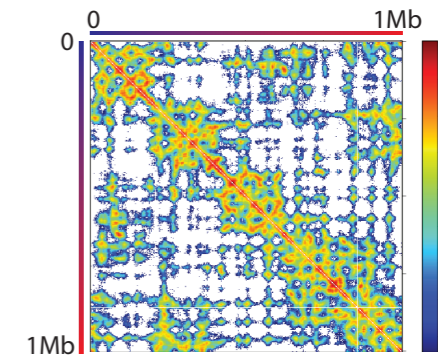
Toy models



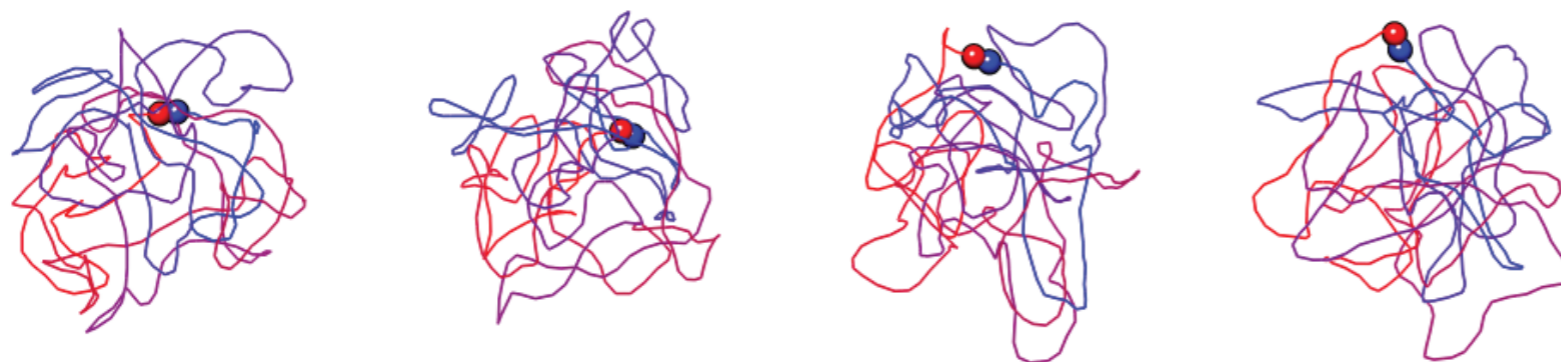
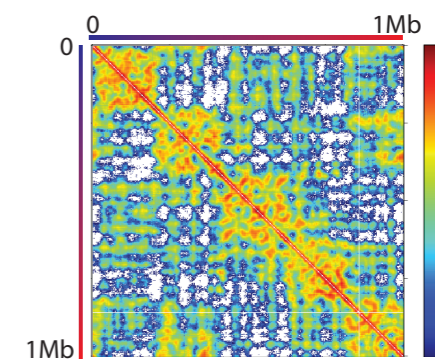
Toy interaction matrices



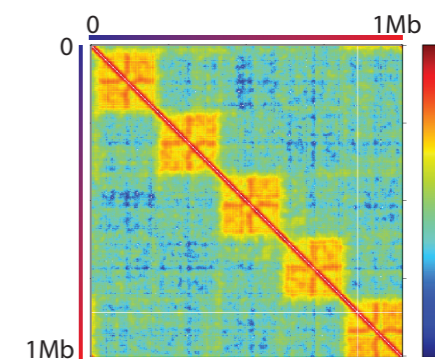
set 0 ($\Delta ts = 10^0$)



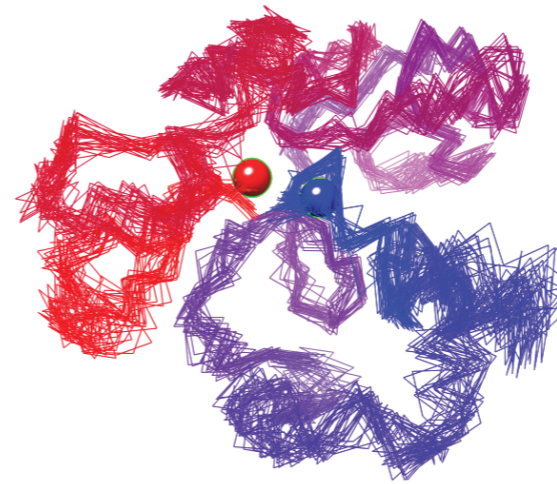
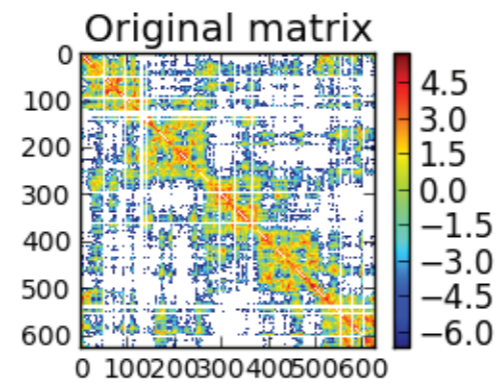
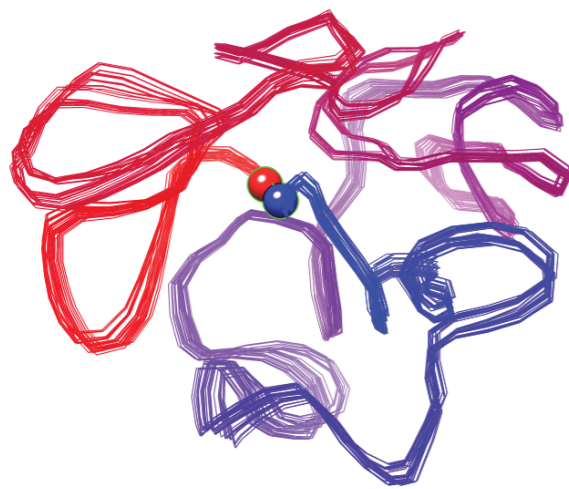
set 4 ($\Delta ts = 10^4$)



set 6 ($\Delta ts = 10^6$)



Reconstructing toy models



chr40_TAD

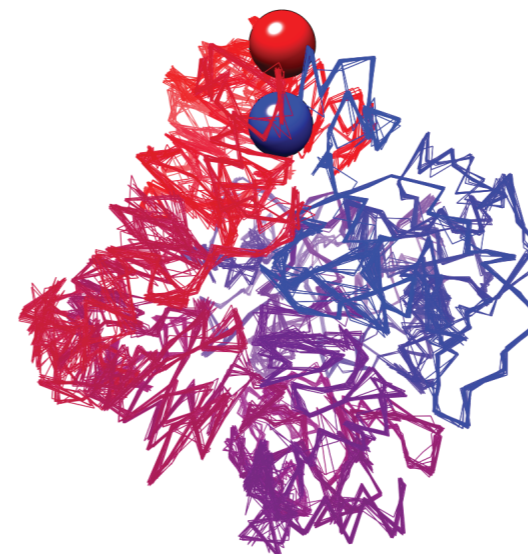
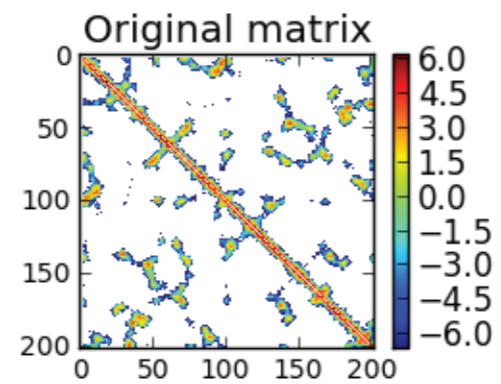
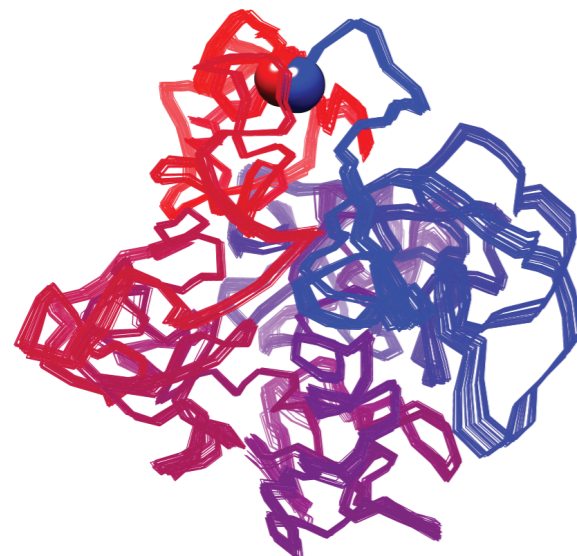
$\alpha=100$

$\Delta ts=10$

TADbit-SCC: 0.91

$\langle dRMSD \rangle$: 32.7 nm

$\langle dSCC \rangle$: 0.94



chr150_TAD

$\alpha=50$

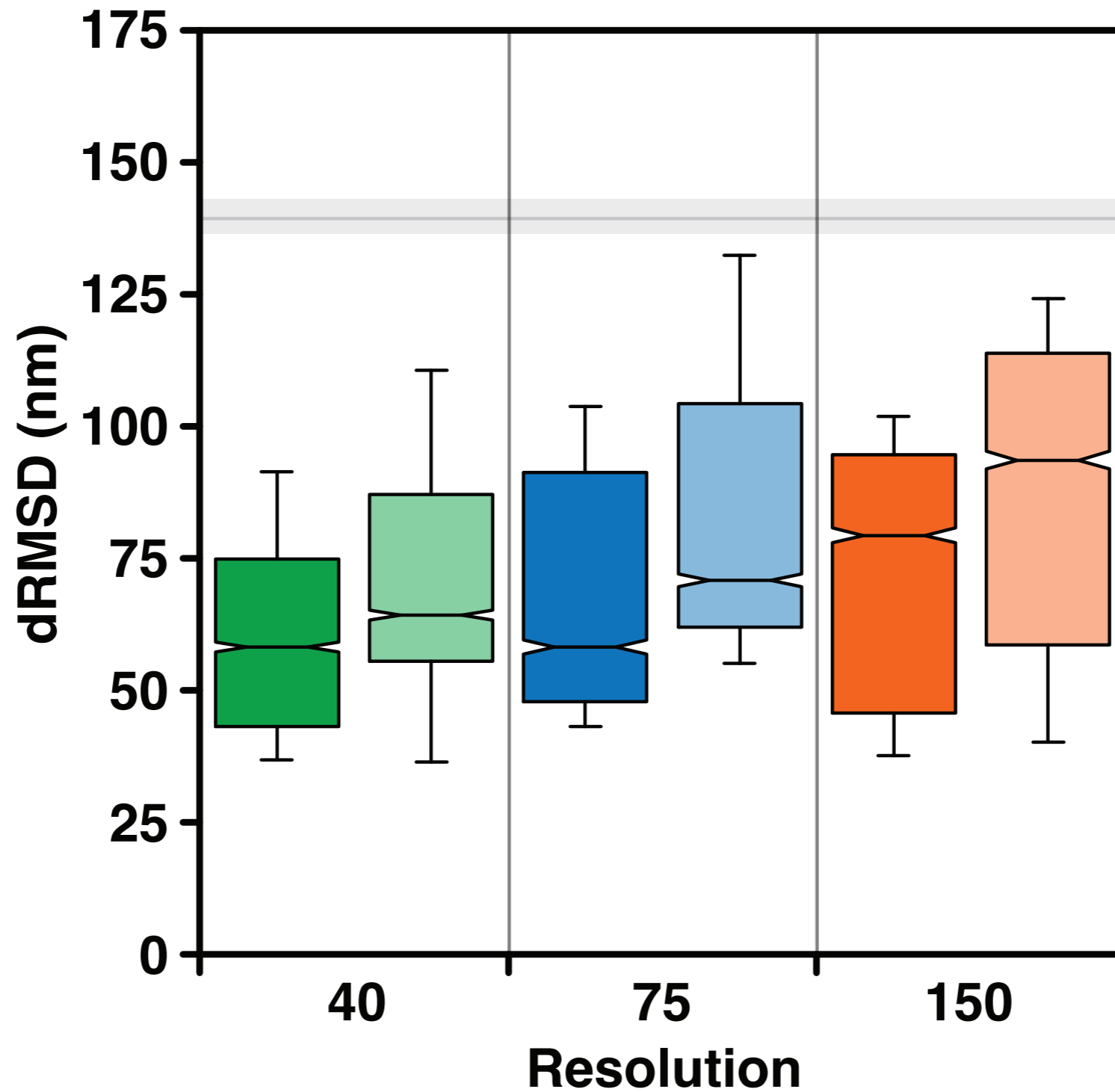
$\Delta ts=1$

TADbit-SCC: 0.82

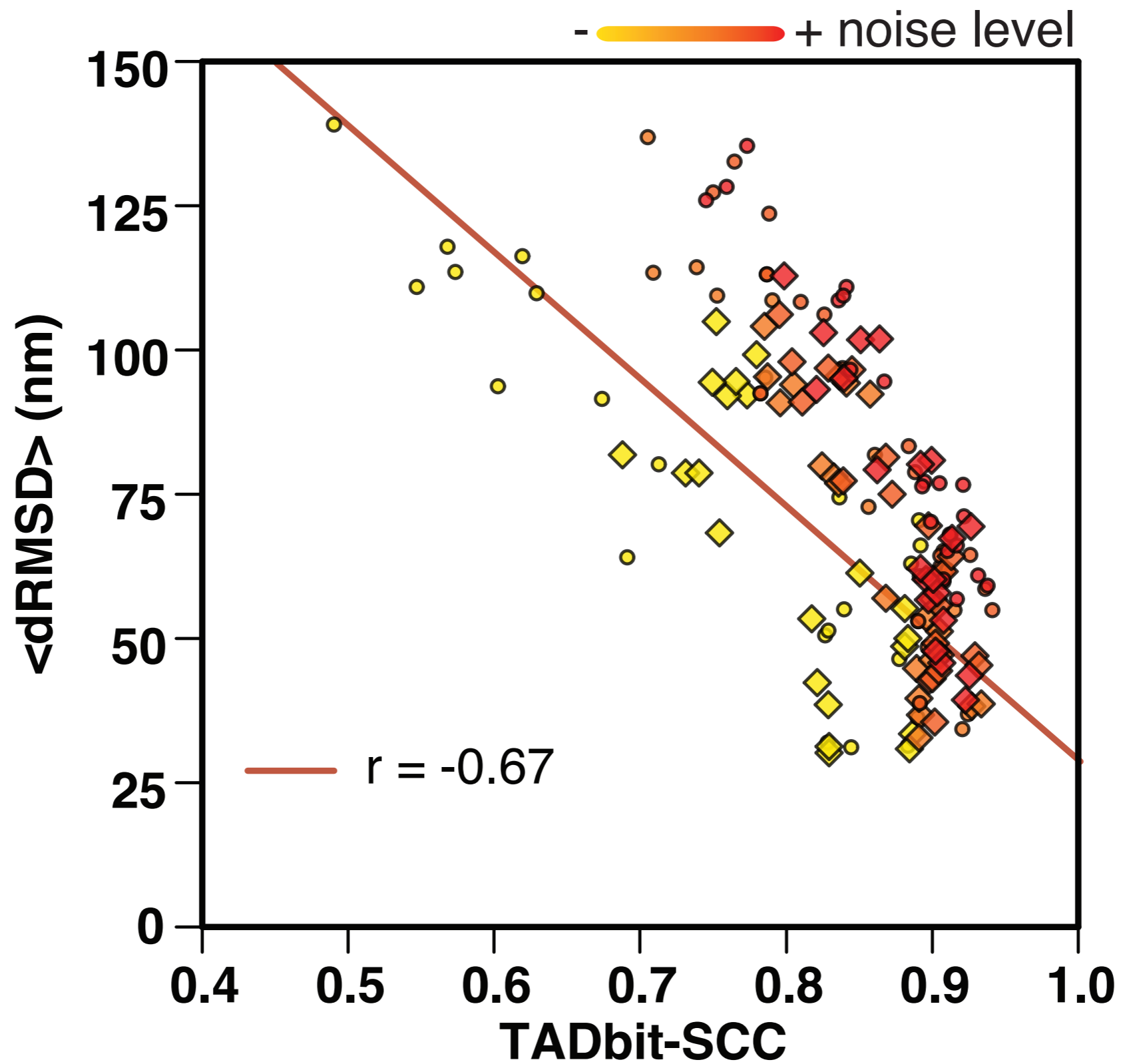
$\langle dRMSD \rangle$: 45.4 nm

$\langle dSCC \rangle$: 0.86

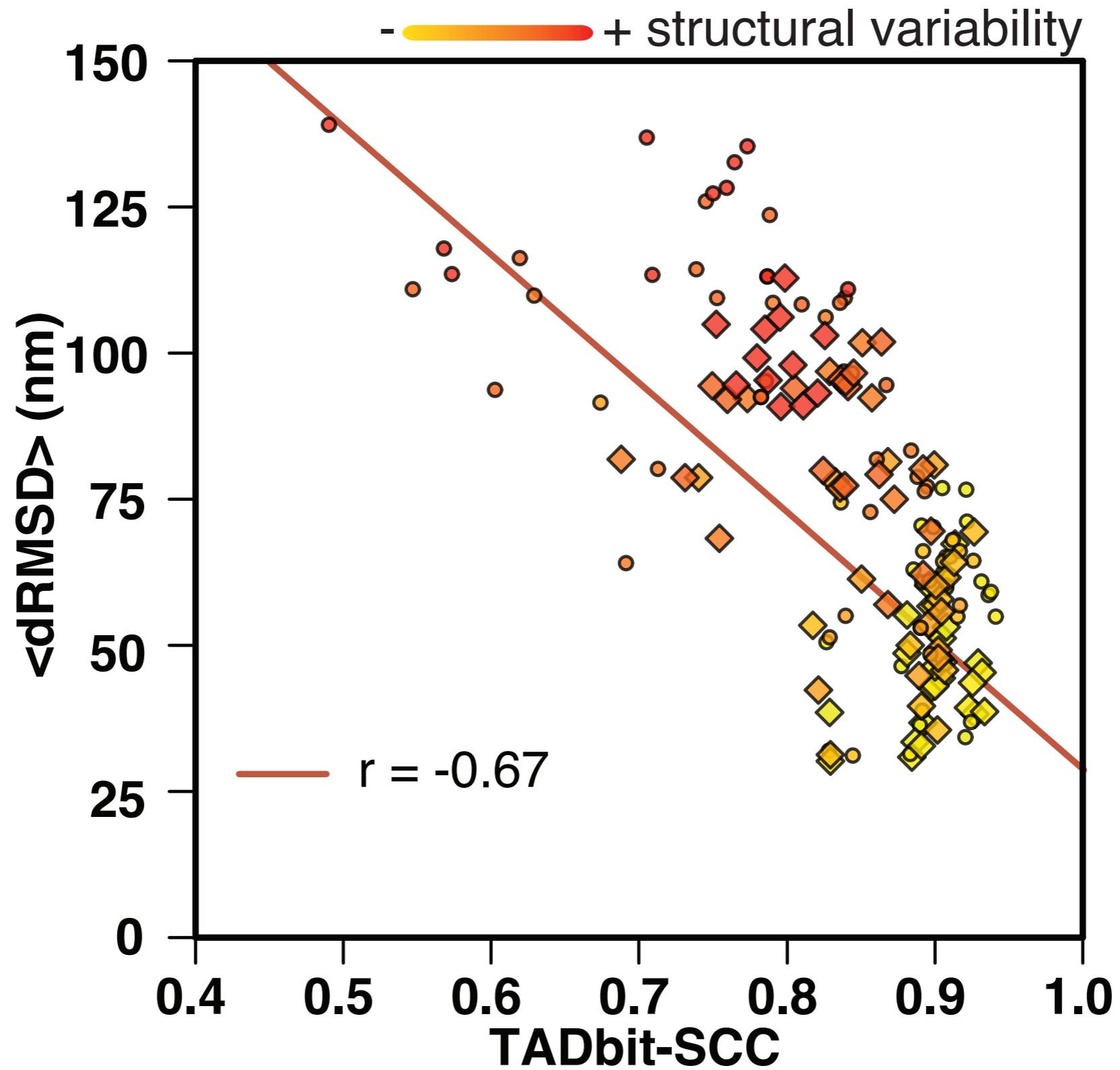
TADs & higher-res are "good"



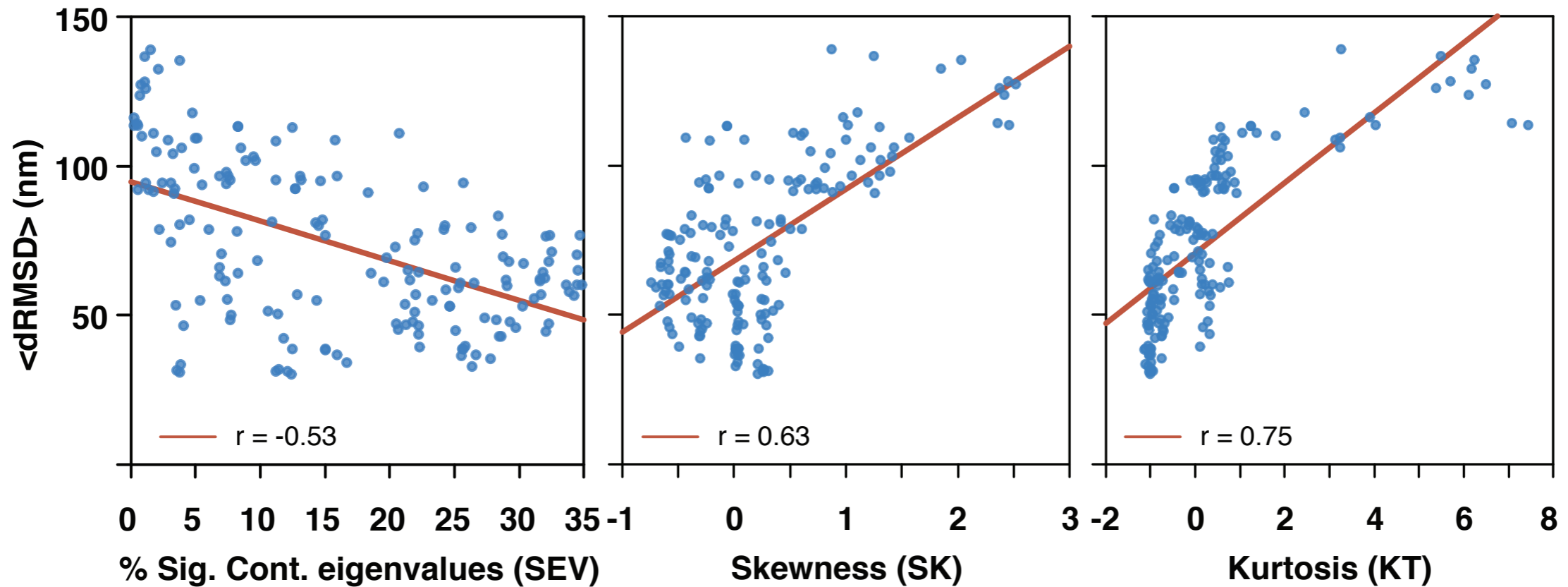
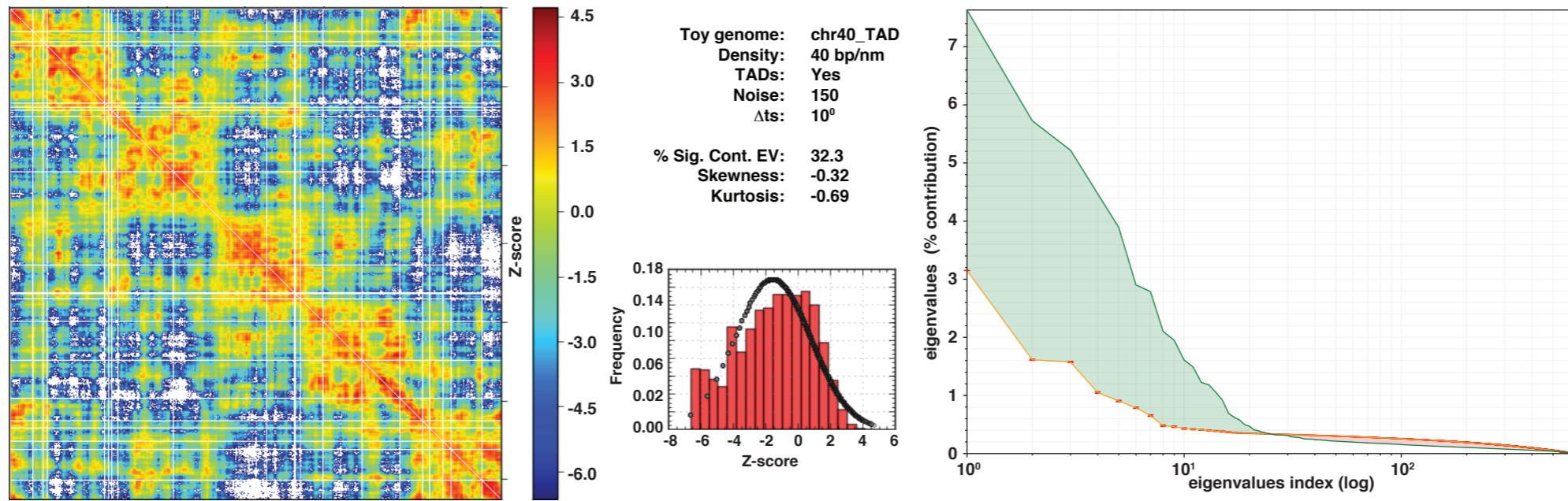
Noise is "OK"



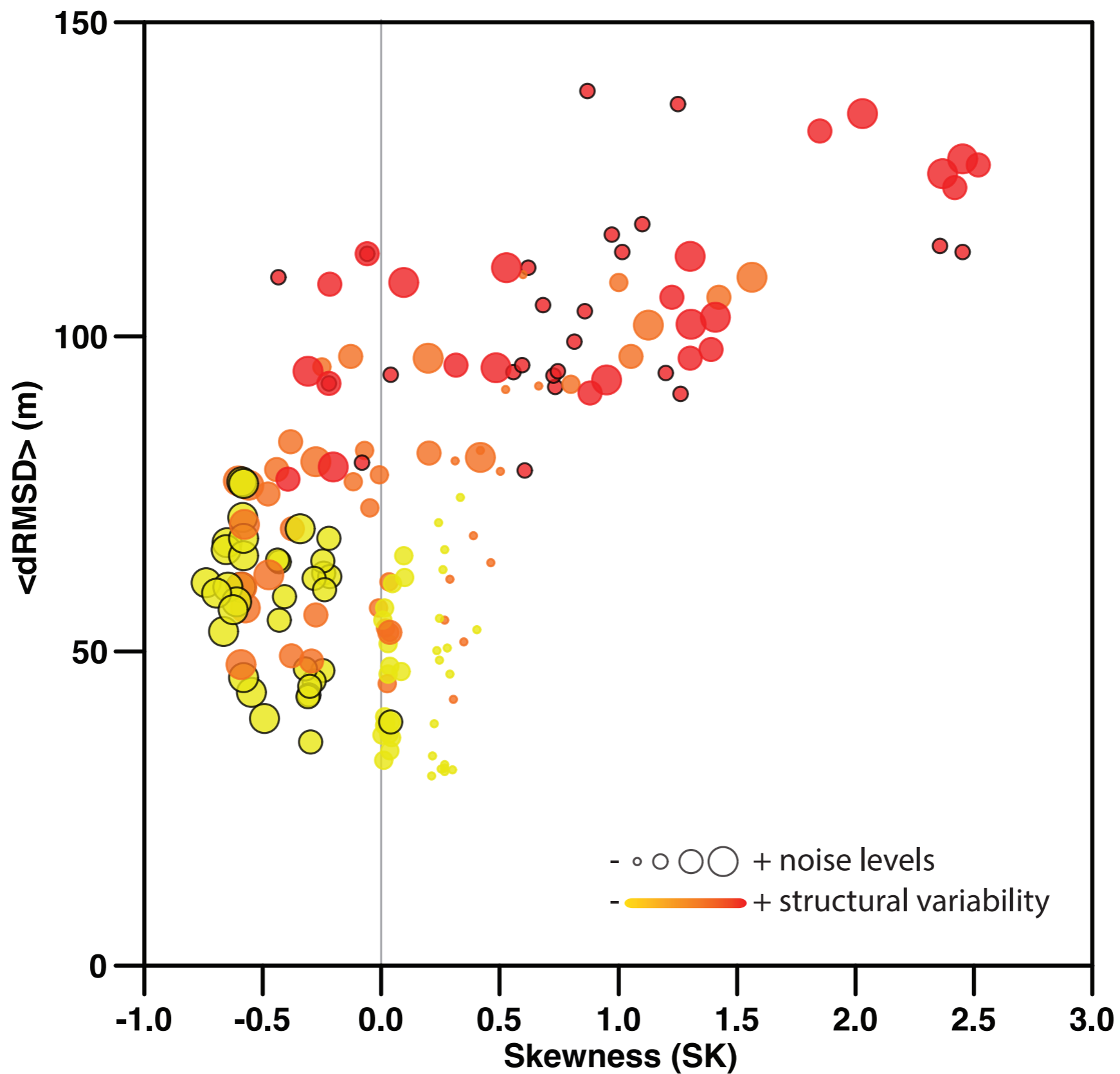
Structural variability is "NOT OK"



Can we predict the accuracy of the models?

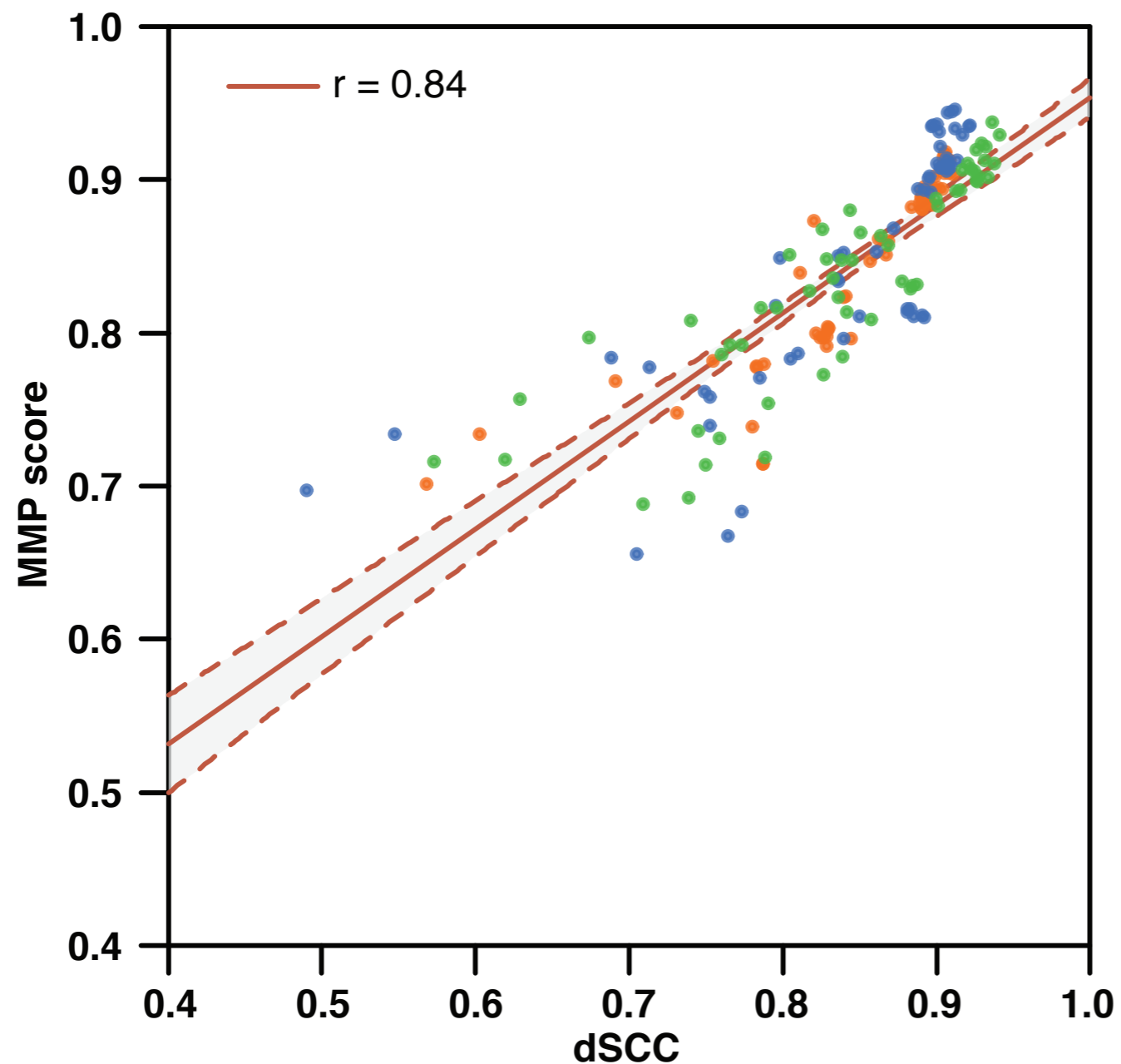


Skewness "side effect"

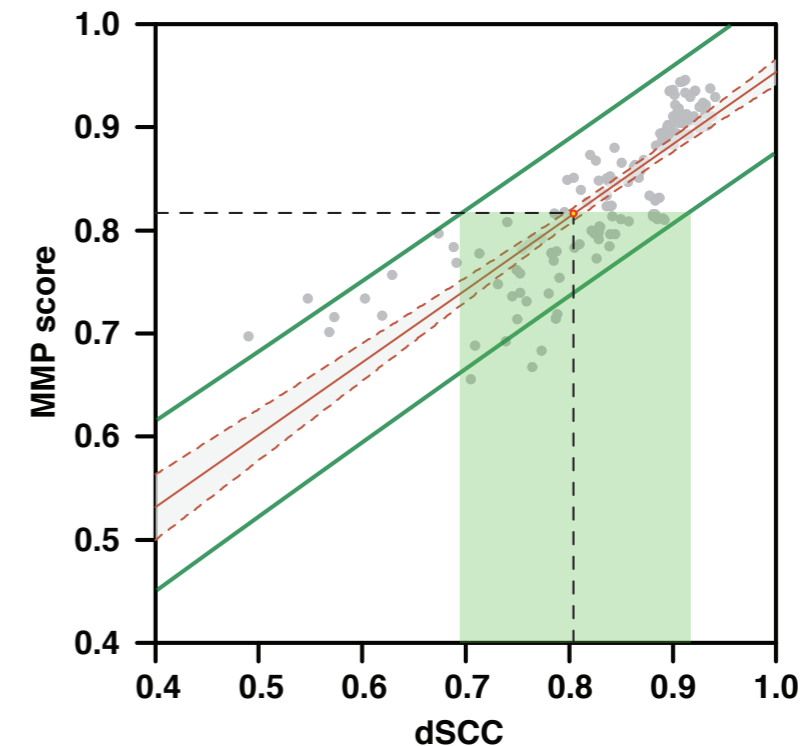
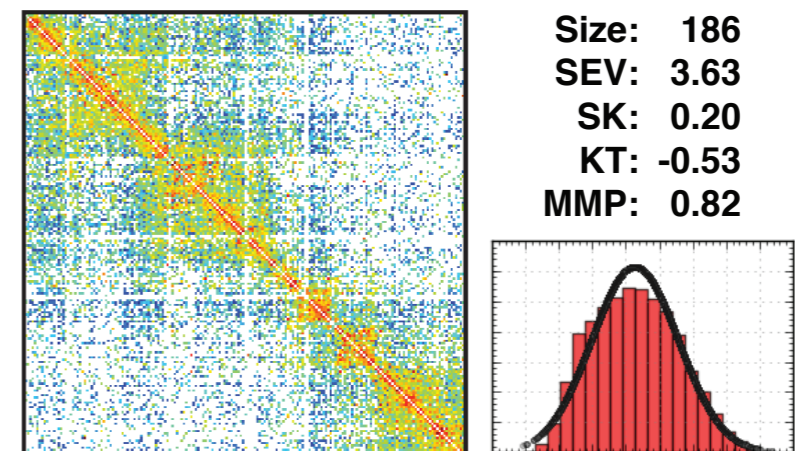


Can we predict the accuracy of the models?

$$\text{MMP} = -0.0002 * \text{Size} + 0.0335 * \text{SK} - 0.0229 * \text{KU} + 0.0069 * \text{SEV} + 0.8126$$



Human Chr1:120,640,000-128,040,000



Higher-res is "good"

put your \$\$ in sequencing

Noise is "OK"

no need to worry much

Structural variability is "NOT OK"

homogenize your cell population!

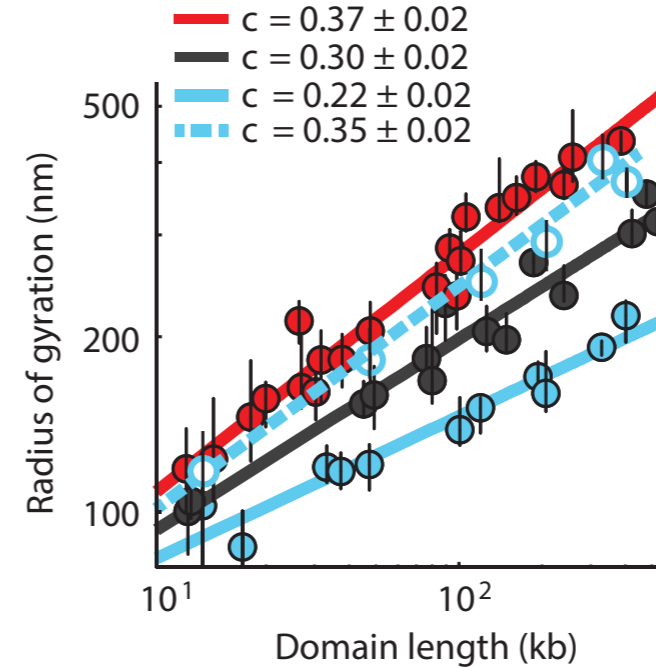
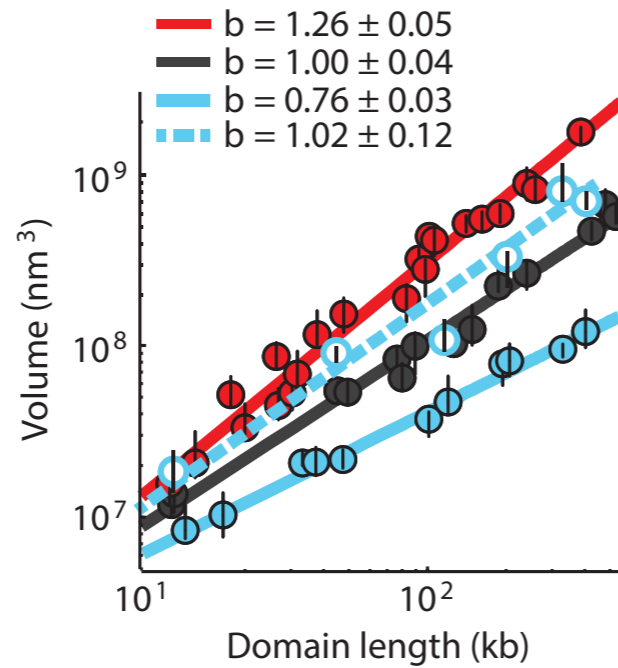
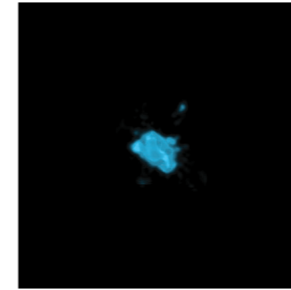
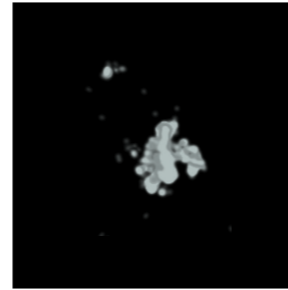
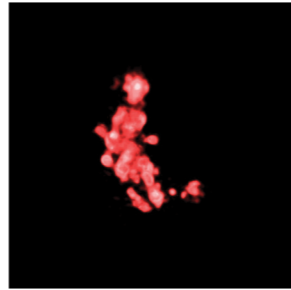
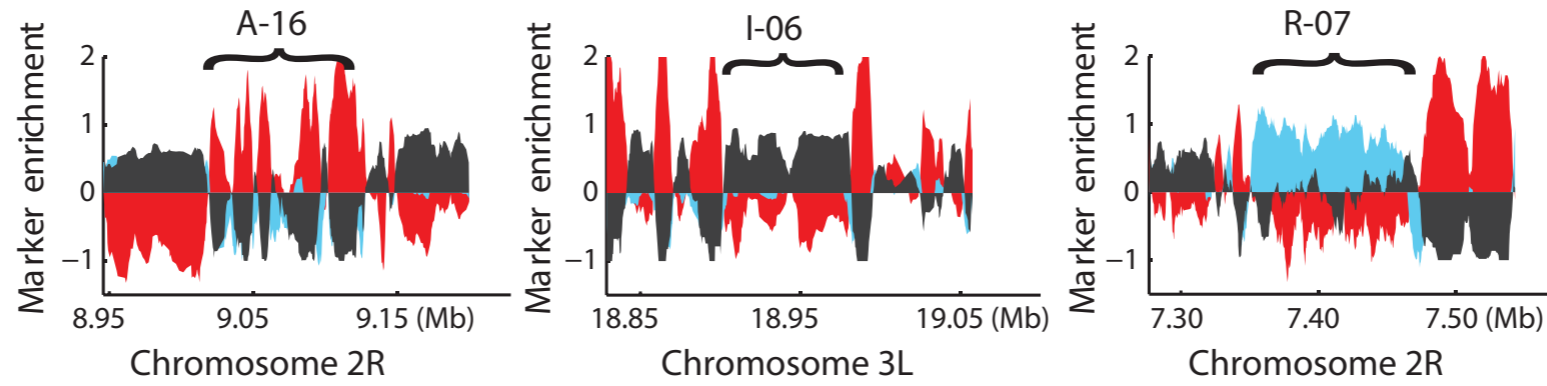
...but we can differentiate between noise and structural variability

and we can a priori predict the accuracy of the models

But... what about direct validation of models?

Model accuracy

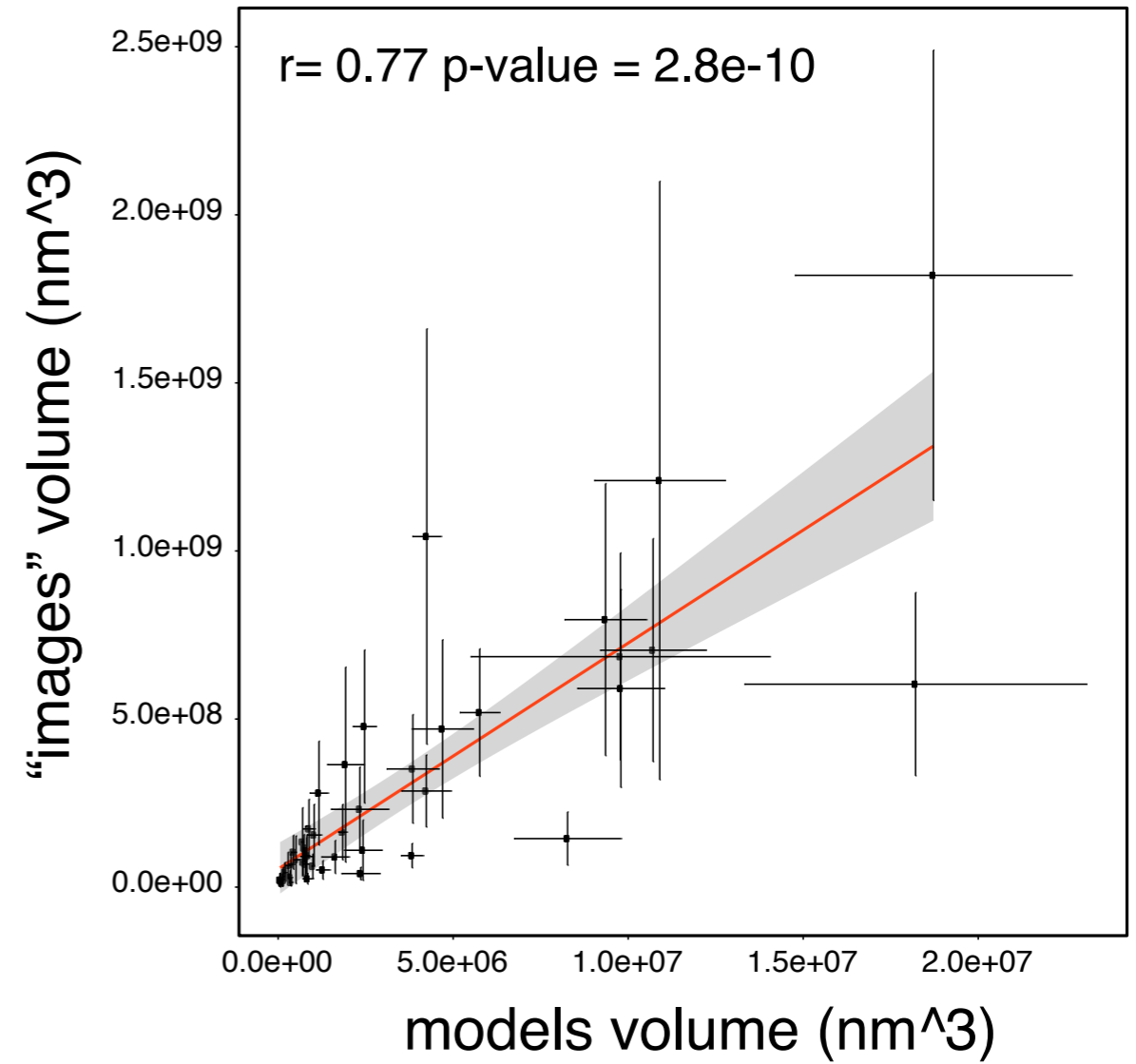
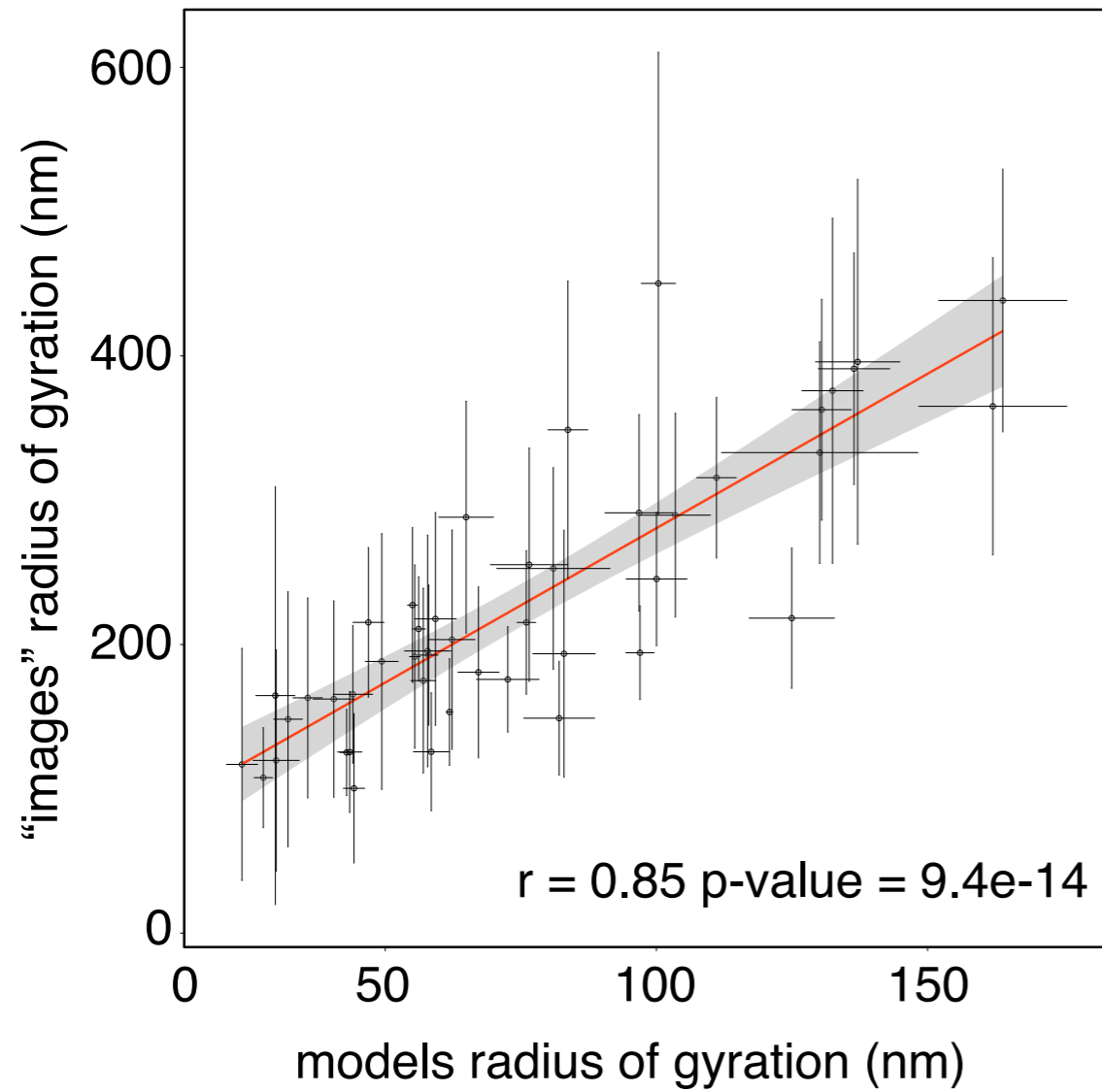
Boettiger, A. N., et al. (2016). *Nature*, 529, 418–422.



● Active ● Inactive ● Repressed ● Repressed (Ph KD)

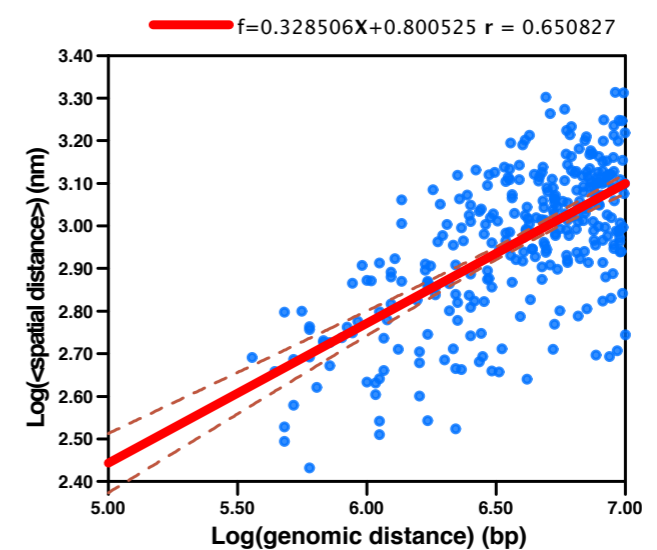
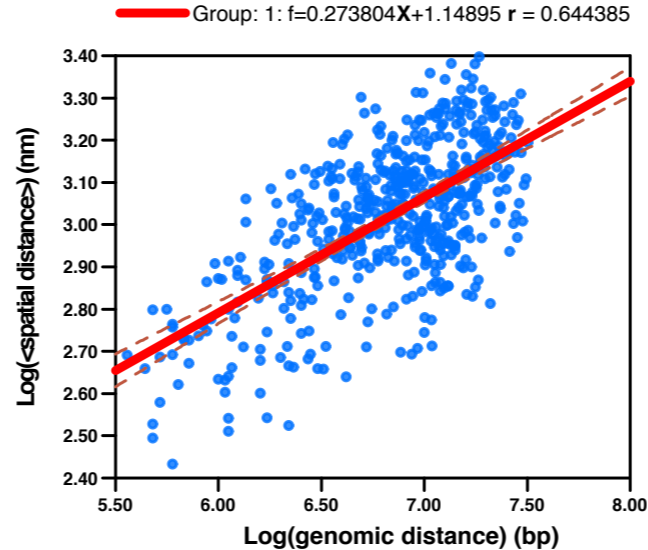
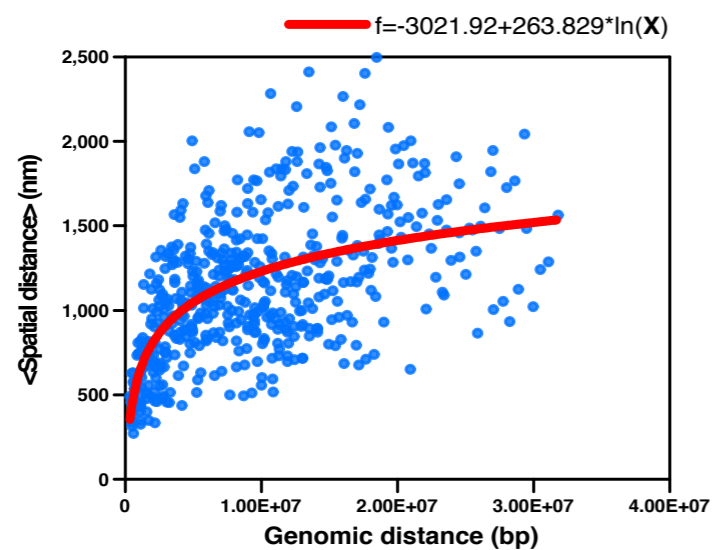
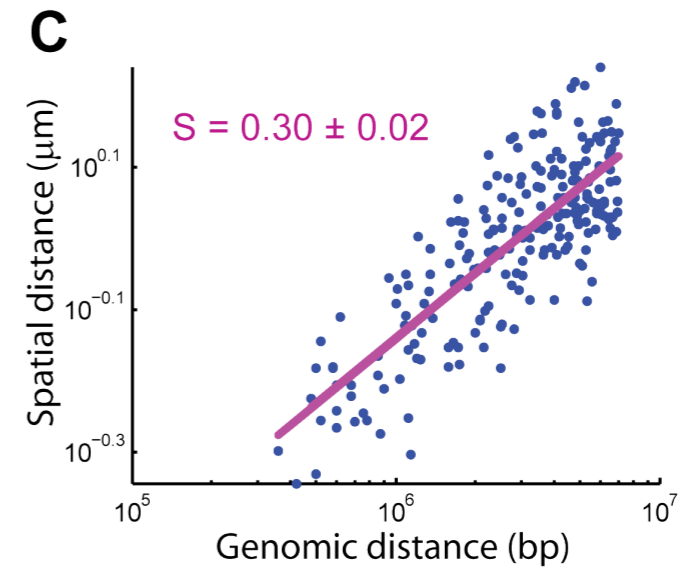
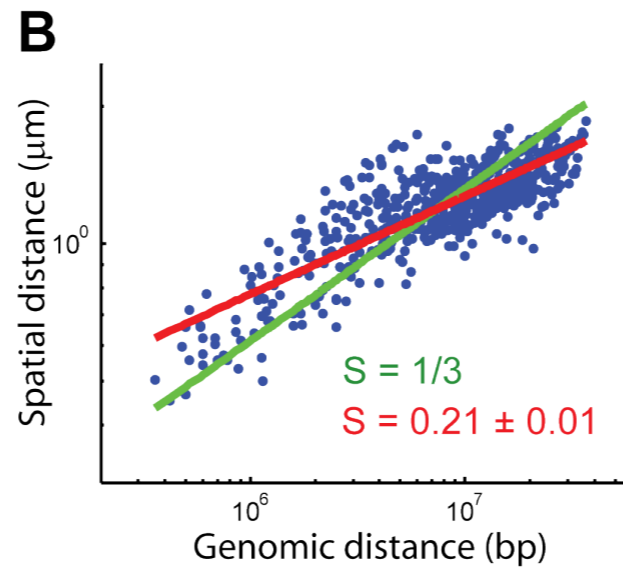
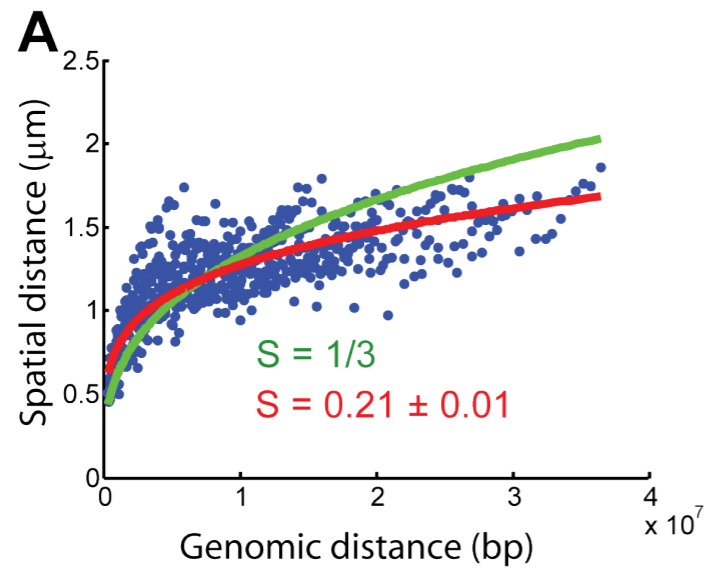
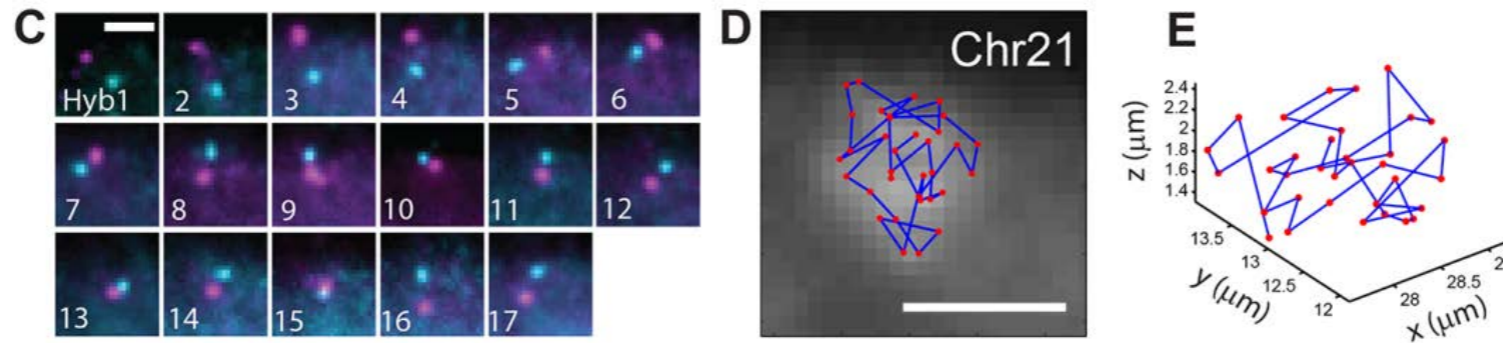
Model accuracy (fly@2Kb)

Boettiger, A. N., et al. (2016). Nature, 529, 418–422.



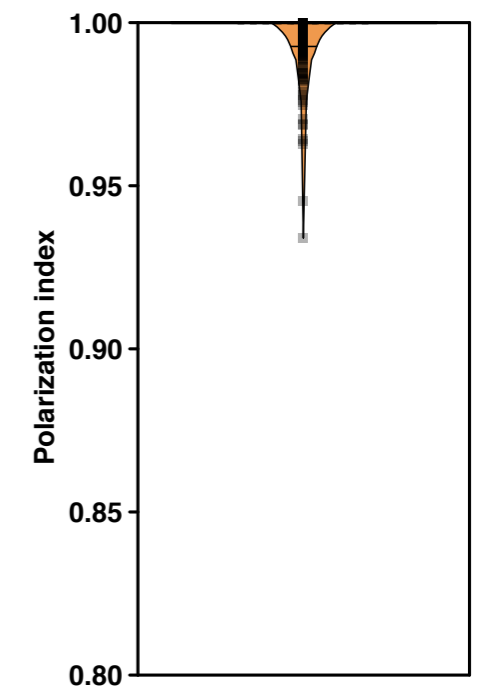
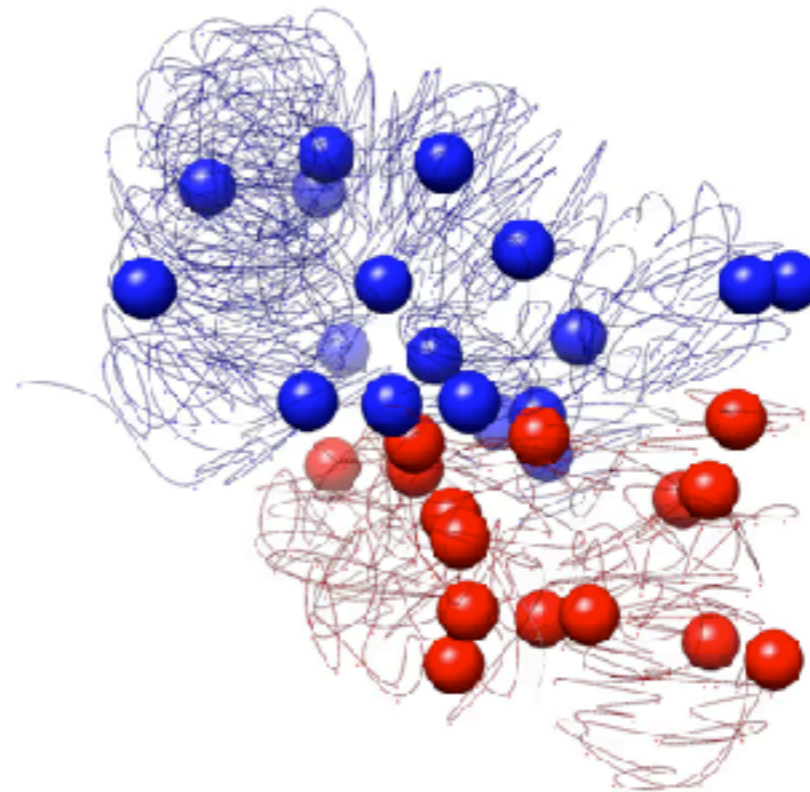
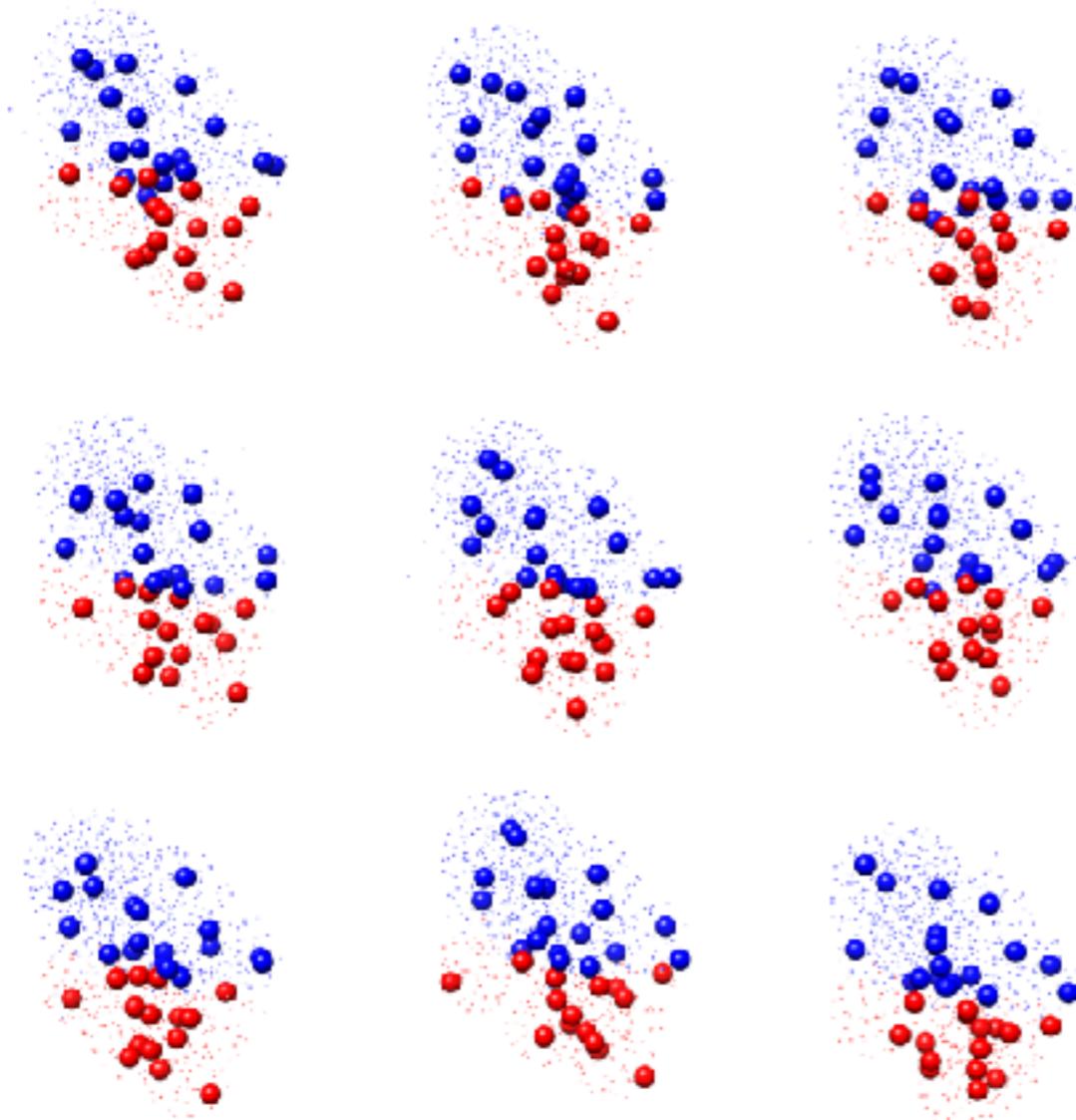
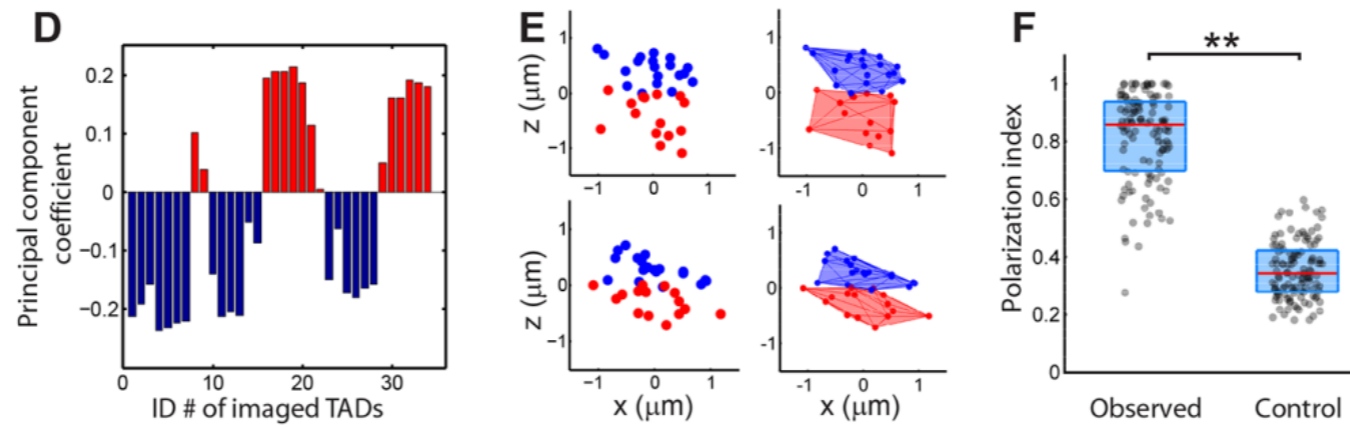
Model accuracy (Human Chr21@40Kb)

Wang, S., et al. (2016). Science 353, 598–602.



Model accuracy (Human Chr21@40Kb)

Wang, S., et al. (2016). *Science* 353(6299), 598–602.



<http://marciuslab.org>
<http://3DGenomes.org>

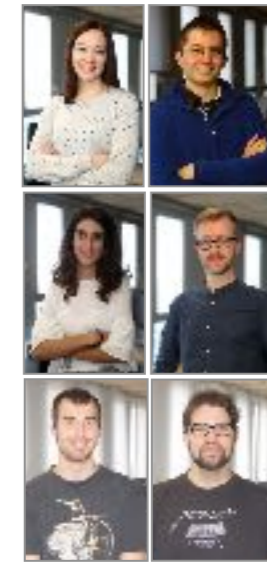
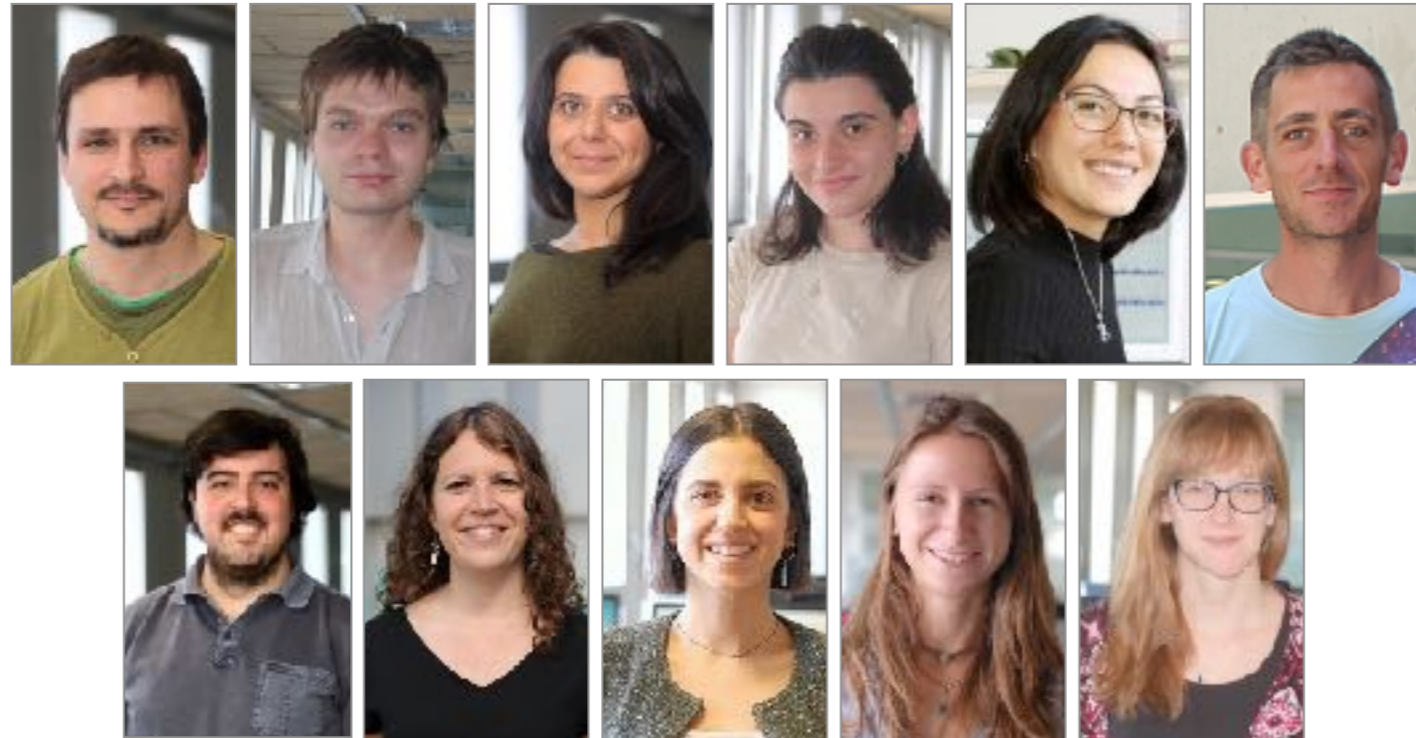


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