3DGenomics

Marc A. Marti-Renom Structural Genomics Group CNAG-CRG



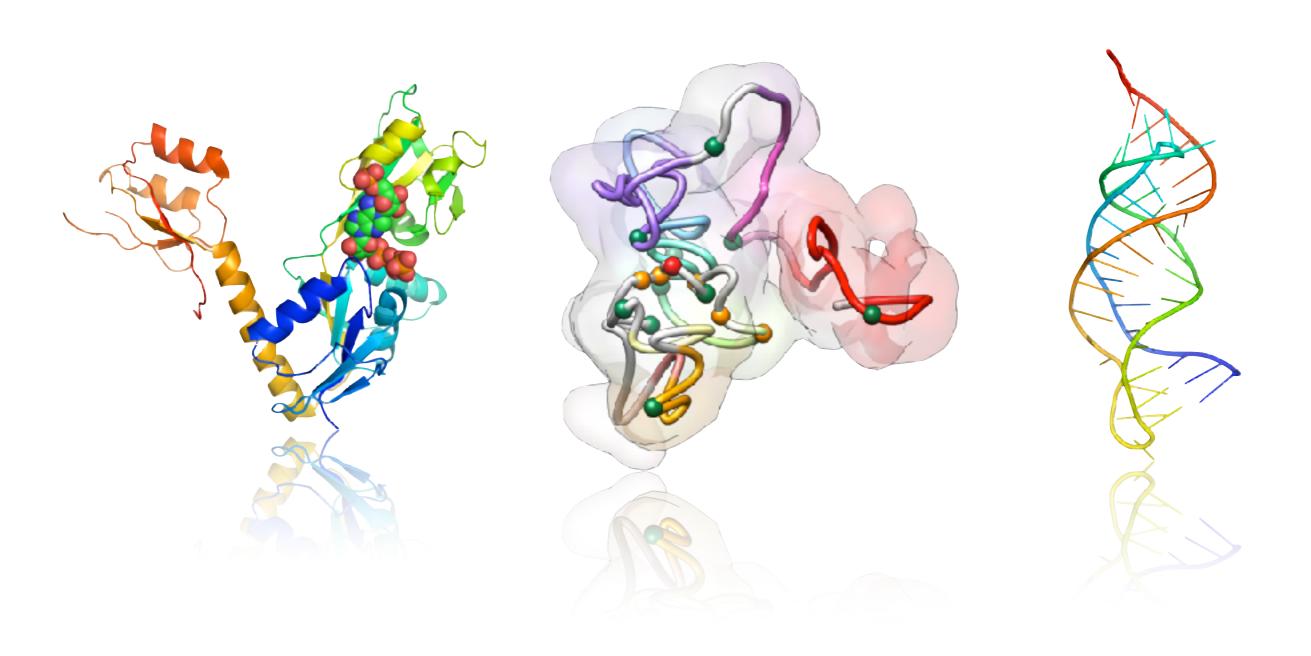






Structural Genomics Group

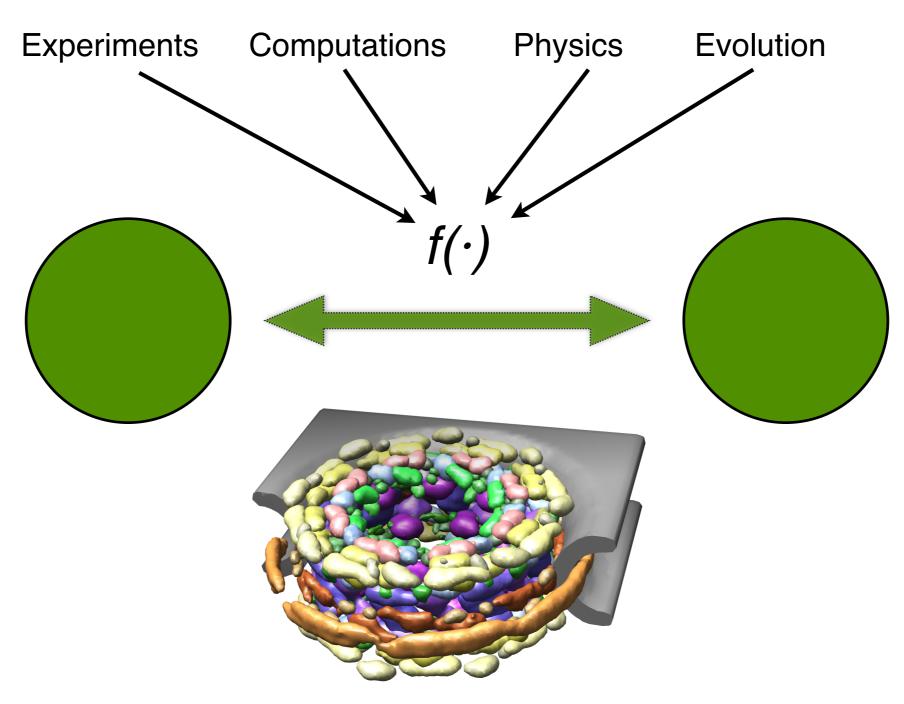
http://www.marciuslab.org





Integrative Modeling Platform

http://www.integrativemodeling.org

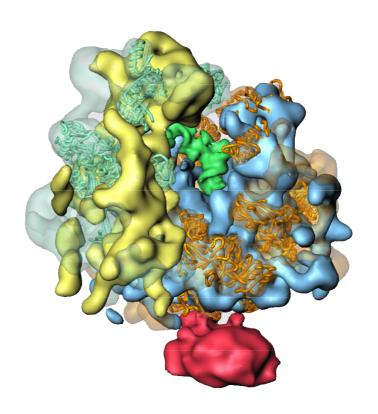


From: Russel, D. et al. PLOS Biology 10, e1001244 (2012).

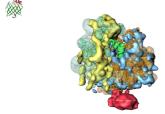


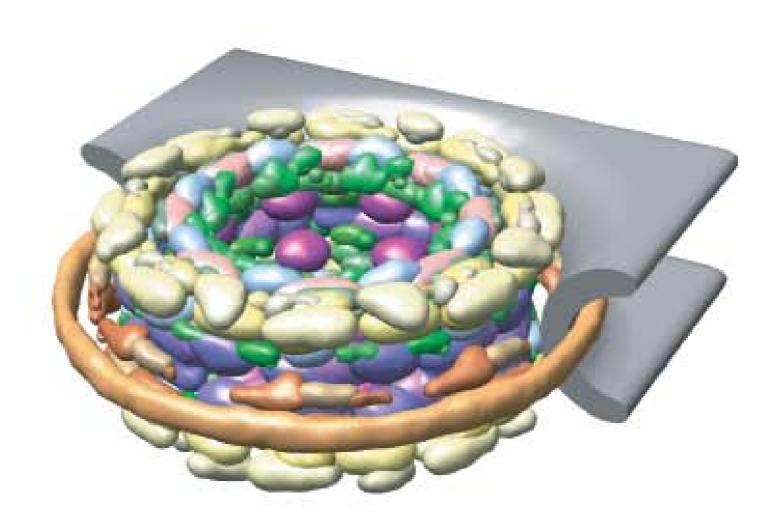
Data Integration





Data Integration



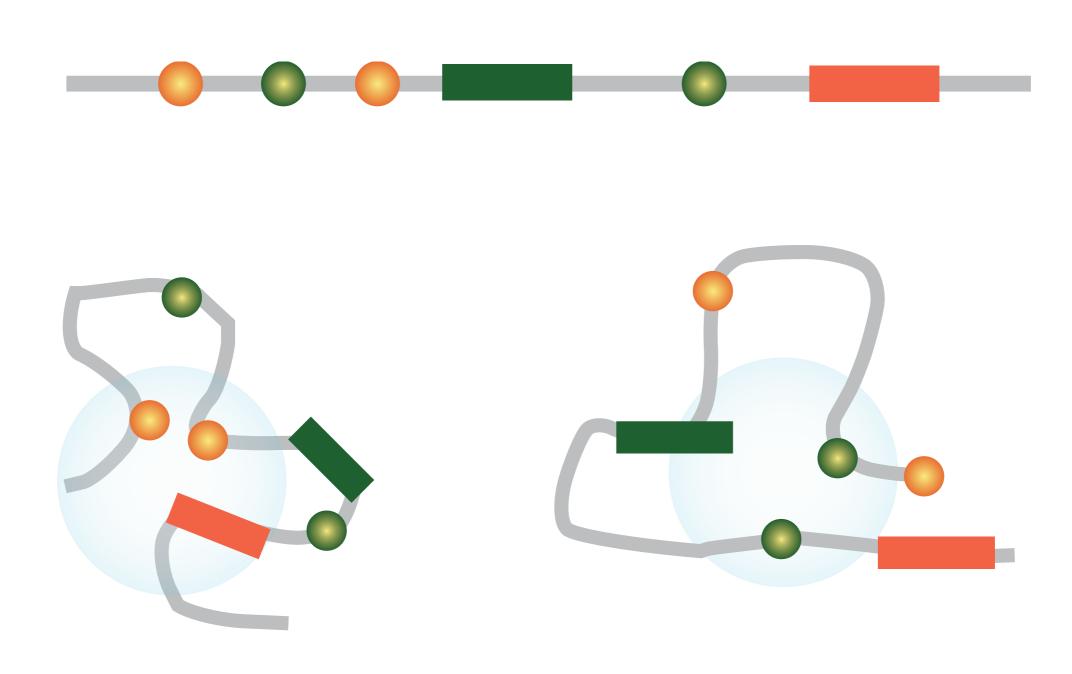




Data Integration



Complex genome organization



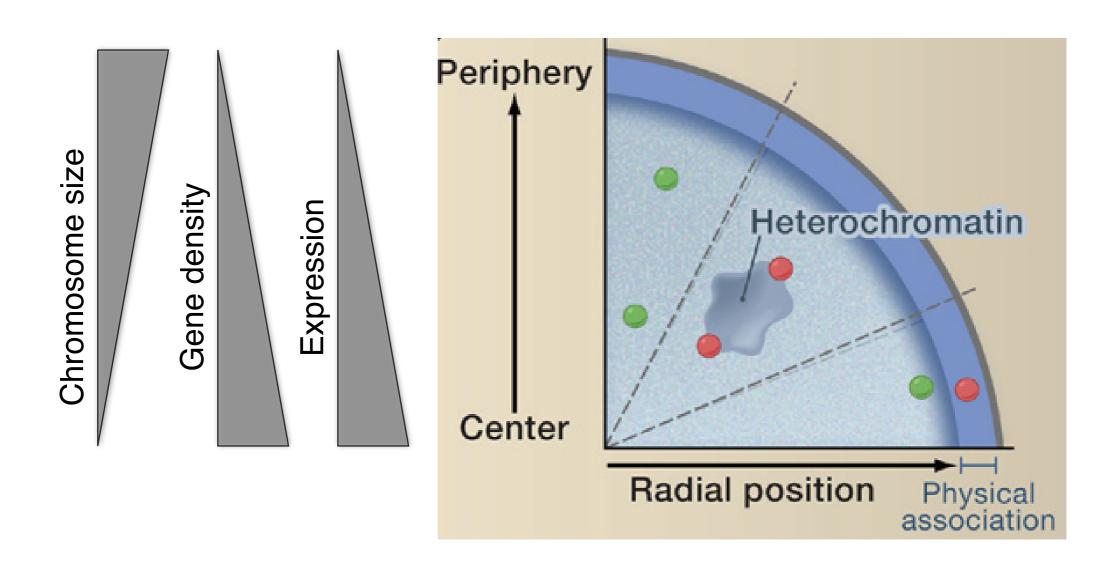
Resolution Gap

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

Knowl	ledge								
******					IDM			5 11 8 X 12 15 6 10 5 13 7 2 16 9 7 18	
10 ⁰		10 ³			10 ⁶			DNA length 10 ⁹	nt
								Volume	1
10 ⁻⁹	10 ⁻⁶		10 ⁻³		10°			10 ³	μm³
							Time		
10 ⁻¹⁰	10 ⁻⁸	10 ⁻⁶	10 ⁻⁴	10 ⁻²		10°	10 ²	10 ³	S
								Resolution	
10 ⁻³			10 ⁻²				10 ⁻¹		μ

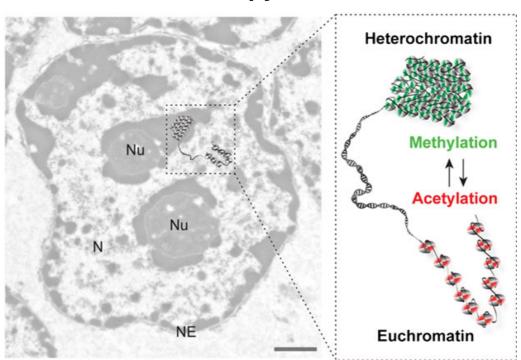
Level I: Radial genome organization

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



Level II: Euchromatin vs heterochromatin

Electron microscopy



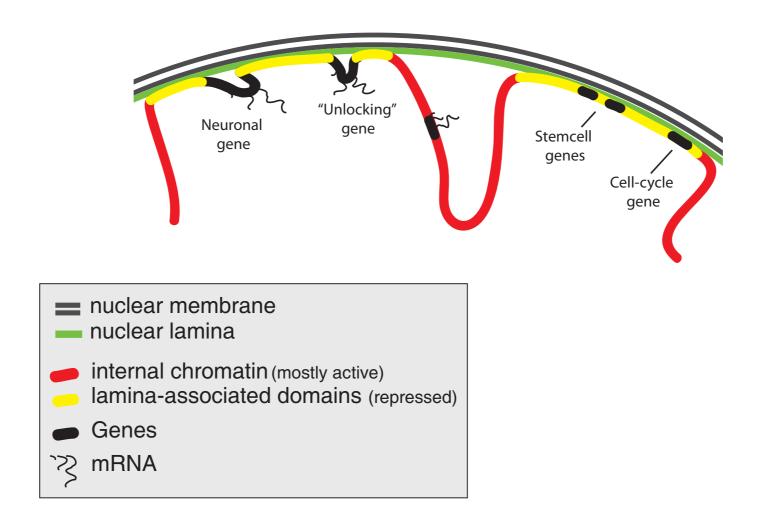
Euchromatin:

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

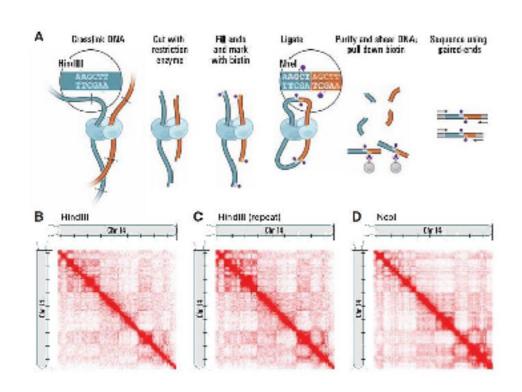


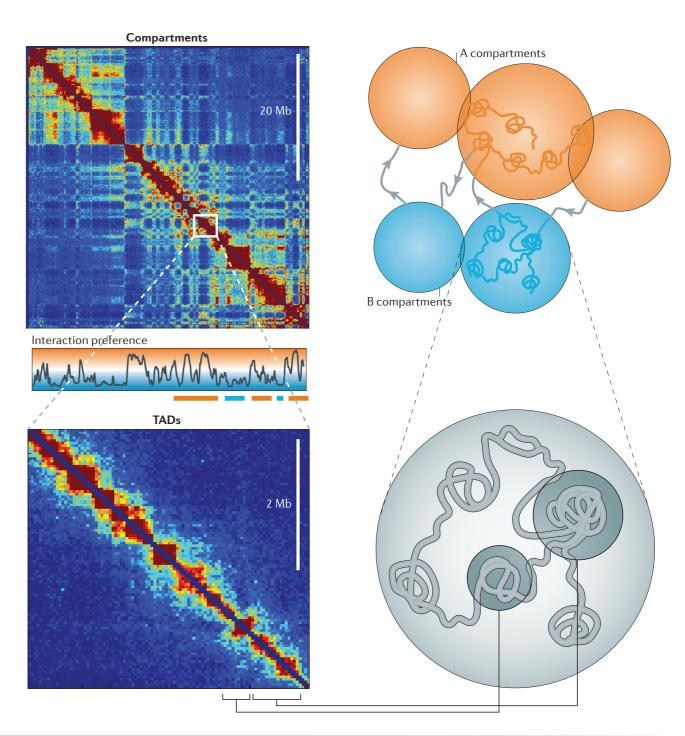
Most genes in Lamina Associated Domains are transcriptionally silent, suggesting that lamina-genome interactions are widely involved in the control of gene expression

Level IV: Higher-order organization

Dekker, J., Marti-Renom, M. A. & Mirny, L. A. Exploring the three-dimensional organization of genomes: interpreting chromatin interaction data.

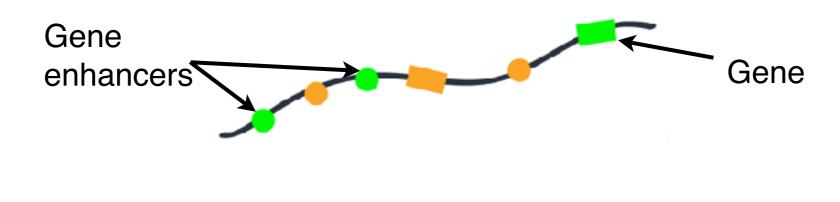
Nat Rev Genet 14, 390–403 (2013).

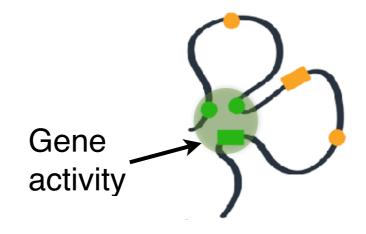






Level V: Chromatin loops





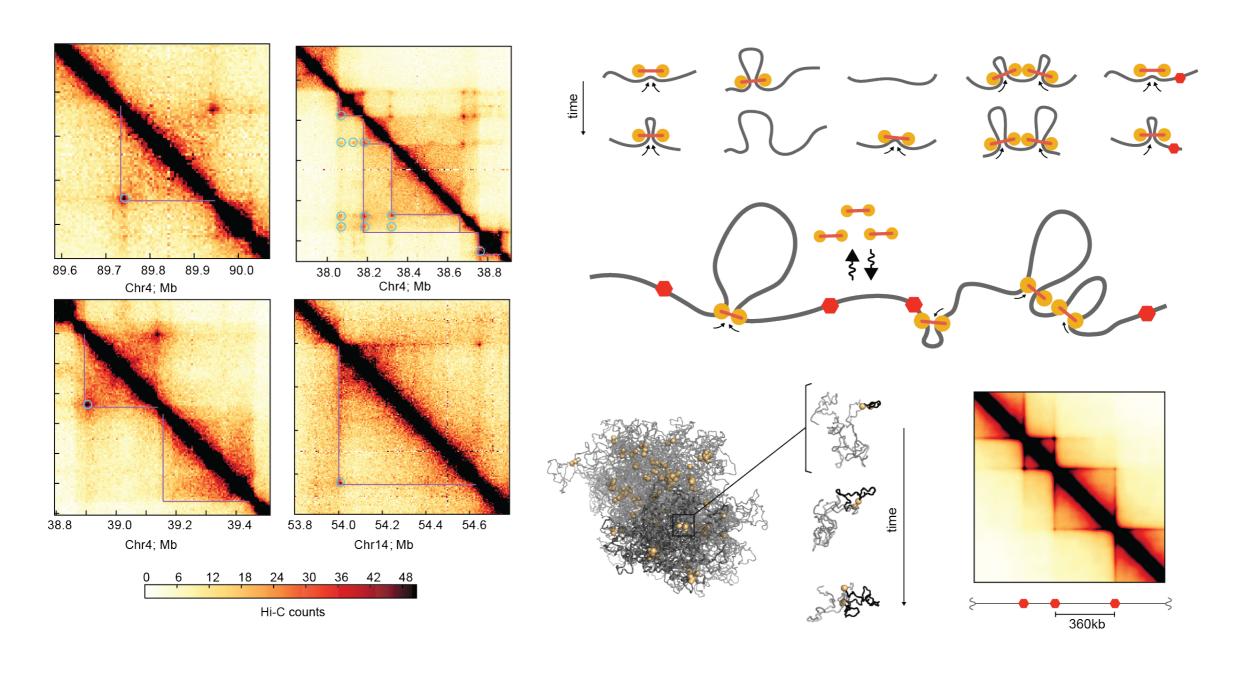
Loops bring distal genomic regions in close proximity to one another

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)

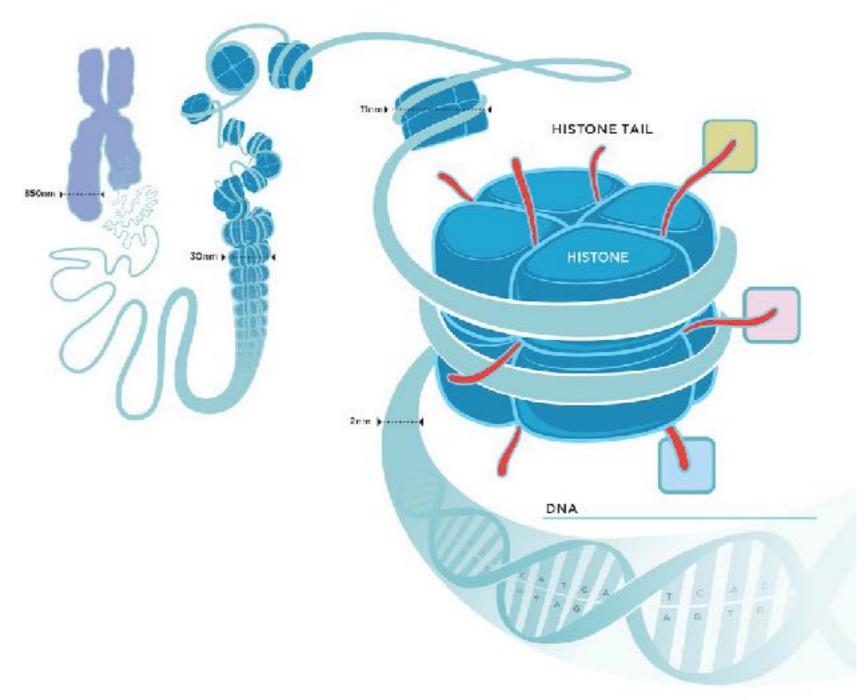
Level V: Loop-extrusion as a driving force

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

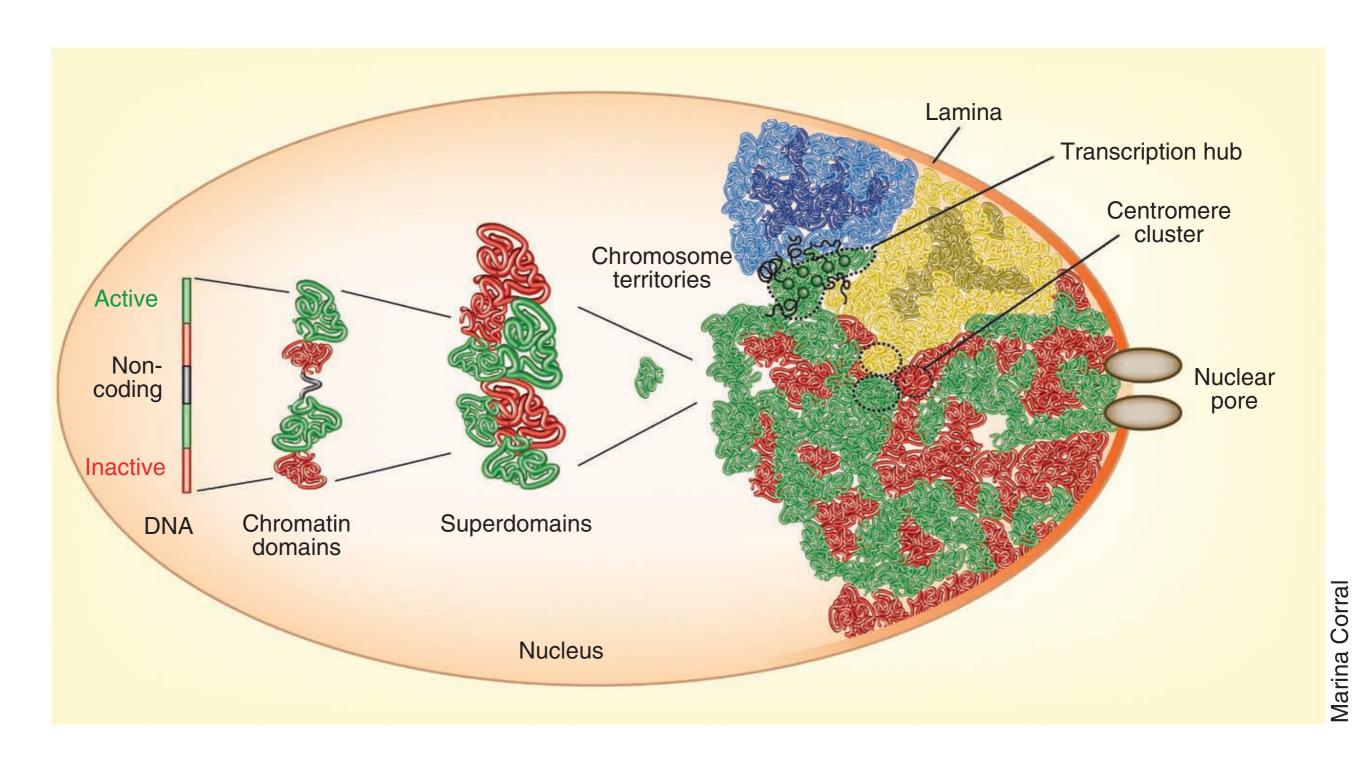
Chromosome Chromatin fibre Nucleosome





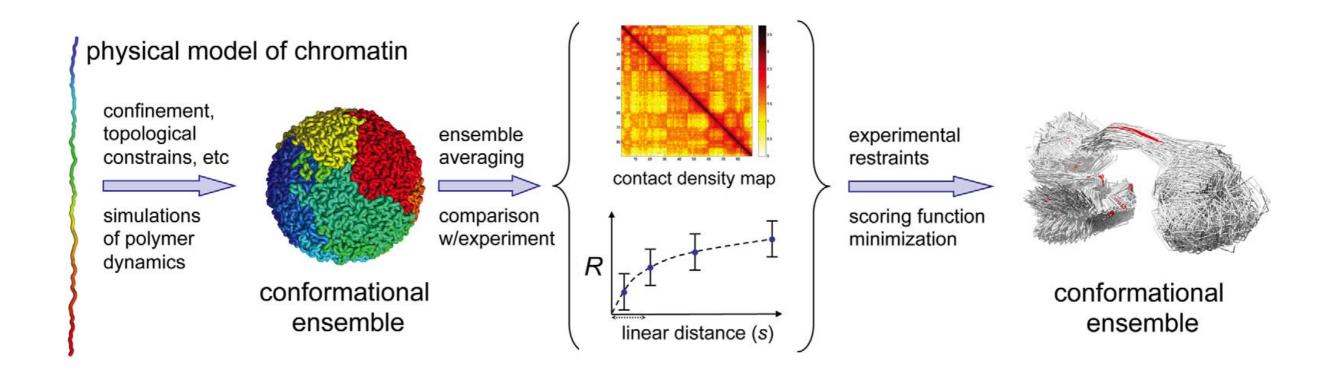
Complex genome organization

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

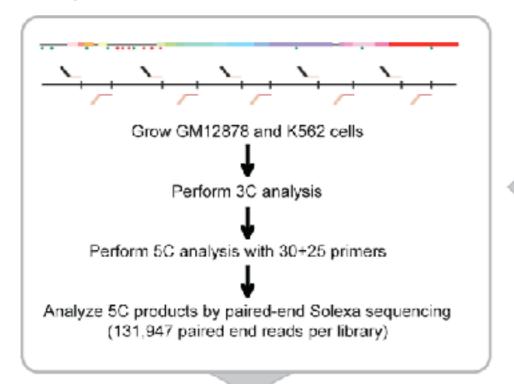


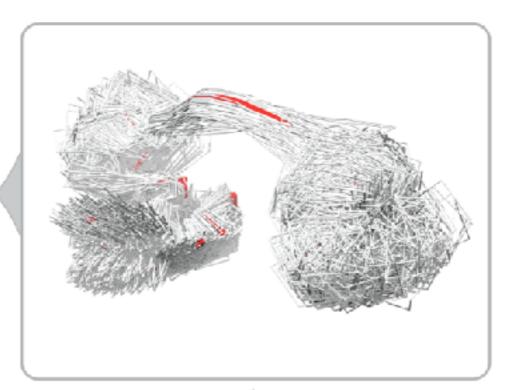
Modeling Genomes

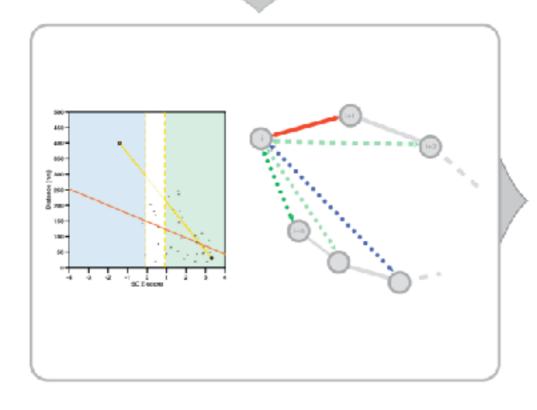
Marti-Renom, M. A. & Mirny, L. A. PLoS Comput Biol 7, e1002125 (2011)

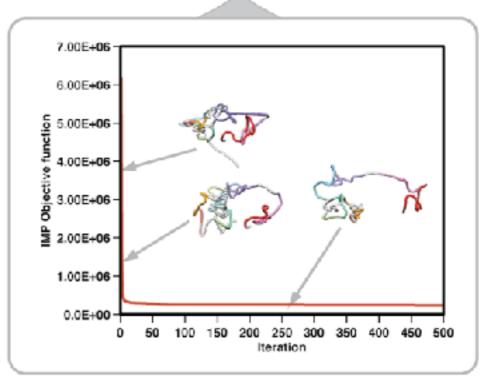


Experiments





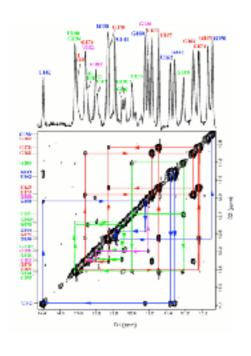




Computation

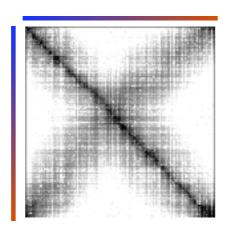






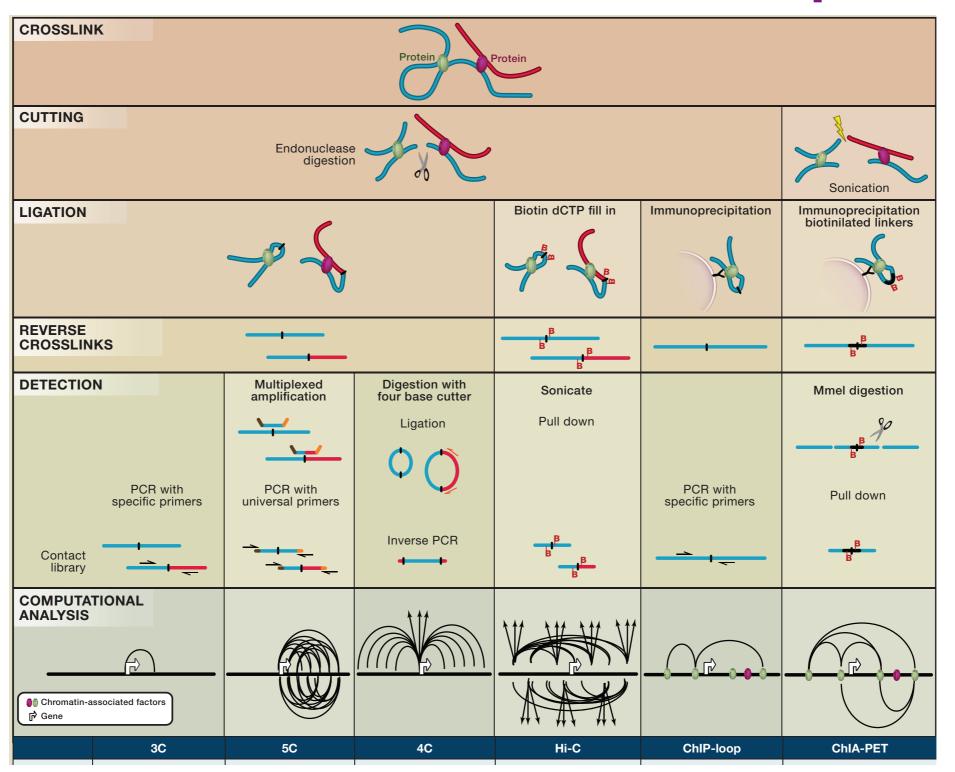
Biomolecular structure determination 2D-NOESY data





Chromosome structure determination 5C data

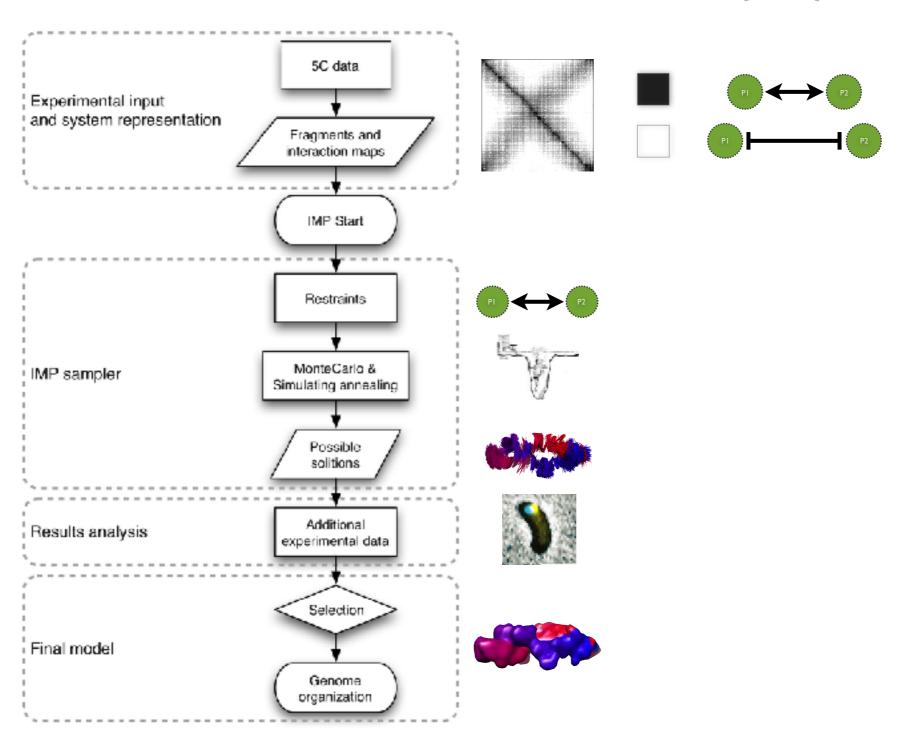
Chromosome Conformation Capture



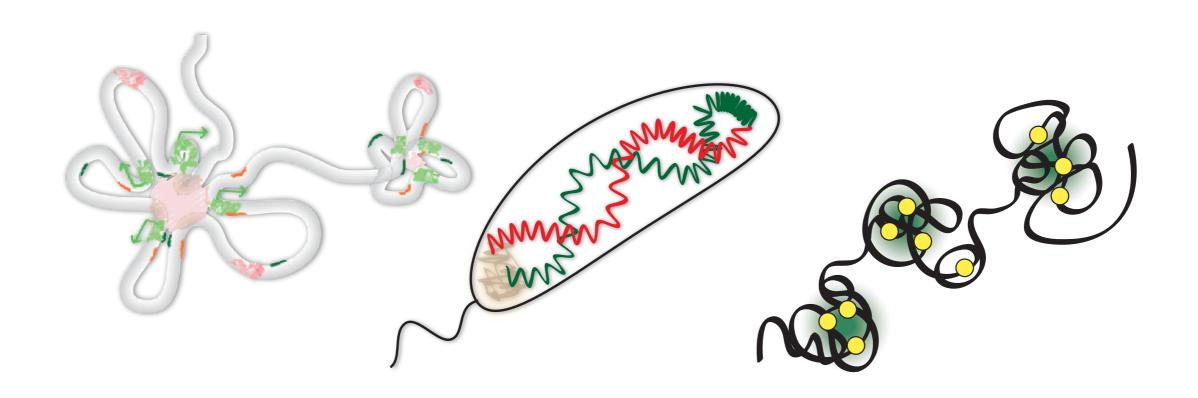


Modeling 3D Genomes

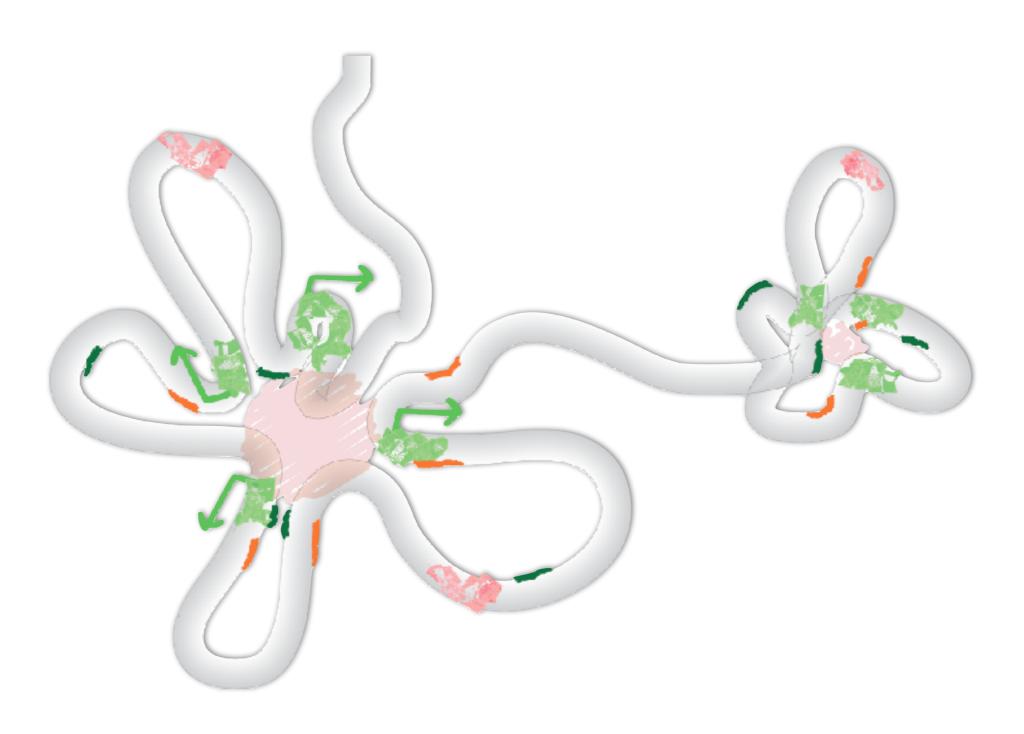
Baù, D. & Marti-Renom, M. A. Methods 58, 300-306 (2012).



Examples...

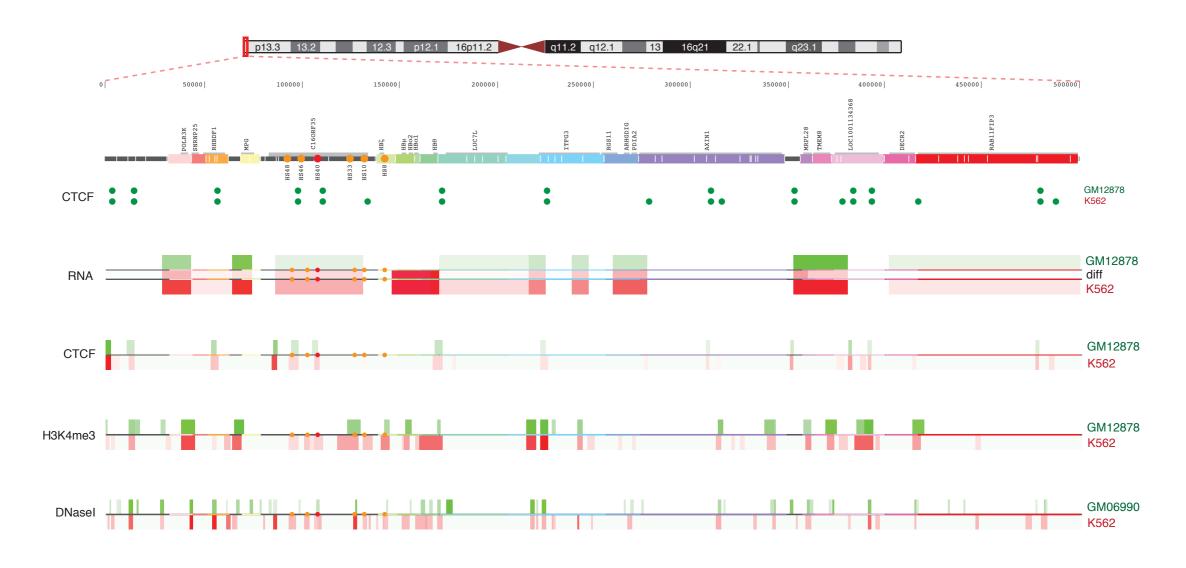


Human α -globin domain



Human α -globin domain

ENm008 genomic structure and environment



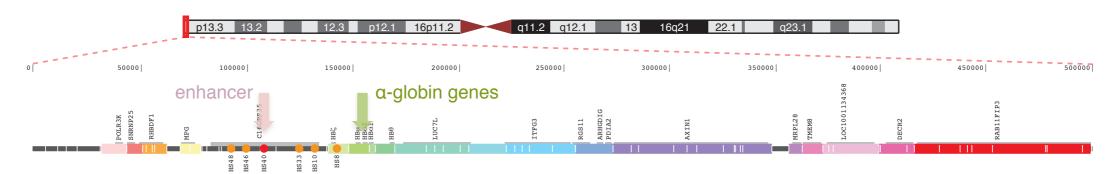
The ENCODE data for ENm008 region was obtained from the UCSC Genome Browser tracks for: RefSeq annotated genes, Affymetrix/CSHL expression data (Gingeras Group at Cold Spring Harbor), Duke/NHGRI DNasel Hypersensitivity data (Crawford Group at Duke University), and Histone Modifications by Broad Institute ChIP-seq (Bernstein Group at Broad Institute of Harvard and MIT).

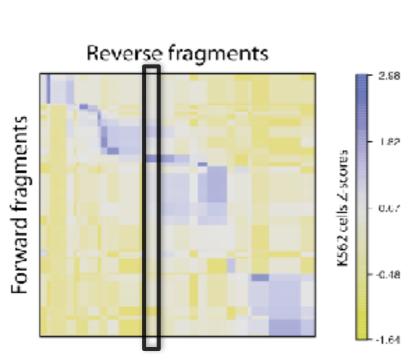
ENCODE Consortium. Nature (2007) vol. 447 (7146) pp. 799-816



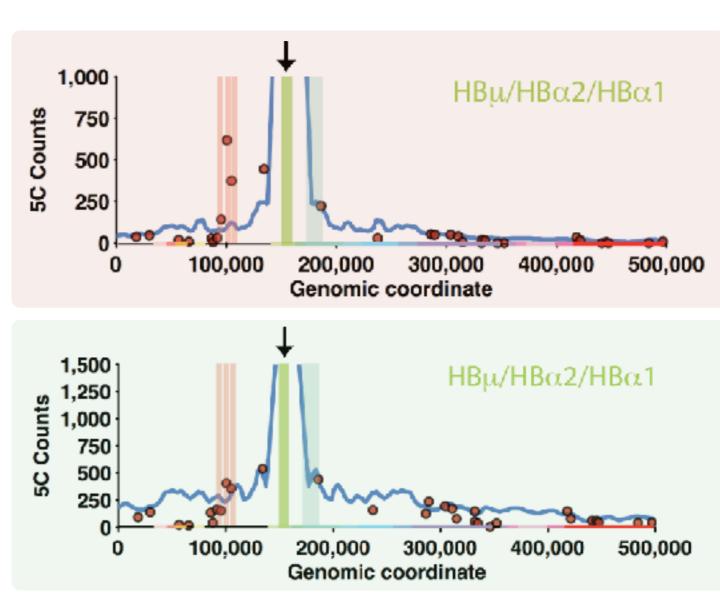
Human α -globin domain

ENm008 genomic structure and environment





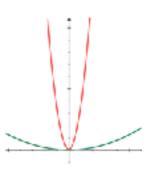
K562 cells: α-globin genes active



Representation

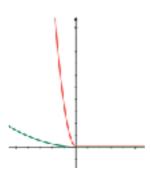
Harmonic

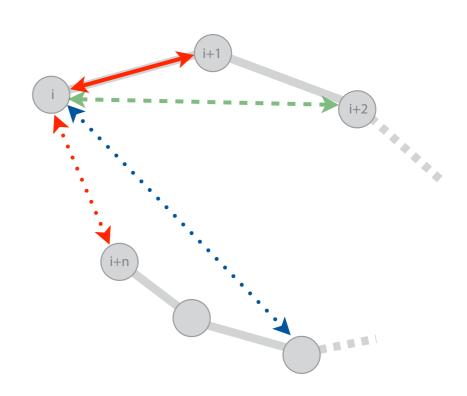
$$H_{i,j} = k \left(d_{i,j} - d_{i,j}^0 \right)^2$$



Harmonic Lower Bound

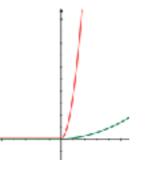
$$\begin{cases} if \ d_{i,j} \le d_{i,j}^{0}; & lbH_{i,j} = k(d_{i,j} - d_{i,j}^{0})^{2} \\ if \ d_{i,j} > d_{i,j}^{0}; & lbH_{i,j} = 0 \end{cases}$$



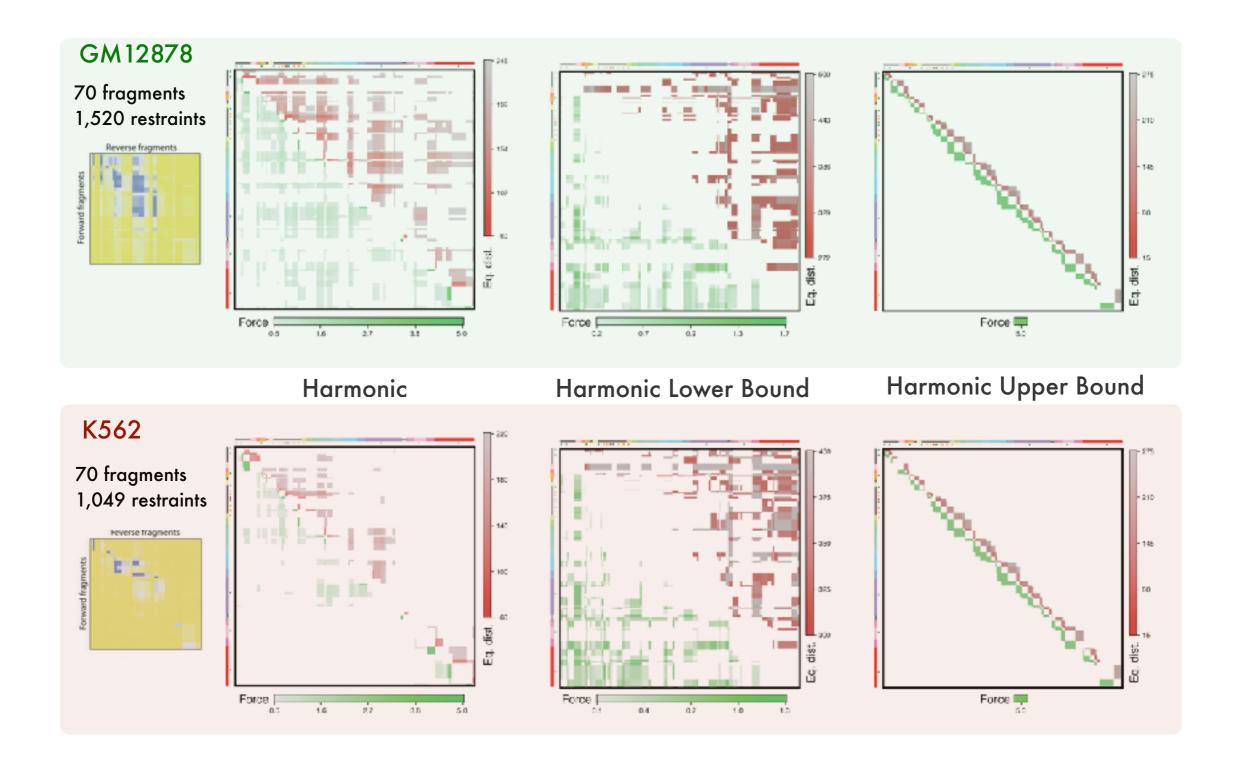


Harmonic Upper Bound

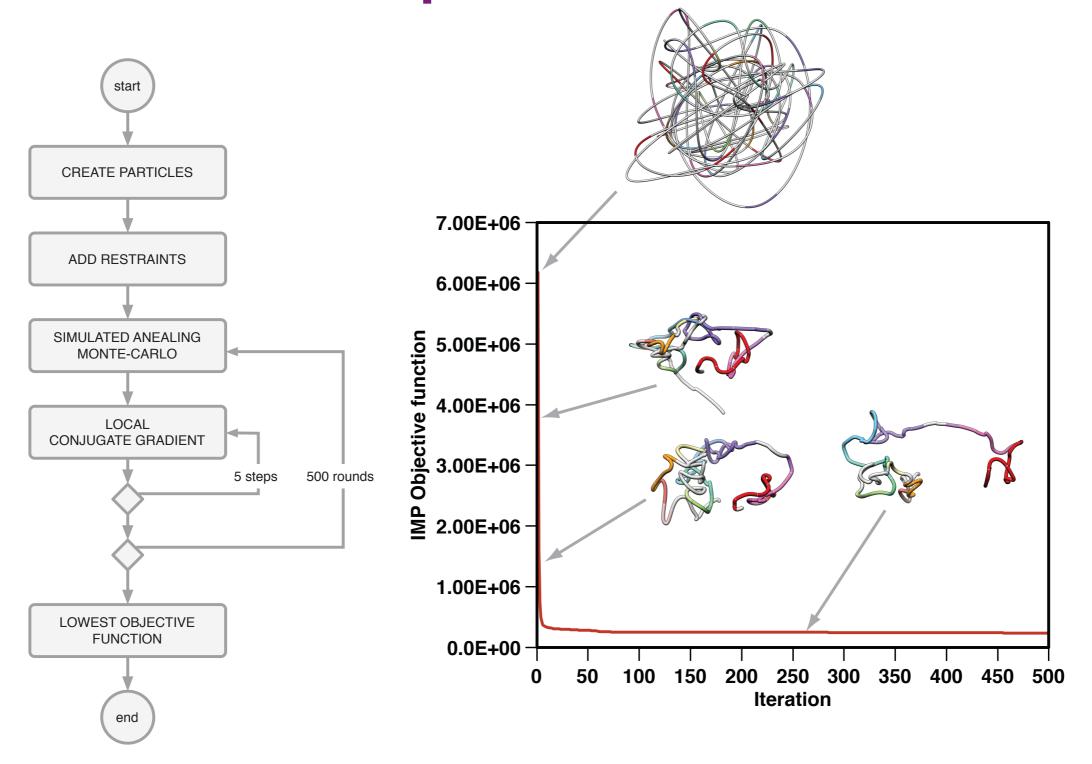
$$\begin{cases} if \ d_{i,j} \ge d_{i,j}^0; & ubH_{i,j} = k(d_{i,j} - d_{i,j}^0)^2 \\ if \ d_{i,j} < d_{i,j}^0; & ubH_{i,j} = 0 \end{cases}$$



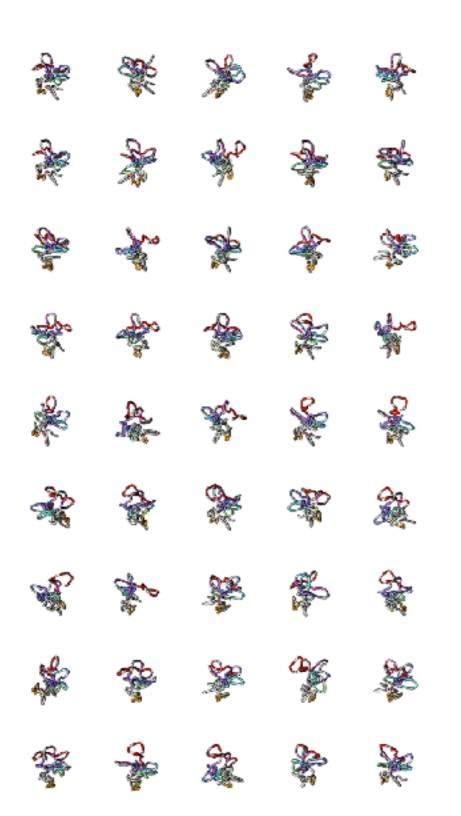
Scoring

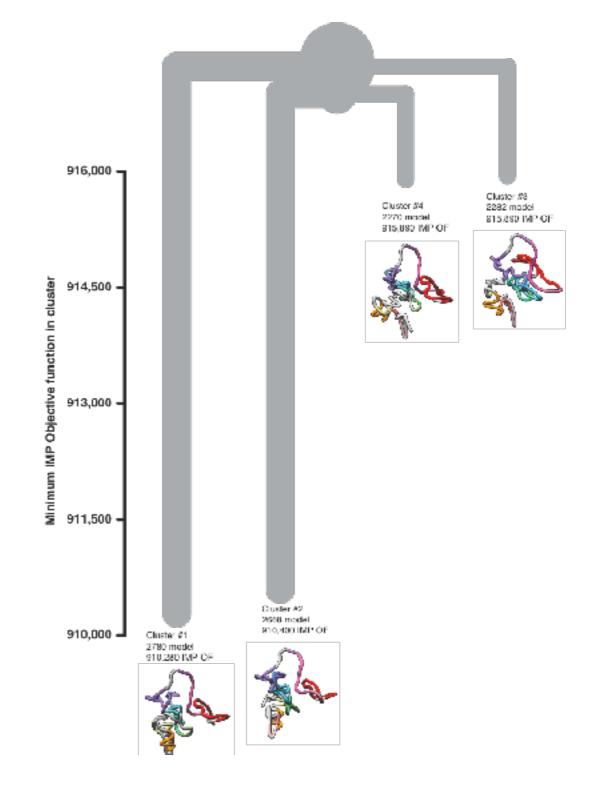


Optimization

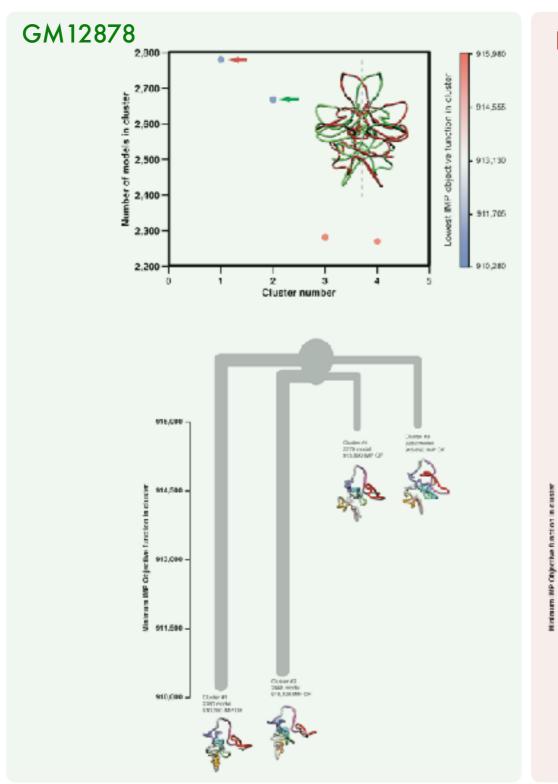


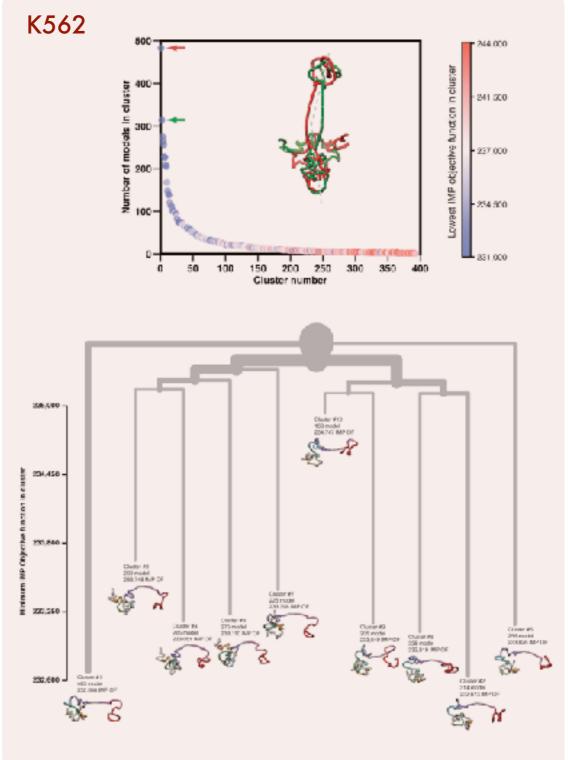
Clustering





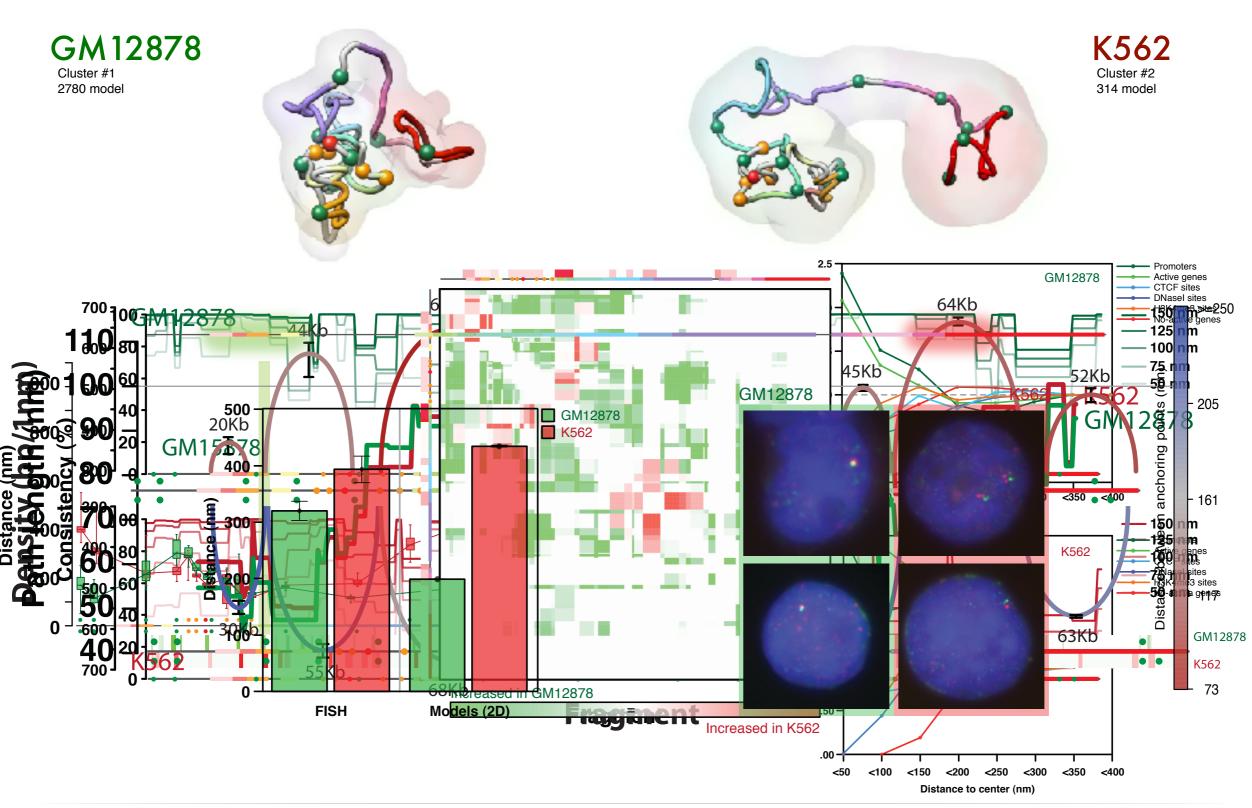
Not just one solution







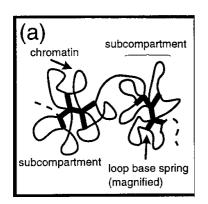
Regulations



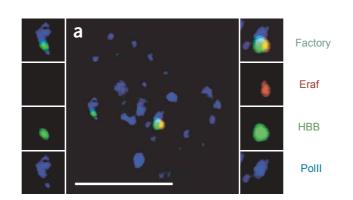


The "Chromatin Globule" model

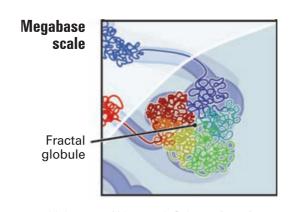








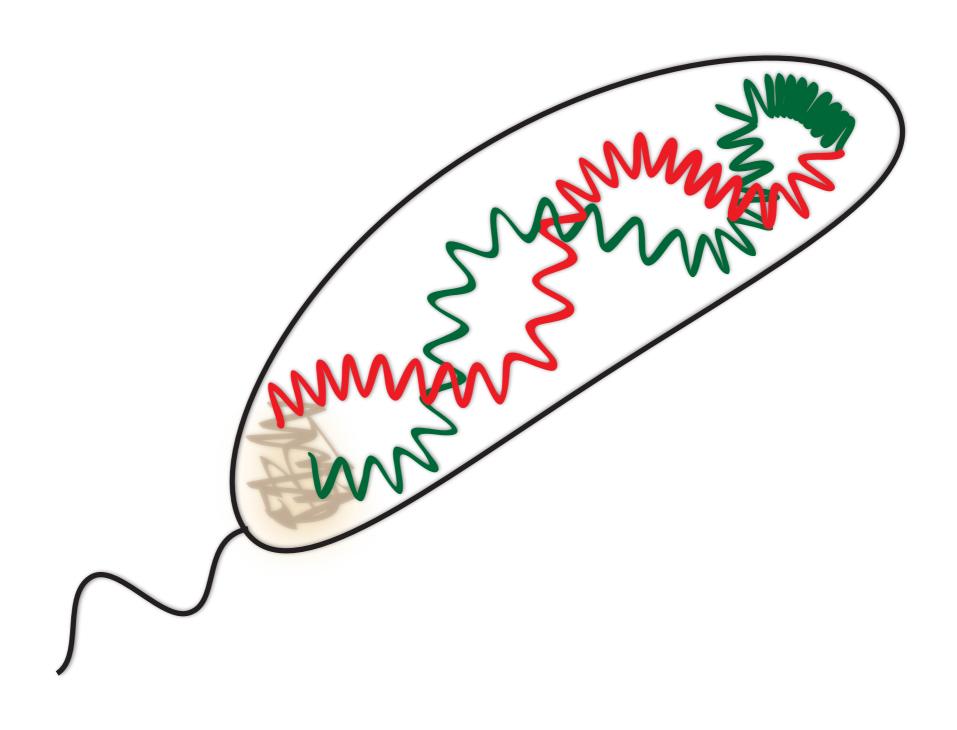
Osborne et al. Nat Genet (2004)



Lieberman-Aiden et al. Science (2009)

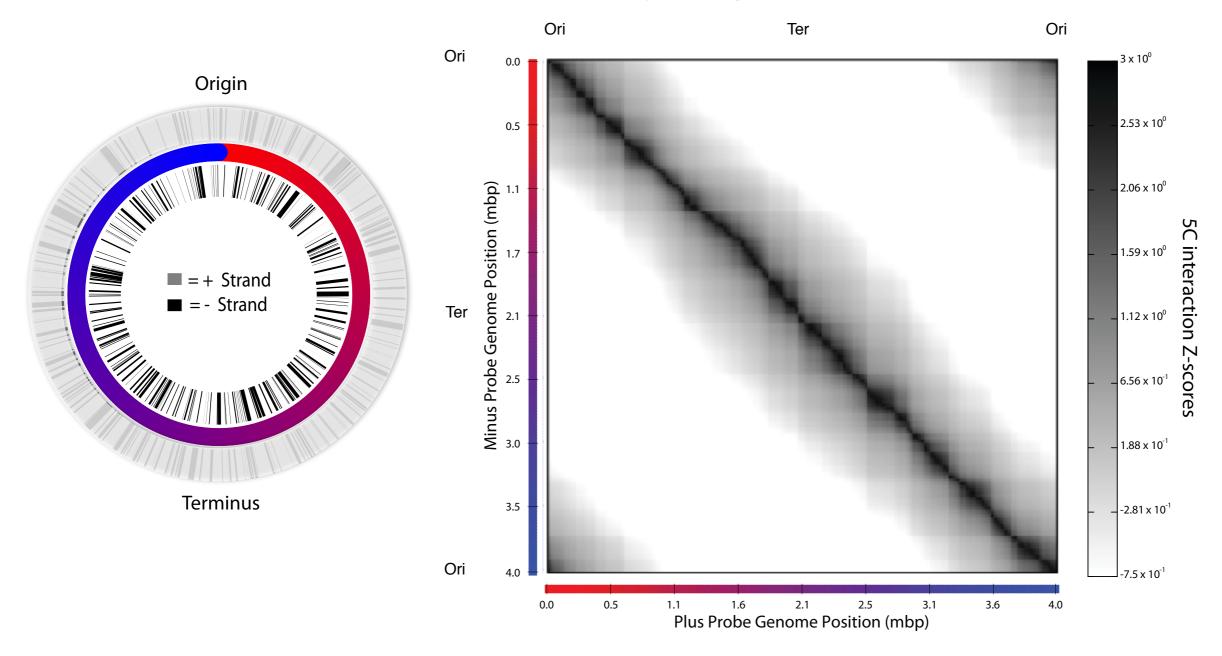
D. Baù et al. Nat Struct Mol Biol (2011) 18:107-14 A. Sanyal et al. Current Opinion in Cell Biology (2011) 23:325-33.

Caulobacter crescentus genome



The 3D architecture of Caulobacter Crescentus

4,016,942 bp & 3,767 genes



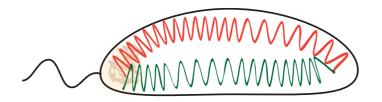
169 5C primers on + strand 170 5C primers on - strand 28,730 chromatin interactions

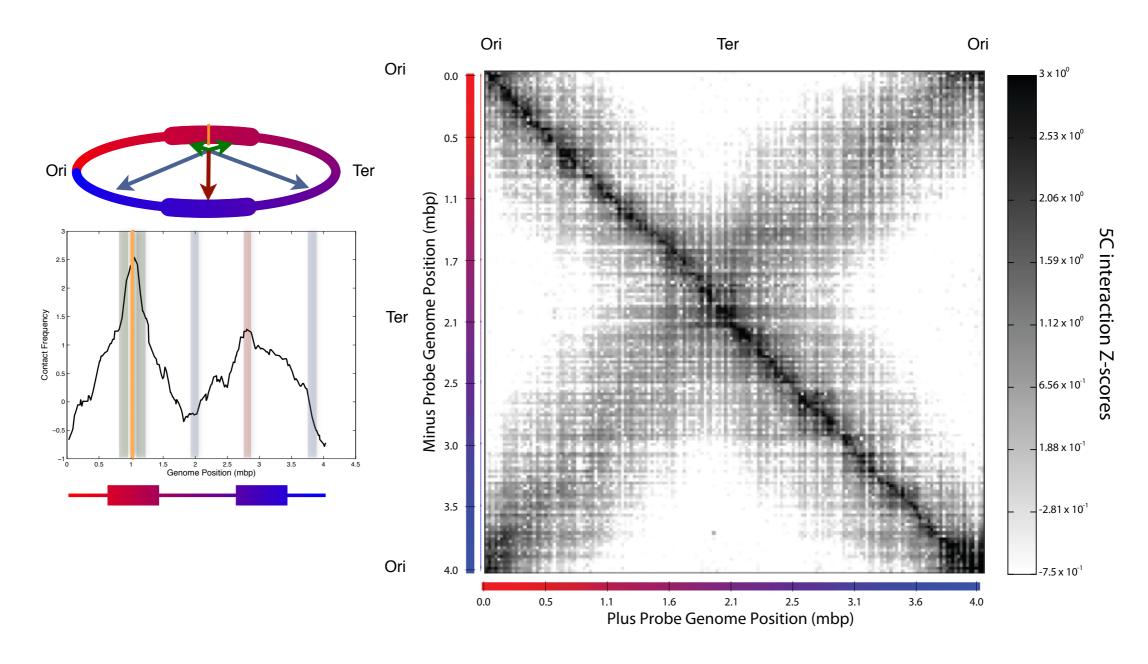




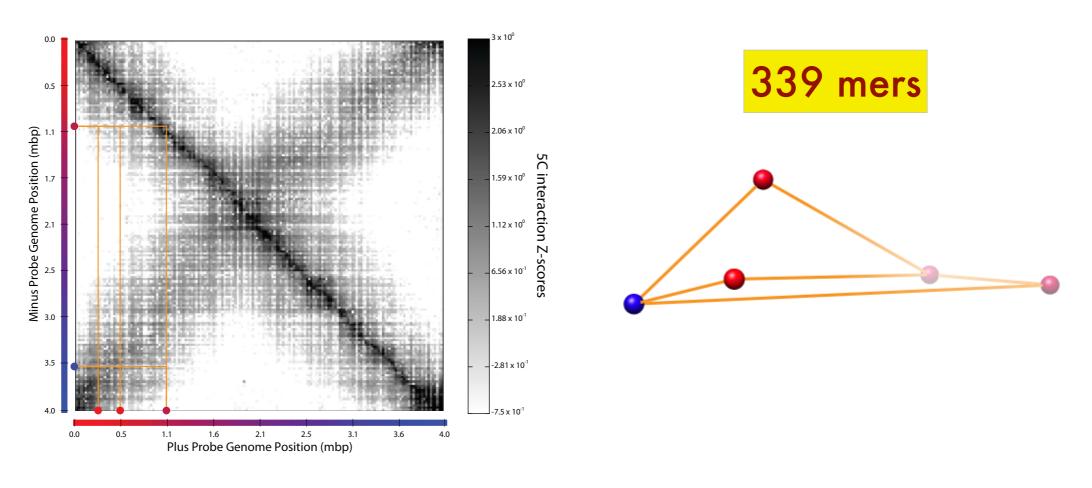
5C interaction matrix

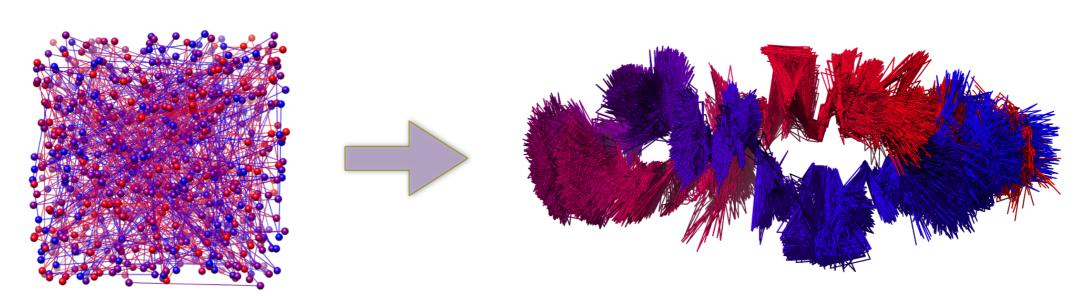
ELLIPSOID for Caulobacter cresentus



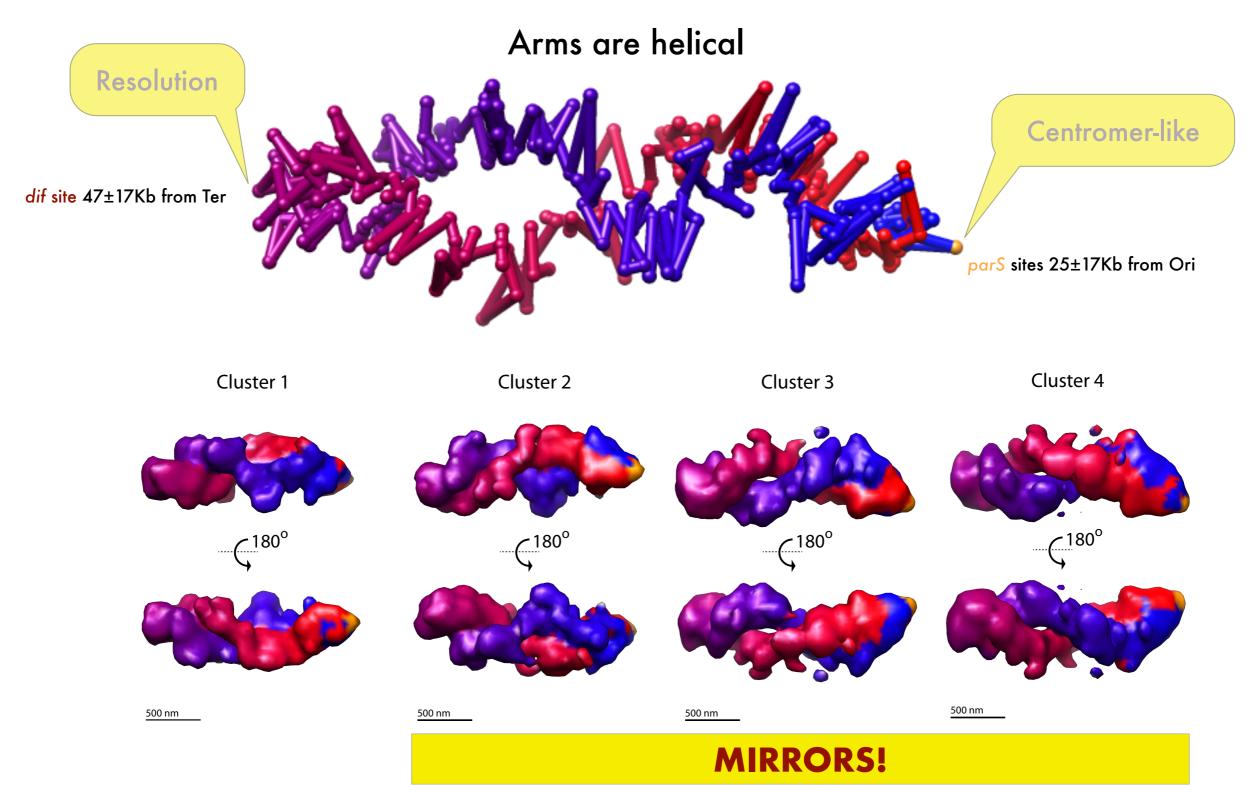


3D model building with the 5C + IMP approach

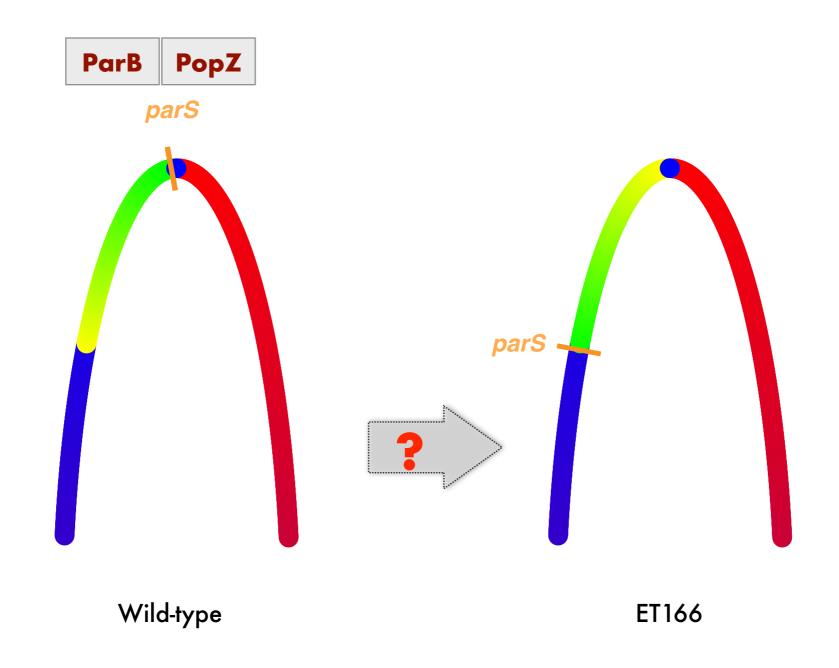




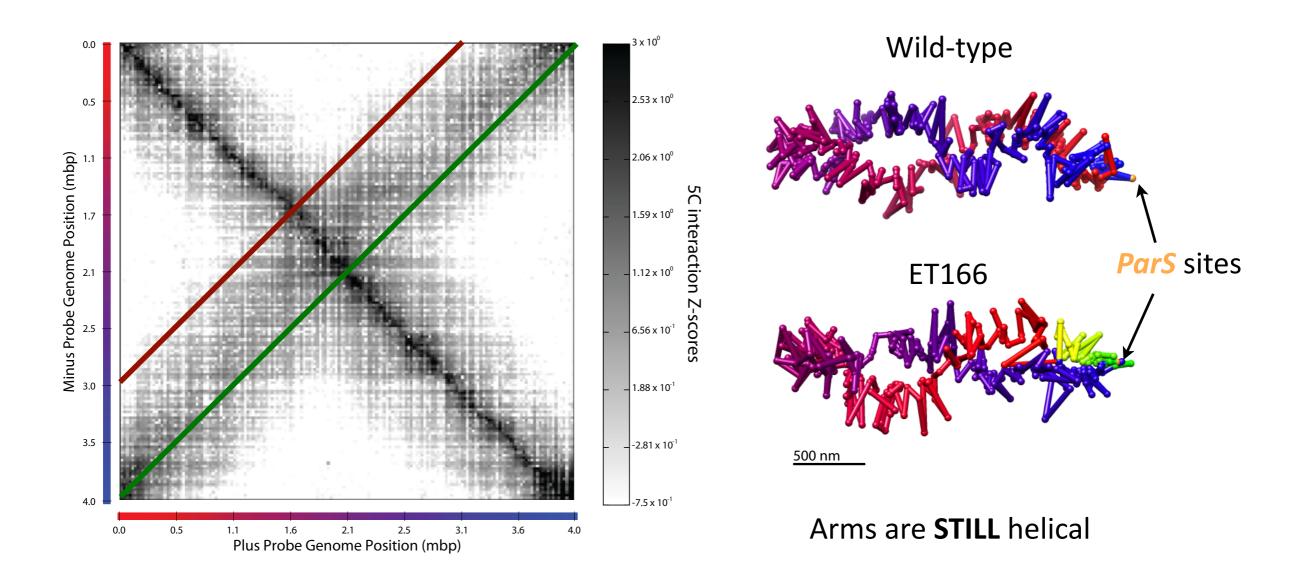
Genome organization in Caulobacter crescentus



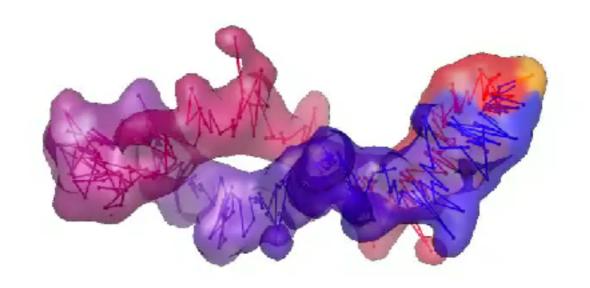
Moving the parS sites 400 Kb away from Ori

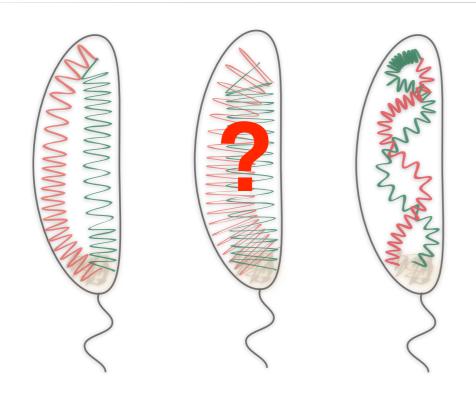


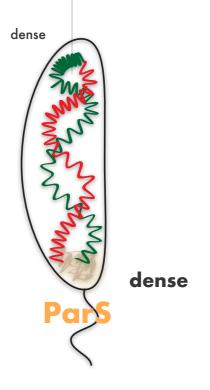
Moving the parS sites results in whole genome rotation!



Genome architecture in Caulobacter





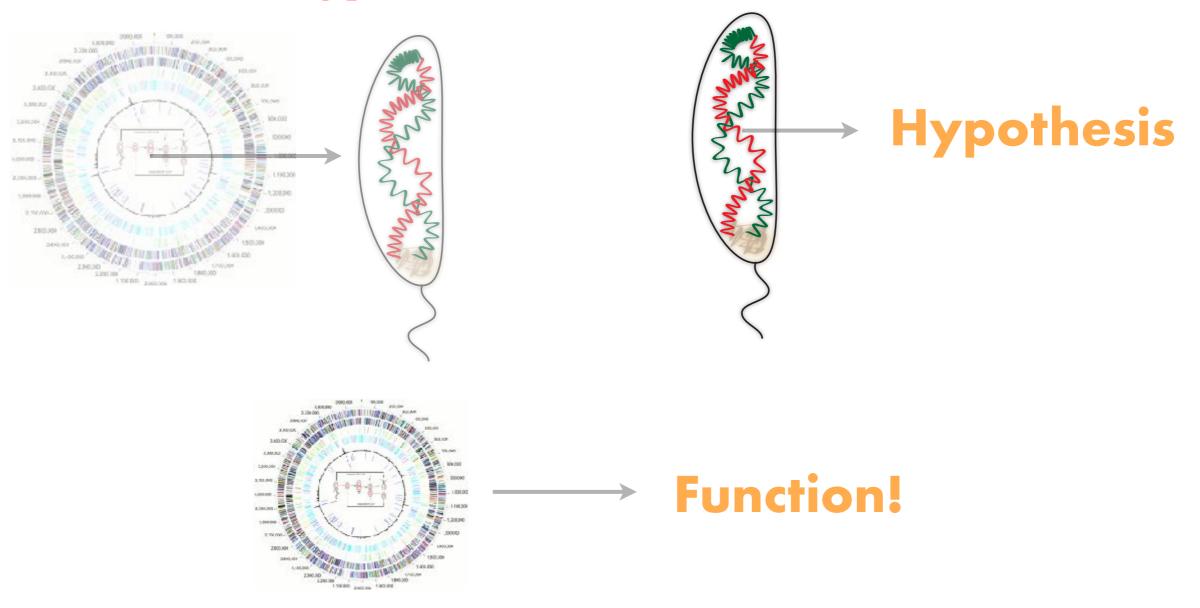


M.A. Umbarger, et al. Molecular Cell (2011) 44:252-264

From Sequence to Function

5C + IMP

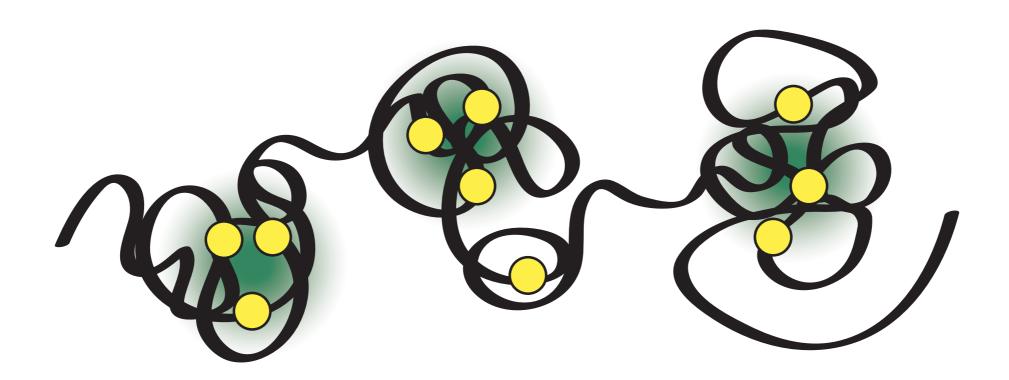
Technology



D. Baù and M.A. Marti-Renom Chromosome Res (2011) 19:25-35.



On TADs and hormones





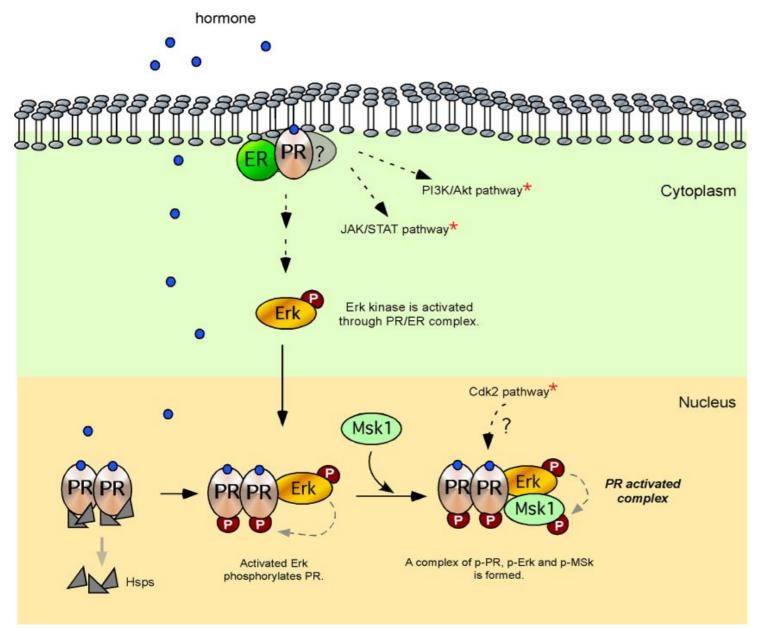
Davide Baù



François le Dily



Progesterone-regulated transcription in breast cancer



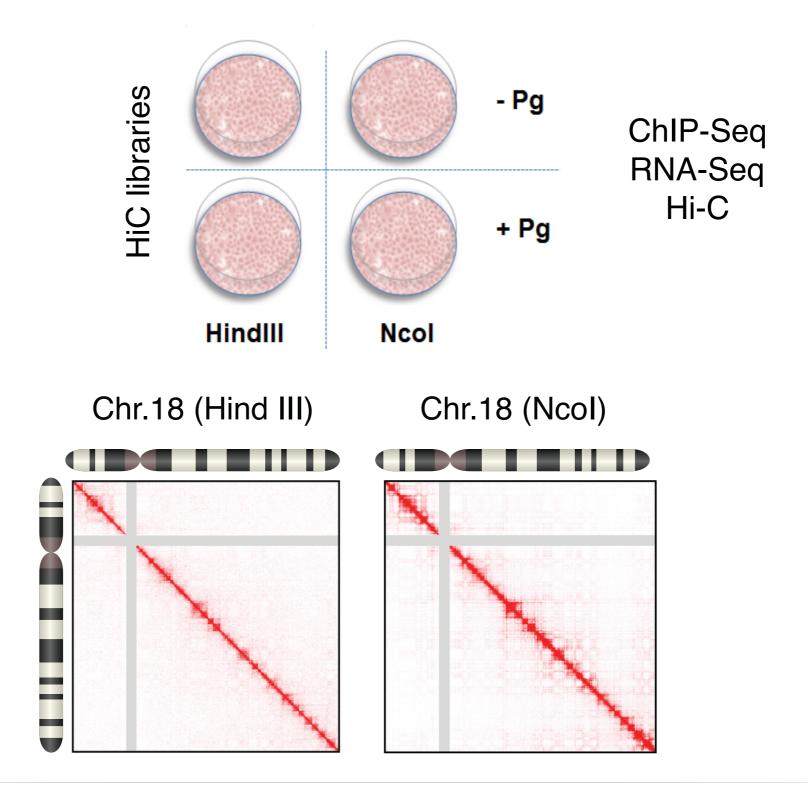
Vicent et al 2011, Wright et al 2012, Ballare et al 2012

- > 2,000 genes **Up**-regulated
- > 2,000 genes **Down**-regulated

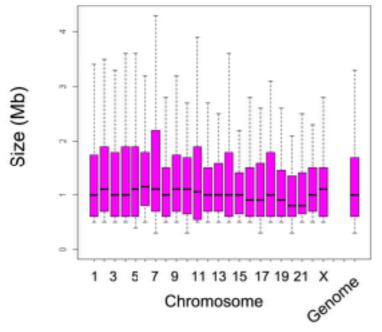
Regulation in 3D?



Experimental design

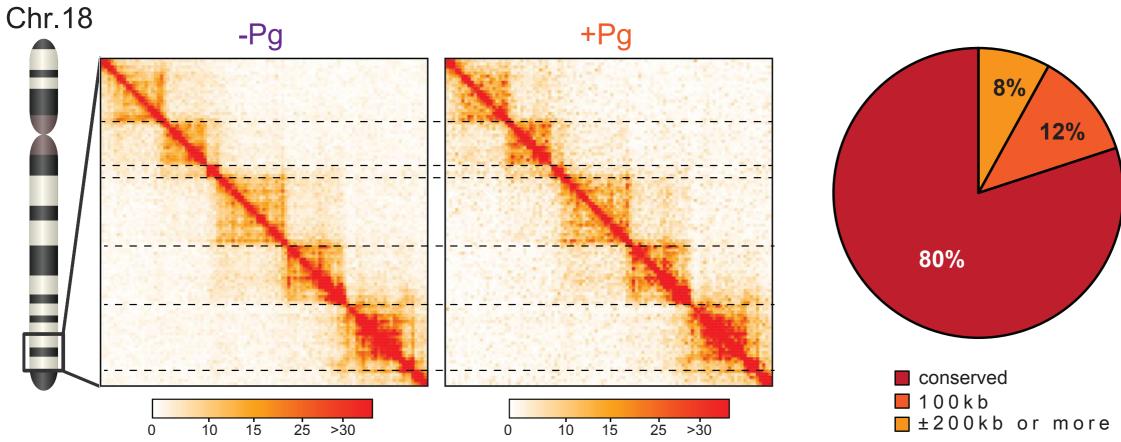


Are there TADs? how robust?



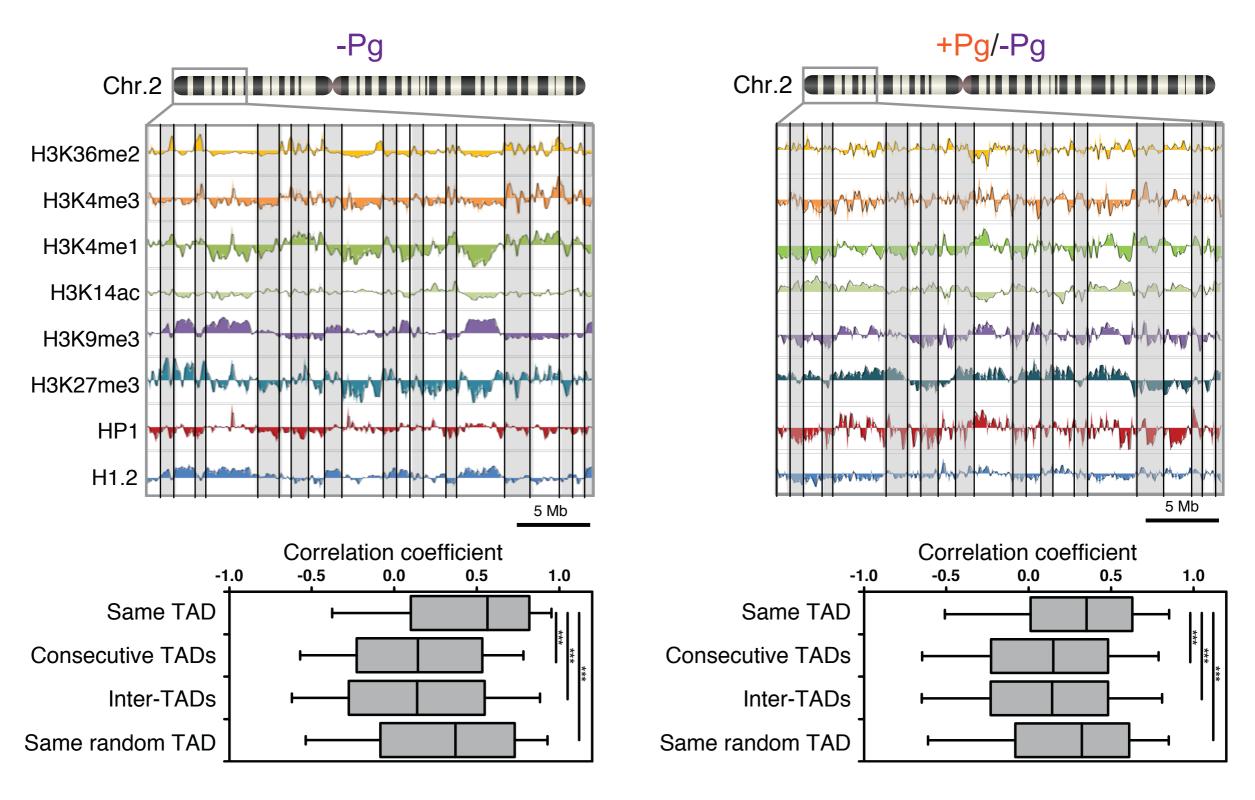
>2,000 detected TADs



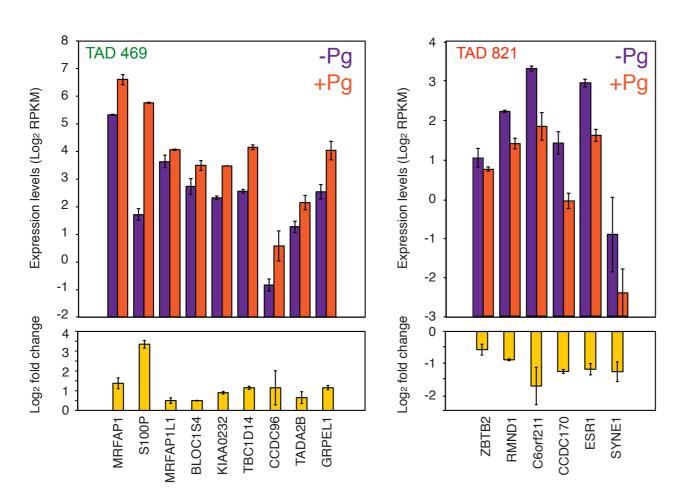


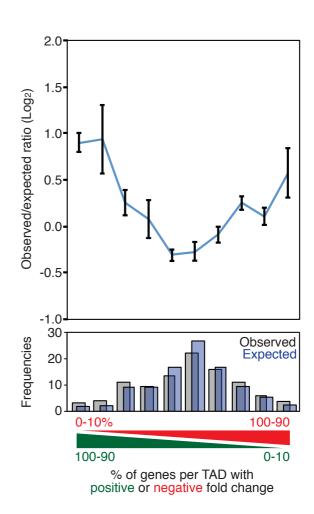


Are TADs homogeneous?

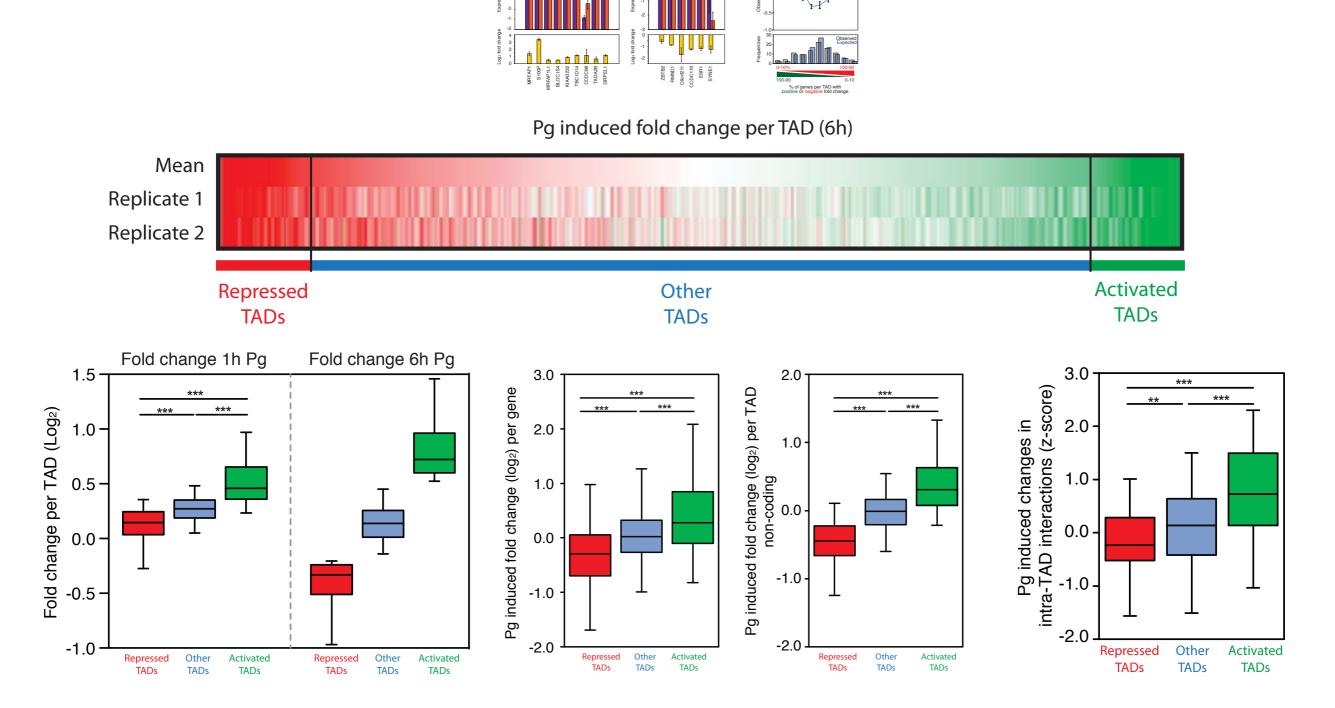


Do TADs respond differently to Pg treatment?

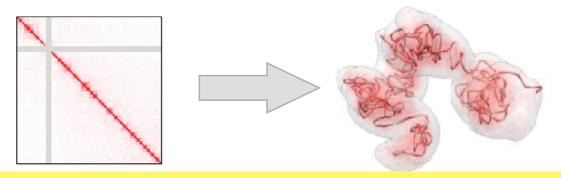




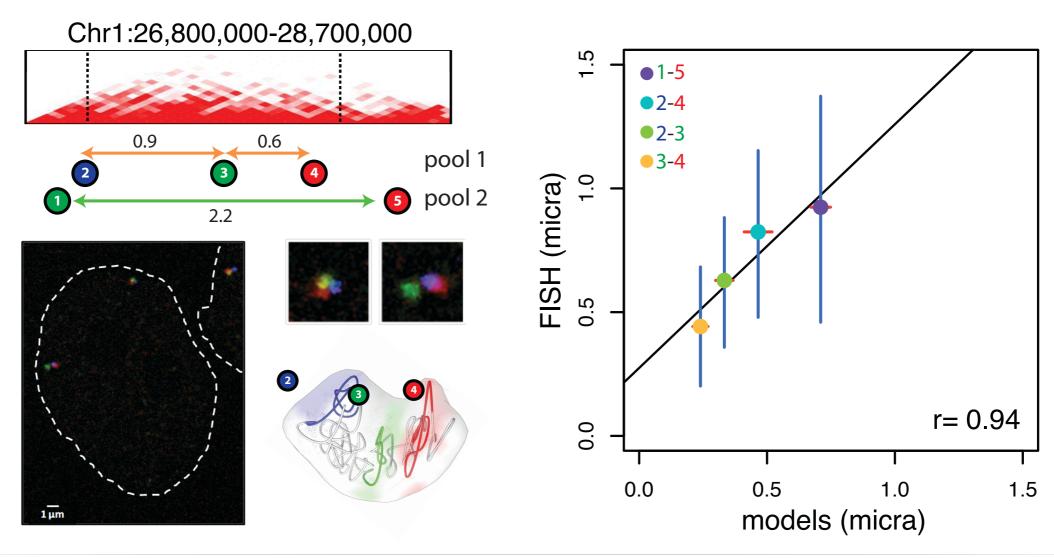
Do TADs respond differently to Pg treatment?



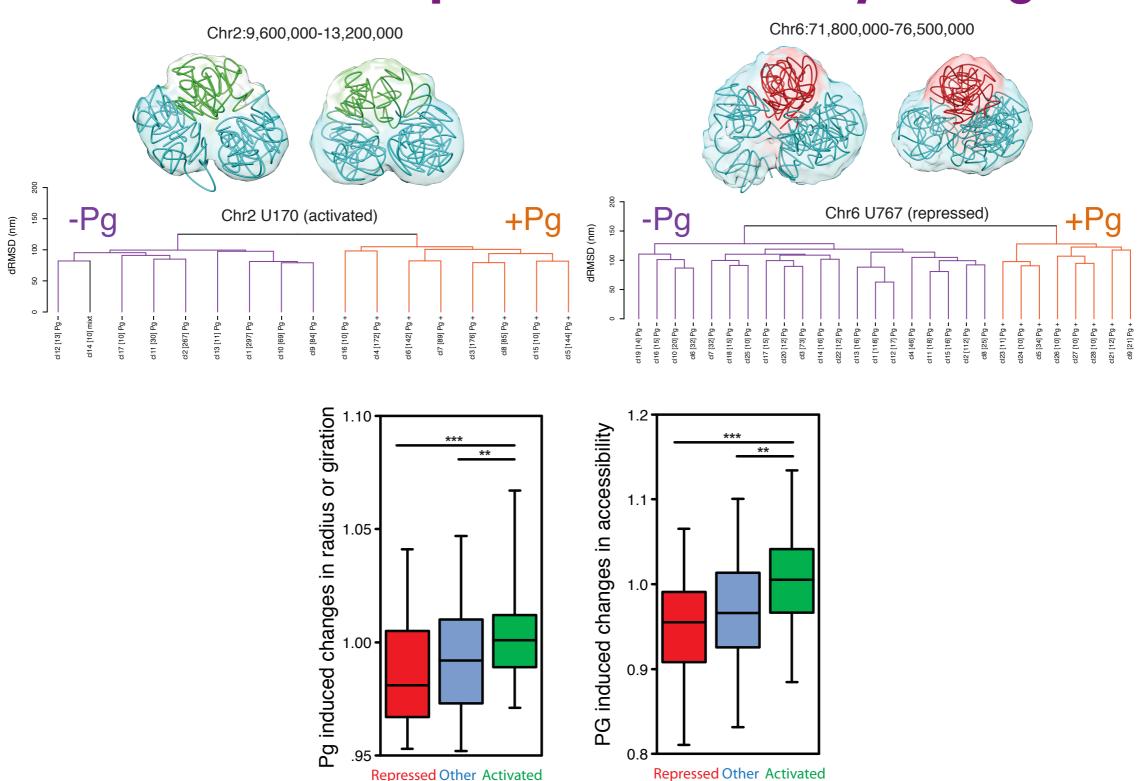
Modeling 3D TADs



61 genomic regions containing 209 TADs covering 267Mb



How TADs respond structurally to Pg?



TADs

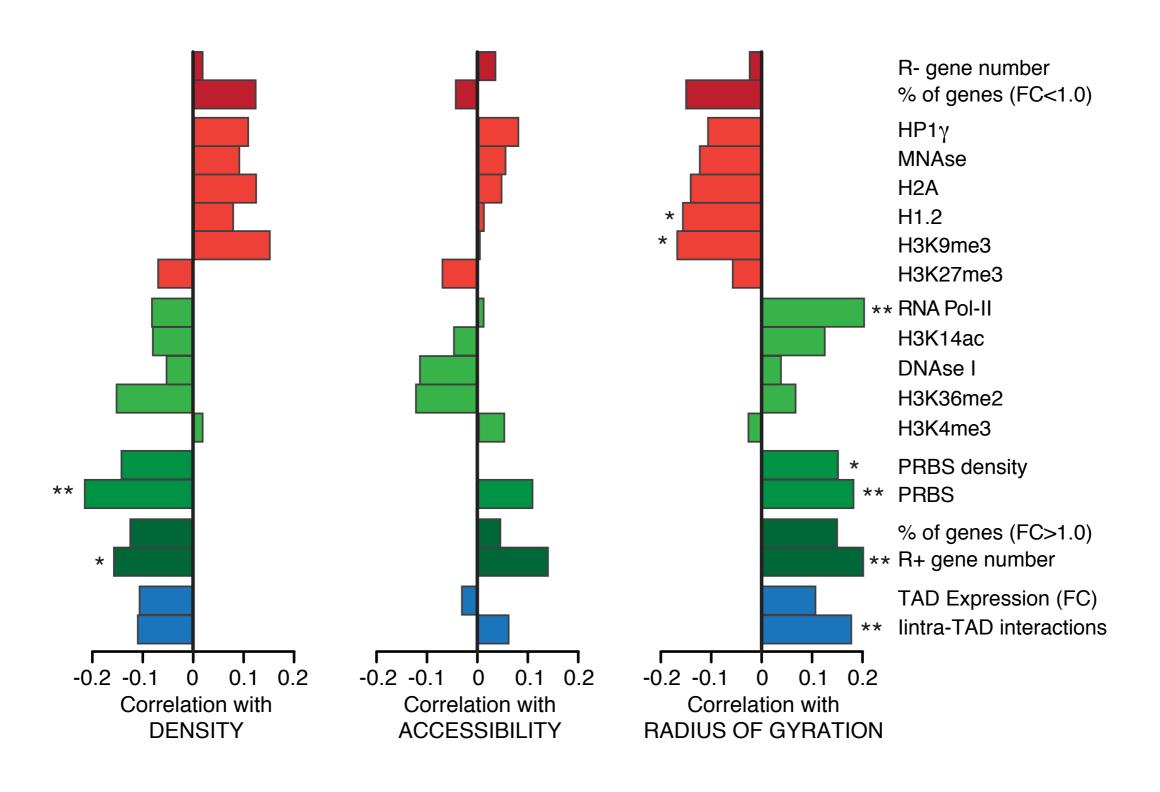
TADs

TADs

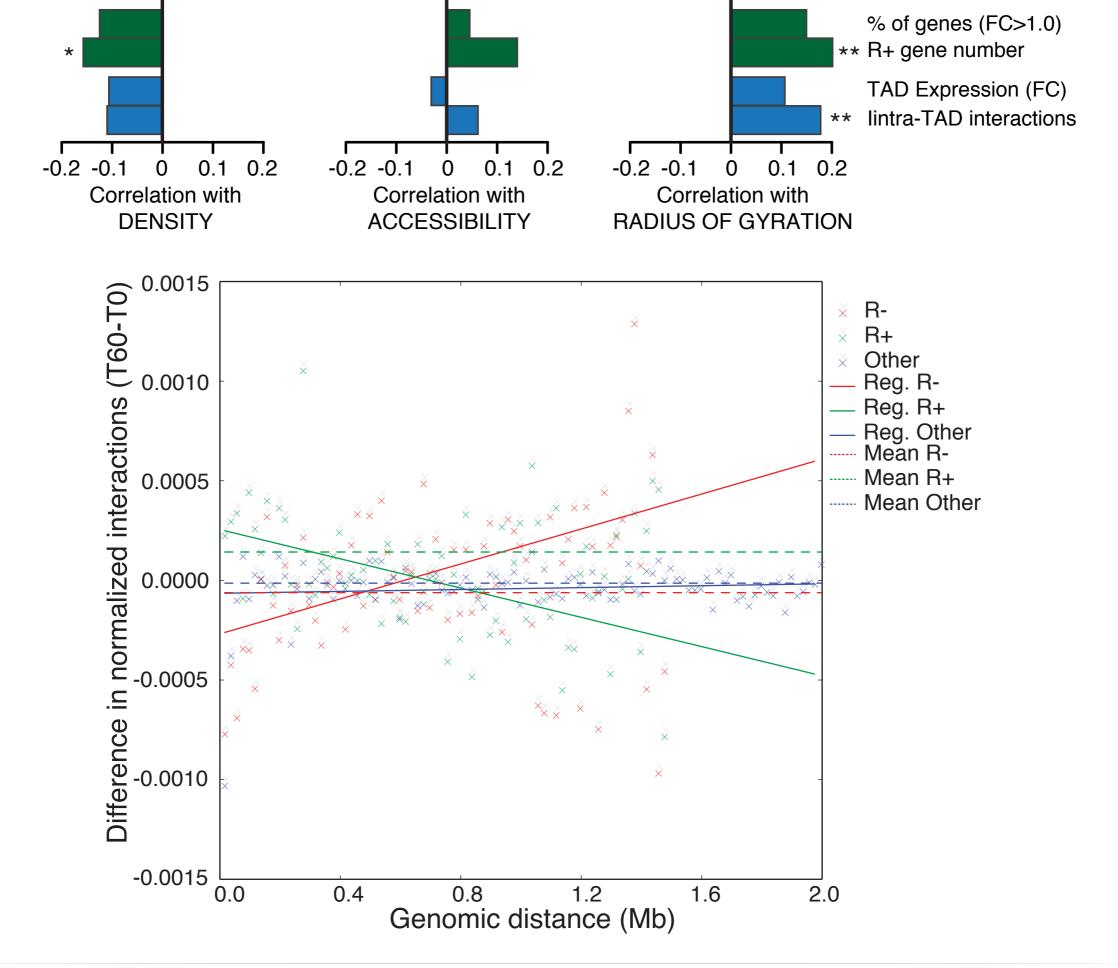
TADs

TADs

How TADs respond structurally to Pg?

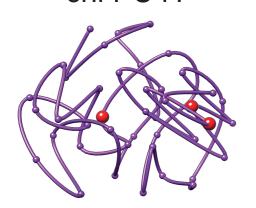




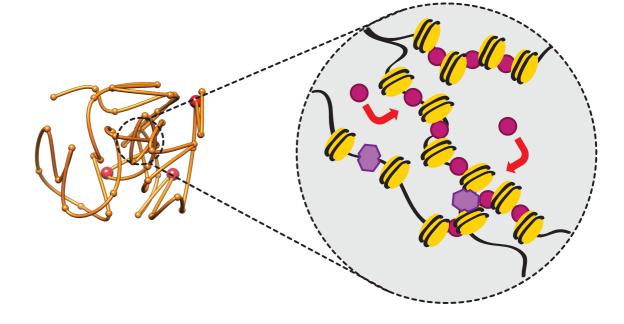


Model for TAD regulation

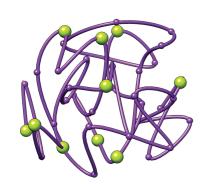
Repressed TAD chr1 U41



DHS
HP1
H1.2
H2A
MNAse
H3K27me3
H3K4me1
H3K4me1
H3K4me3



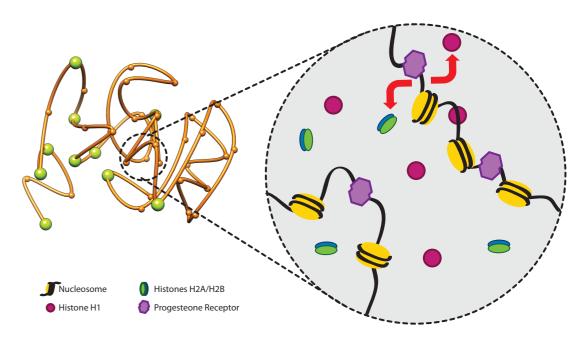
Activated TAD chr2 U207



Structural transition

+Pg





PLoS CB Outlook

Marti-Renom MA, Mirny LA (2011) PLoS Comput Biol 7(7): e1002125.

OPEN & ACCESS Freely available online

 $PLoS \, {\sf computational \, Biology}$

Review

Bridging the Resolution Gap in Structural Modeling of 3D Genome Organization

Marc A. Marti-Renom¹*, Leonid A. Mirny²

1 Structural Genomics Laboratory, Bioinformatics and Genomics Department, Centro de Investigación Príncipe Felipe, Valencia, Spain, 2 Harvard-MIT Division of Health Sciences and Technology, and Department of Physics, Massachusetts Institute of Technology, Cambridge, Massachusetts, United States of America

Abstract: Over the last decade, and especially after the advent of fluorescent in situ hybridization imaging and chromosome conformation capture methods, the availability of experimental data on genome three-dimensional organization has dramatically increased. We now have access to unprecedented details of how genomes organize within the interphase nucleus. Development of new computational approaches to leverage this data has already resulted in the first three-dimensional structures of genomic domains and genomes. Such approaches expand our knowledge of the chromatin folding principles, which has been classically studied using polymer physics and molecular simulations. Our outlook describes computational approaches for integrating experimental data with polymer physics, thereby bridging the resolution gap for structural determination of genomes and genomic domains.

This is an "Editors' Outlook" article for PLoS Computational Biology

Recent experimental and computational advances are resulting in an increasingly accurate and detailed characterization of how genomes are organized in the three-dimensional (3D) space of the nucleus (Figure 1) [1]. At the lowest level of chromatin organization, naked DNA is packed into nucleosomes, which forms the so-called chromatin fiber composed of DNA and proteins. However, this initial packing, which reduces the length of the DNA by about seven times, is not sufficient to explain the higher-order folding of chromosomes during interphase and metaphase. It is now accepted that chromosomes and genes are non-randomly and dynamically positioned in the cell nucleus during the interphase, which challenges the classical representation of genomes as linear static sequences. Moreover, compartmentalization, chromatin organization, and spatial location of genes are associated with gene expression and the functional status of the cell. Despite the importance of 3D genomic architecture. we have a limited understanding of the molecular mechanisms that determine the higher-order organization of genomes and its relation to function. Computational biology plays an important role in the plethora of new technologies aimed at addressing this knowledge gap [2]. Indeed, Thomas Cremer, a pioneer in studying nuclear organization using light microscopy, recently highlighted the importance of computational science in complementing and leveraging experimental observations of genome organization [2]. Therefore, computational approaches to integrate experimental observations with chromatin physics are needed to determine the architecture (3D) and dynamics (4D) of genomes.

We present two complementary approaches to address this challenge: (i) the first approach aims at developing simple polymer models of chromatin and determining relevant interactions (both

physical and biological) that explain experimental observations; (ii) the second approach aims at integrating diverse experimental observations into a system of spatial restraints to be satisfied, thereby constraining possible structural models of the chromatin. The goal of both approaches is dual: to obtain most accurate 3D and 4D representation of chromatin architecture and to understand physical constraints and biological phenomena that determine its organization. These approaches are reminiscent of the proteinfolding field where the first strategy was used for characterizing protein "foldability" and the second was implemented for modeling the structure of proteins using nuclear magnetic resonance and other experimental constraints. In fact, our outlook consistently returns to the many connections between the two fields.

What Does Technology Show Us?

Today, it is possible to quantitatively study structural features of genomes at diverse scales that range from a few specific loci, through chromosomes, to entire genomes (Table 1) [3]. Broadly, there are two main approaches for studying genomic organization: light microscopy and cell/molecular biology (Figure 2). Light microcopy [4], both with fixed and living cells, can provide images of a few loci within individual cells [5,6], as well as their dynamics as a function of time [7] and cell state [8]. On a larger scale, light microscopy combined with whole-chromosome staining reveals chromosomal territories during interphase and their reorganization upon cell division. Immunofluorescence with fluorescent antibodies in combination with RNA, and DNA fluorescence in situ hybridization (FISH) has been used to determine the colocalization of loci and nuclear substructures.

Using cellular and molecular biology, novel chromosome conformation capture (3C)-based methods such 3C [9], 3C-on-chip or circular 3C (the so-called 4C) [10,11], 3C carbon copy (5C) [12], and Hi-C [13] quantitatively measure frequencies of spatial contacts between genomic loci averaged over a large

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