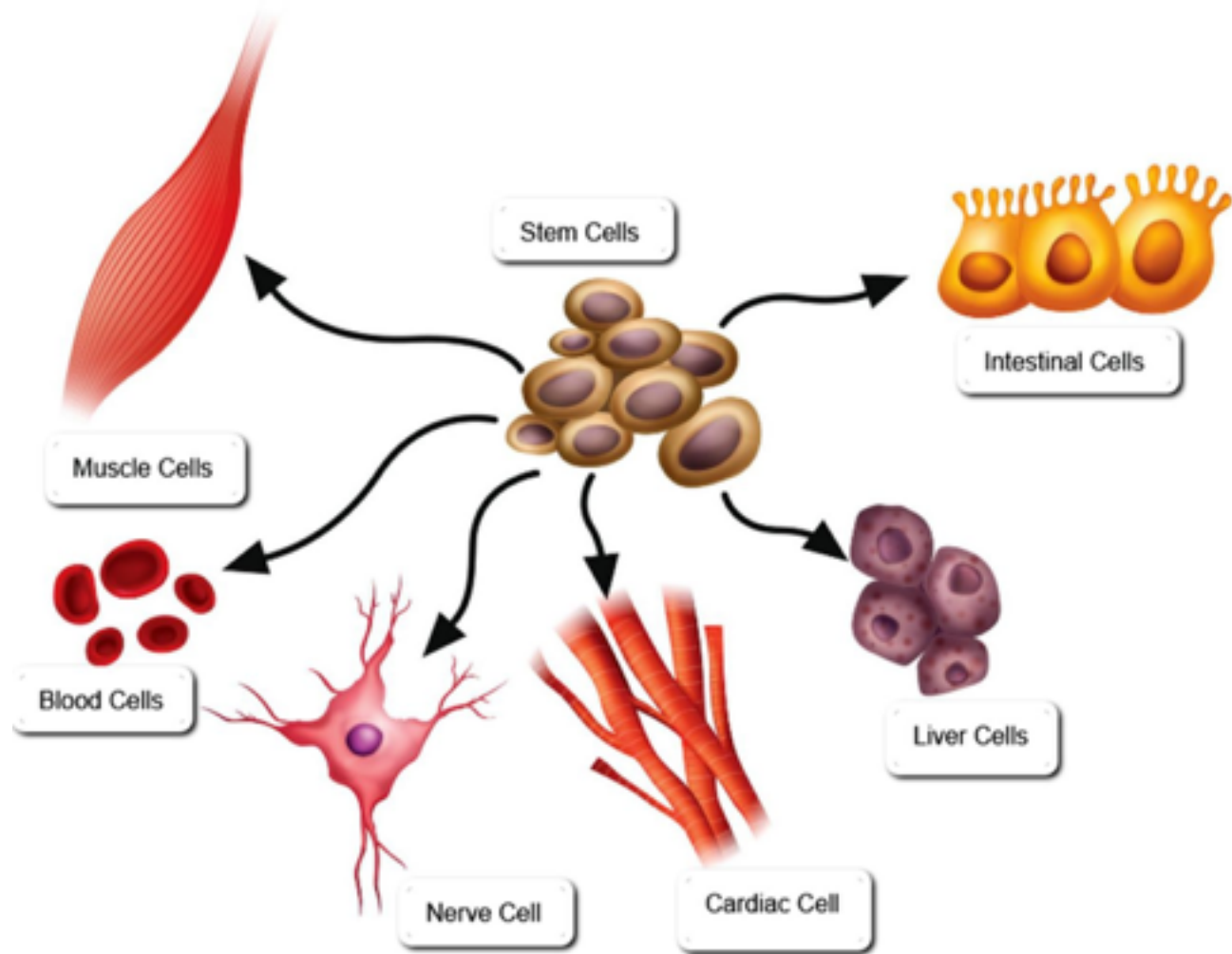


# 3D genome architecture

Brad Krajina  
Alex Yoshikawa

# The Human Genome Encodes for Rich Phenotypic Diversity

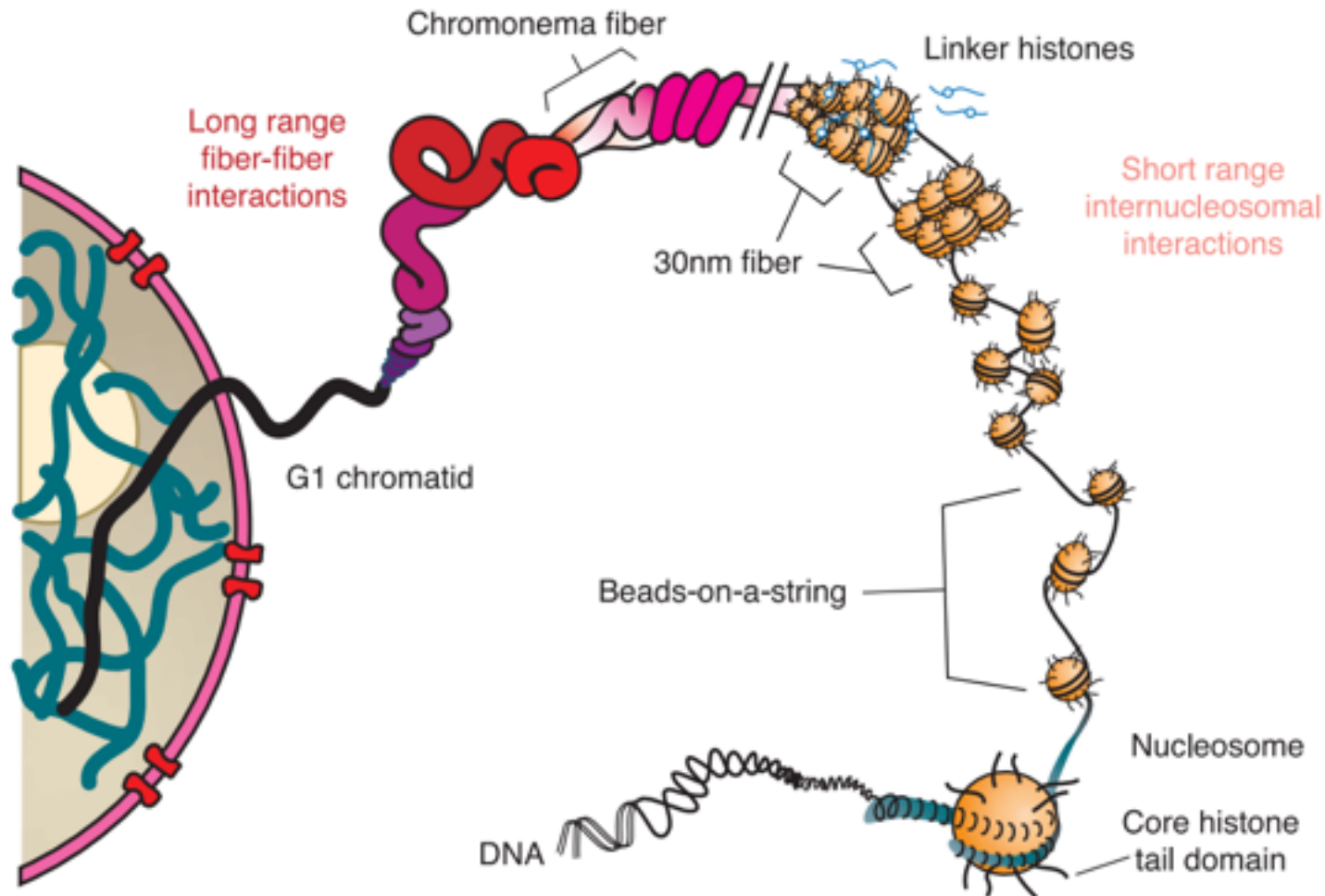


Stephen J. Farenga, Daniel Ness, Michael Hutchinson. *The American Biology Teacher* 77(6). 2015

Unlocking how gene expression is regulated to give rise to a variety of phenotypes from a single genotype may play a central role in understanding human development and disease

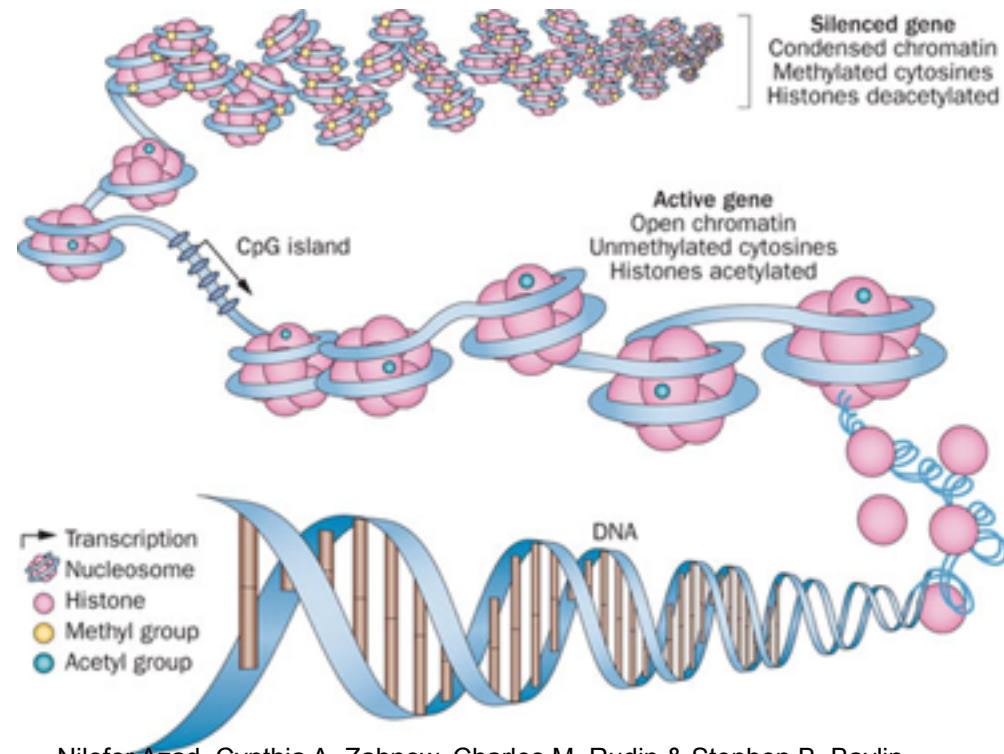
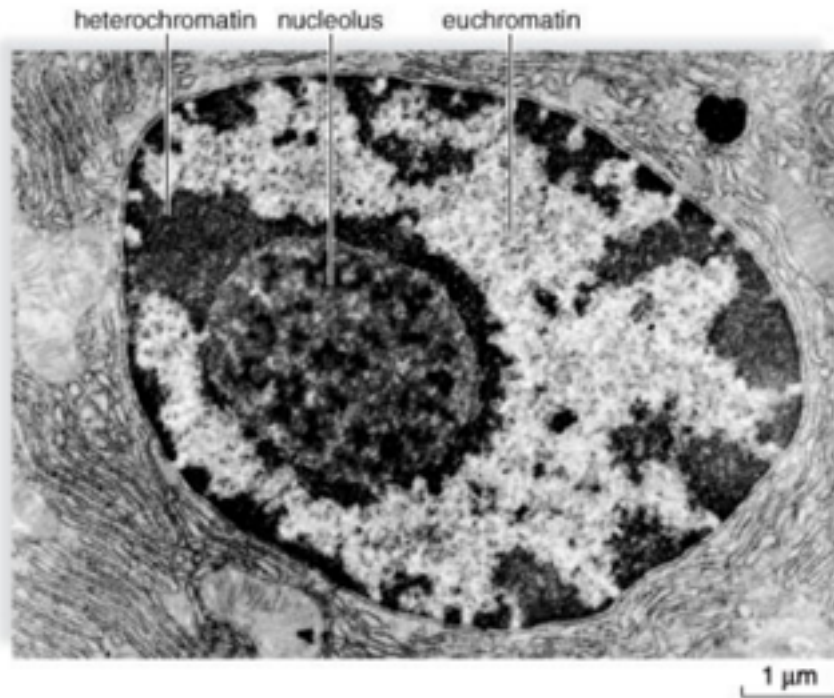
# Beyond the Central Dogma: 3D Chromosome Organization Regulates Genetic Processes

A human cell must compact ~1 m of DNA into a single nucleus, requiring exquisite control of hierarchical 3D organization



# Eukaryotic Chromosomes are Organized into Transcriptionally Active and Transcriptionally Inactive Domains

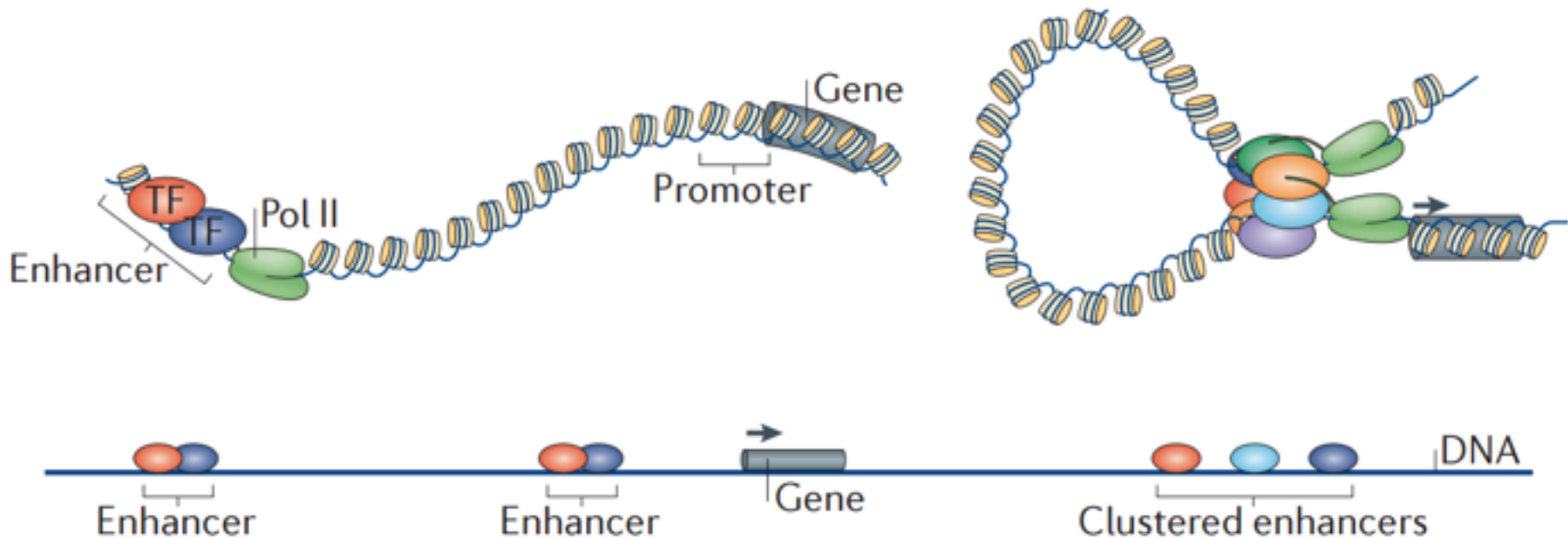
**Heterochromatin** is densely packaged and **transcriptionally silenced**.  
**Euchromatin** is loosely packaged and **transcriptionally active**.



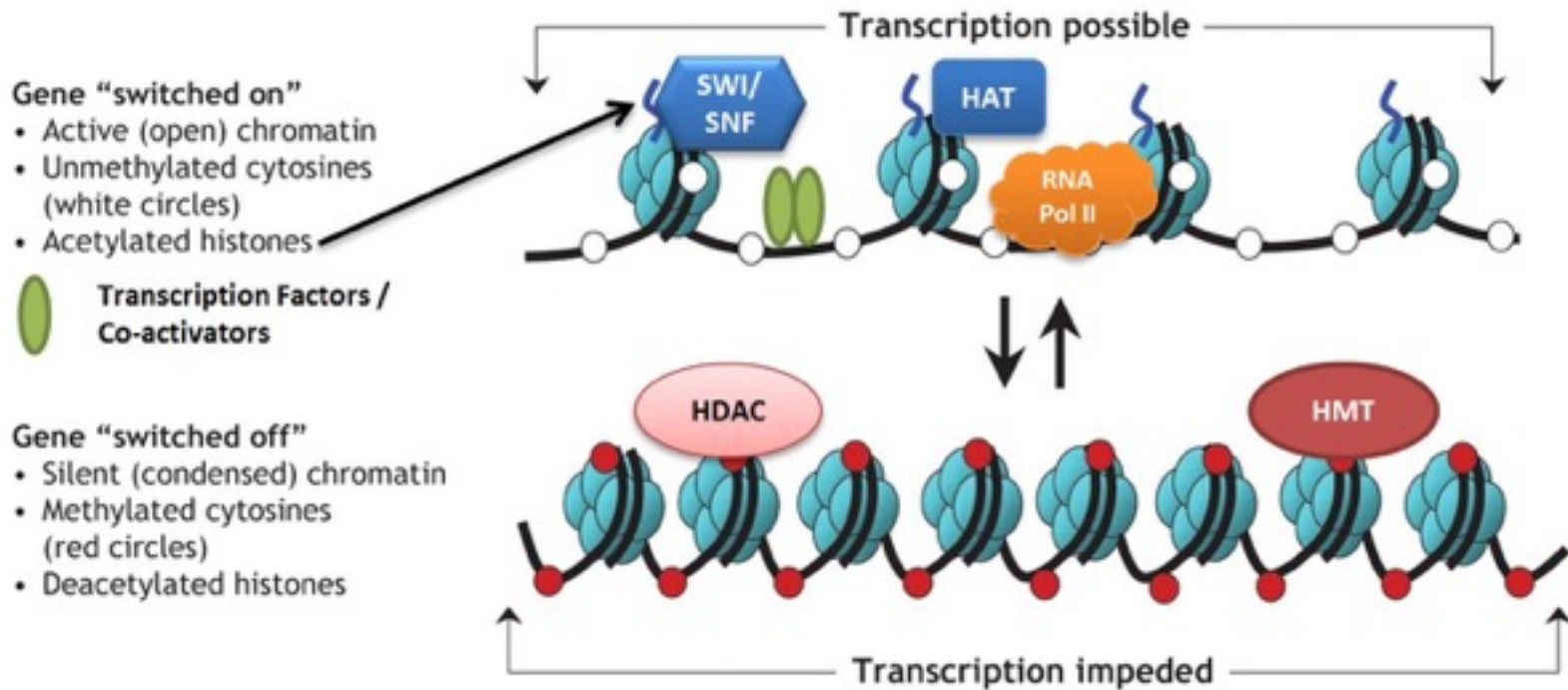
Nilofer Azad, Cynthia A. Zahnow, Charles M. Rudin & Stephen B. Baylin  
Nature Reviews Clinical Oncology 10, 256-266. 2013

# DNA looping regulates gene expression

Chromatin looping enables transcriptional activation by contacts between gene promoters and genomically distal enhancer sequences

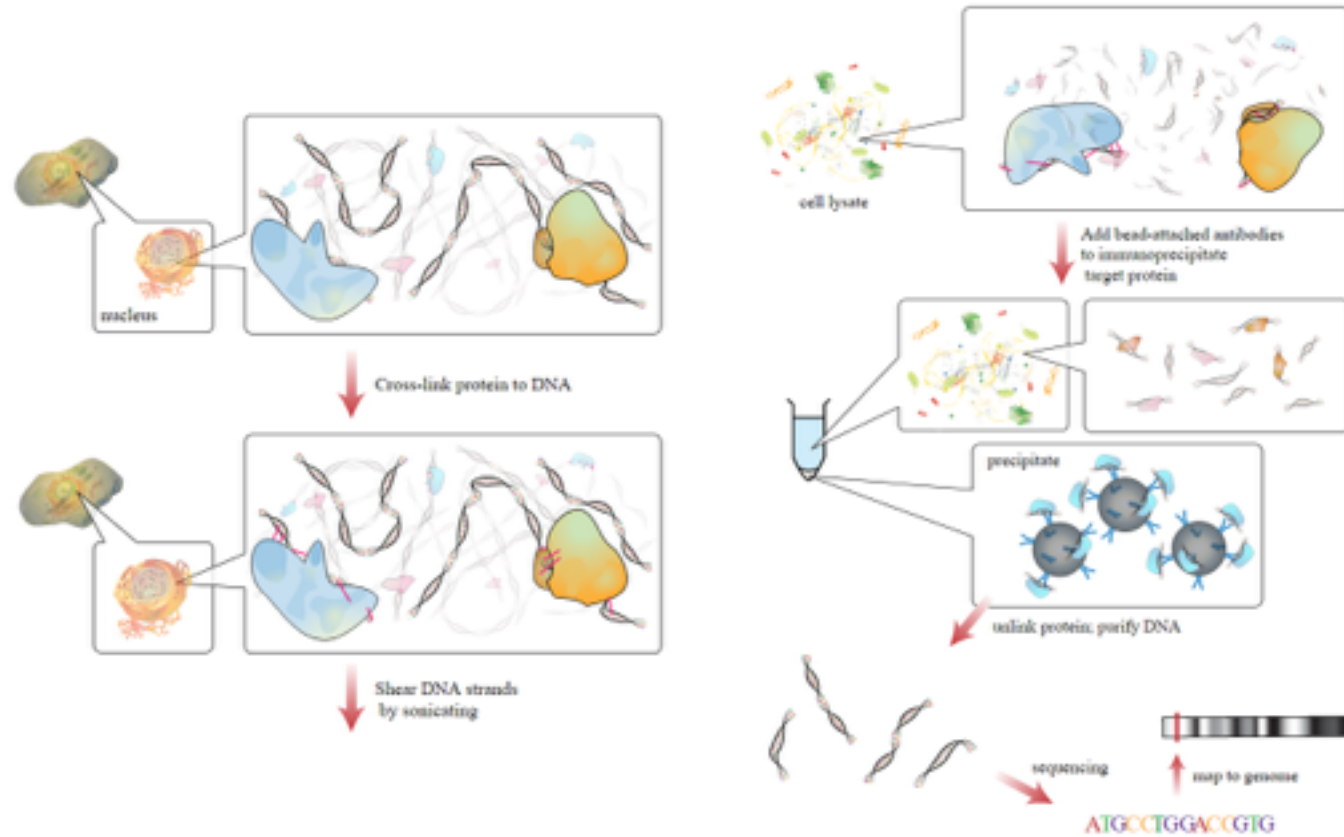


# Proteins regulate genome architecture

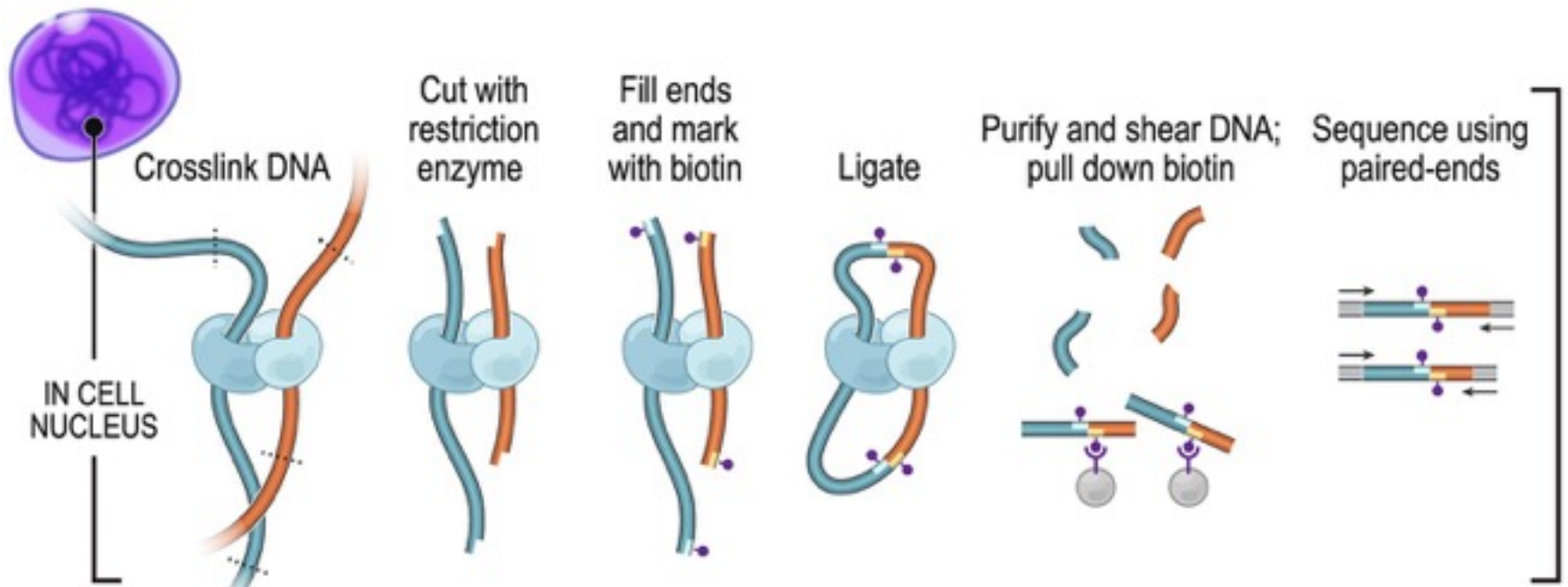


[Restructuring Chromosomes Youtube Link](#)

# Chip-Seq is a method to study protein-DNA interactions



# Hi-C is a method to study 3D genome architecture





# A 3D Map of the Human Genome at Kilobase Resolution Reveals Principles of Chromatin Looping

Suhas S.P. Rao,<sup>1,2,3,4,10</sup> Miriam H. Huntley,<sup>1,2,3,4,5,10</sup> Neva C. Durand,<sup>1,2,3,4</sup> Elena K. Stamenova,<sup>1,2,3,4</sup> Ivan D. Bochkov,<sup>1,2,3</sup> James T. Robinson,<sup>1,4</sup> Adrian L. Sanborn,<sup>1,2,3,6</sup> Ido Machol,<sup>1,2,3</sup> Arina D. Omer,<sup>1,2,3</sup> Eric S. Lander,<sup>4,7,8,\*</sup> and Erez Lieberman Aiden<sup>1,2,3,4,9,\*</sup>

<sup>1</sup>The Center for Genome Architecture, Baylor College of Medicine, Houston, TX 77030, USA

<sup>2</sup>Department of Molecular and Human Genetics, Baylor College of Medicine, Houston, TX 77030, USA

<sup>3</sup>Department of Computer Science, Department of Computational and Applied Mathematics, Rice University, Houston, TX 77005, USA

<sup>4</sup>Broad Institute of MIT and Harvard, Cambridge, MA 02139, USA

<sup>5</sup>School of Engineering and Applied Sciences, Harvard University, Cambridge, MA 02138, USA

<sup>6</sup>Department of Computer Science, Stanford University, Stanford, CA 94305, USA

<sup>7</sup>Department of Biology, Massachusetts Institute of Technology (MIT), Cambridge, MA 02139, USA

<sup>8</sup>Department of Systems Biology, Harvard Medical School, Boston, MA 02115, USA

<sup>9</sup>Center for Theoretical Biological Physics, Rice University, Houston, TX 77030, USA

<sup>10</sup>Co-first author

# HiC reveals organization of human chromosomes

Contact probability maps from HiC reveal regions of the chromosome that preferentially interact

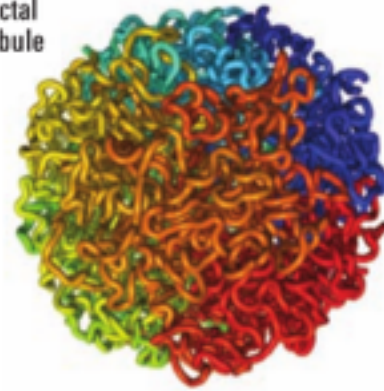
“Unfolded” genome



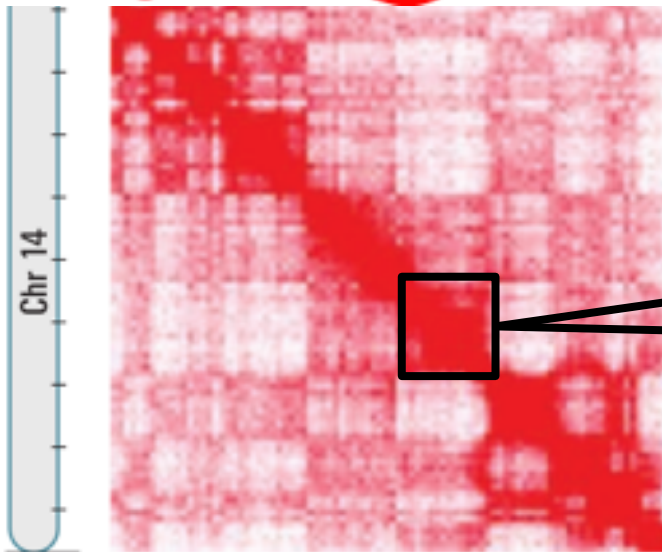
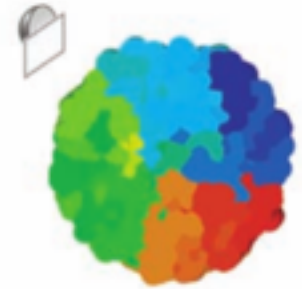
“Folded” genome with contact domains



Fractal globule

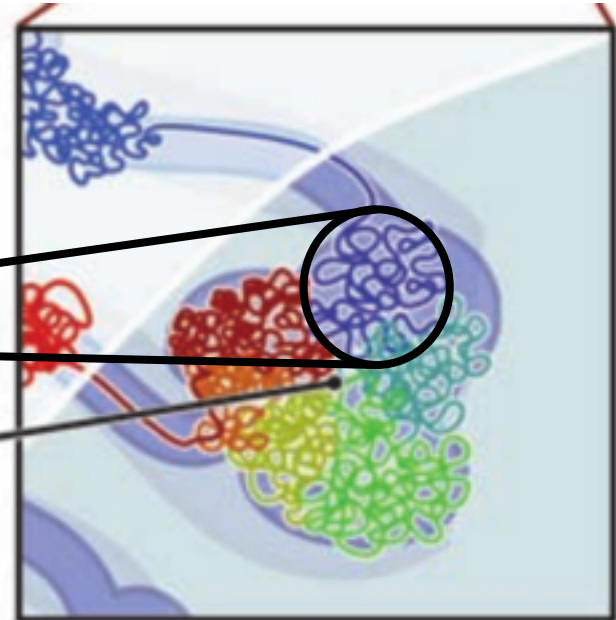


Cross-section view



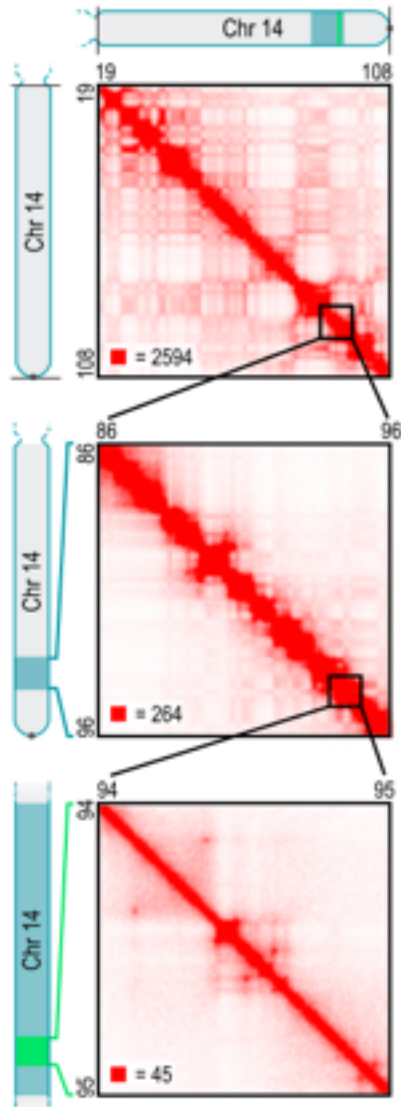
Megabase scale

Fractal globule

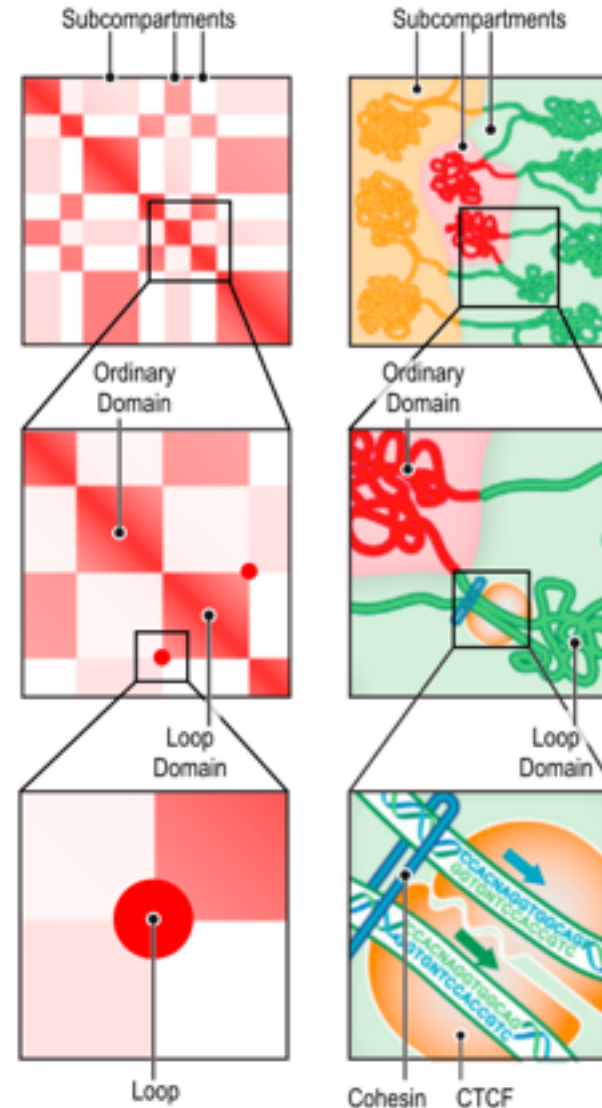


# *In situ* Hi-C reveals hierarchical organization and function of the human genome

*In situ* Hi-C enables kilobase-scale resolution of contact maps



Kilobase-scale resolution elucidates hierarchical genome organization



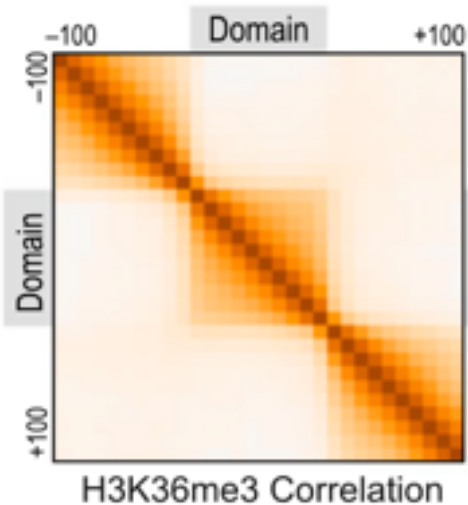
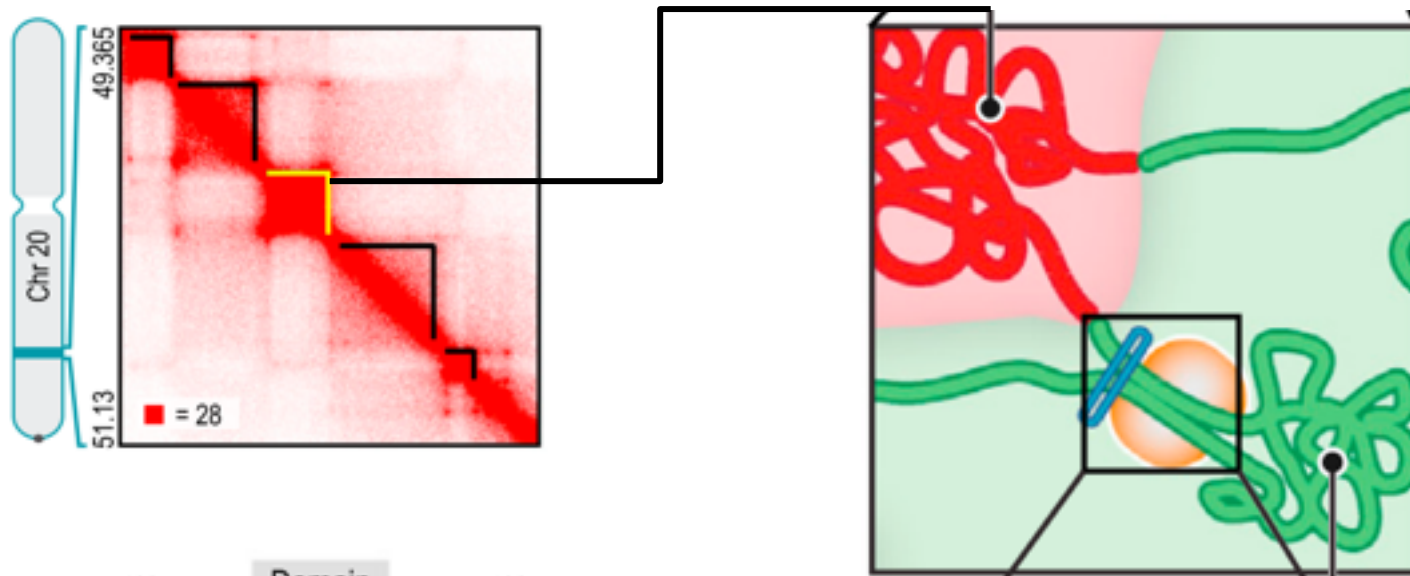
Sub-compartments organize distinct Epigenetic patterns

Megabase domains organize preferential regions of self-contact

Protein-mediated loops demarcate domain boundaries and regulate gene activation

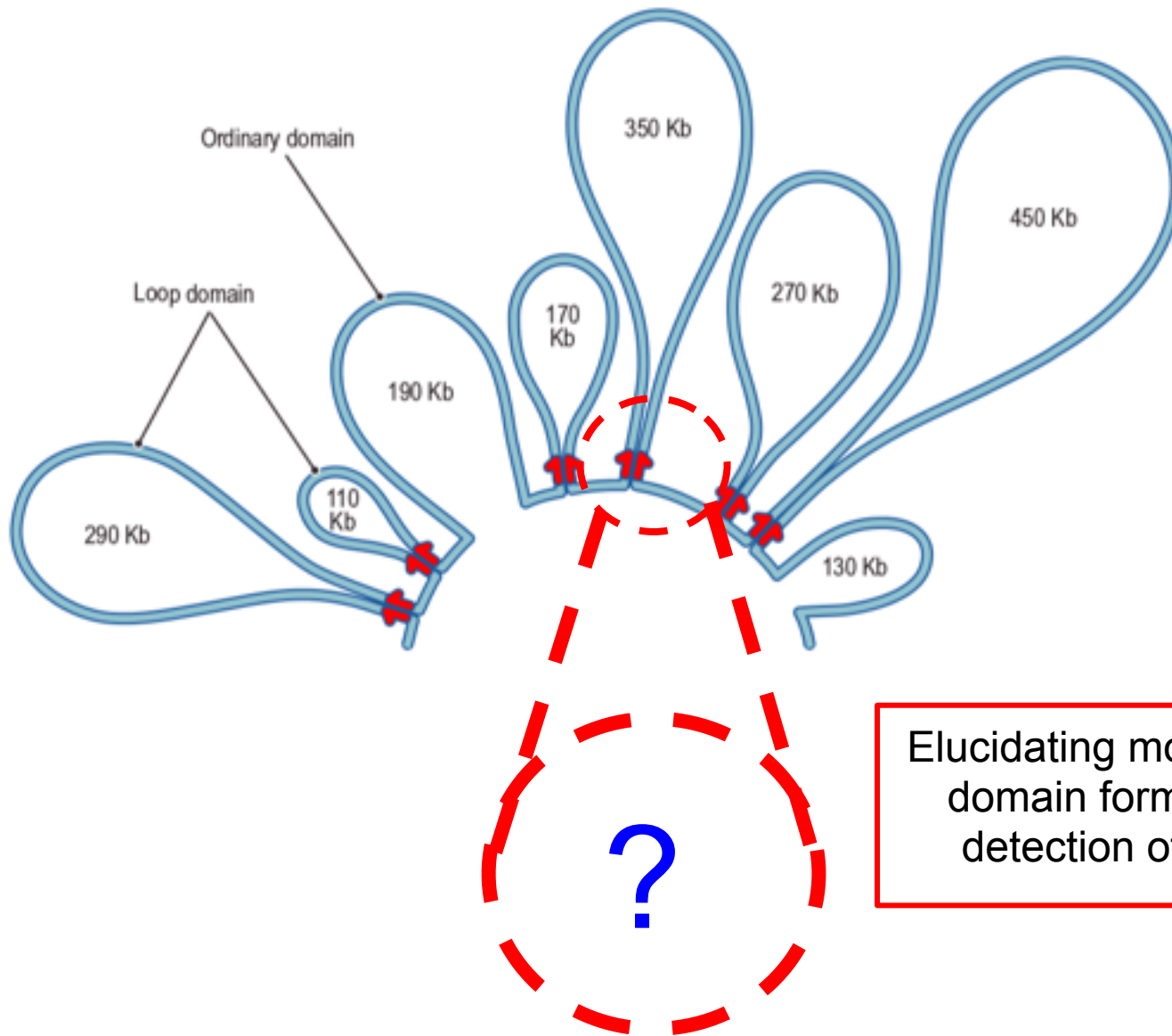
# Contact domains segregate the chromosome into contact domains with distinct epigenetic profiles

Genome contact maps reveal domains with preferential self-contacts



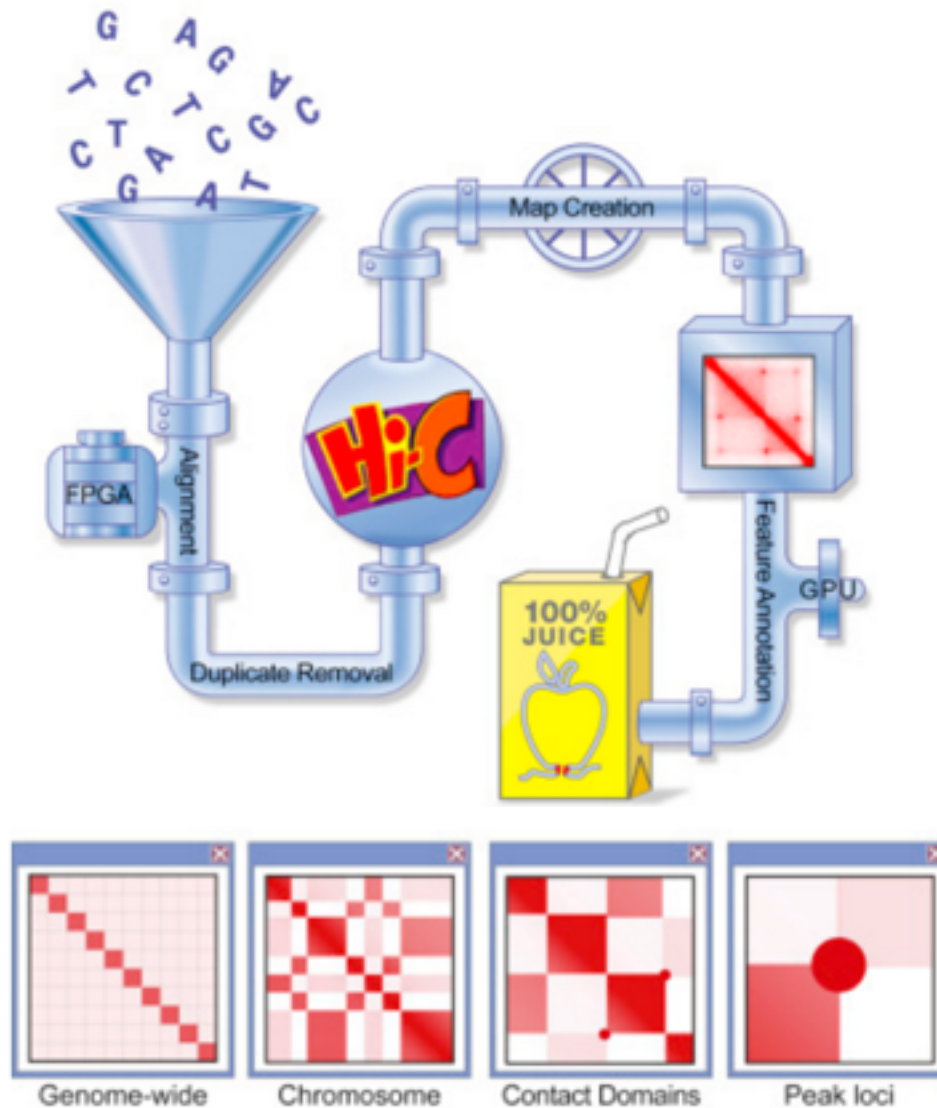
Contact domains exhibit correlations in histone modifications

# What are the “molecular ties” that define domains?



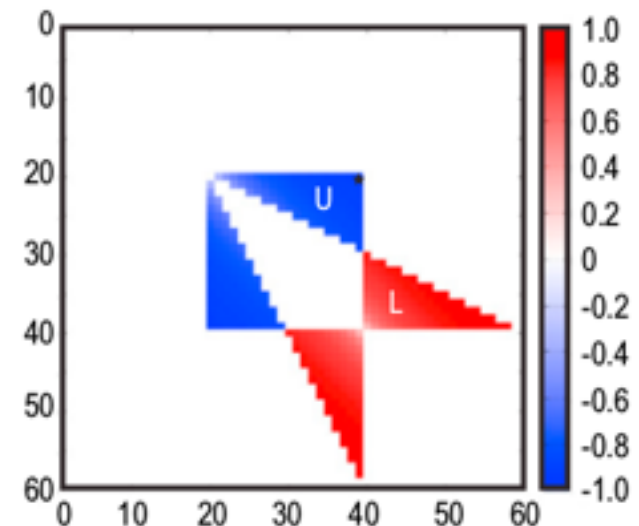
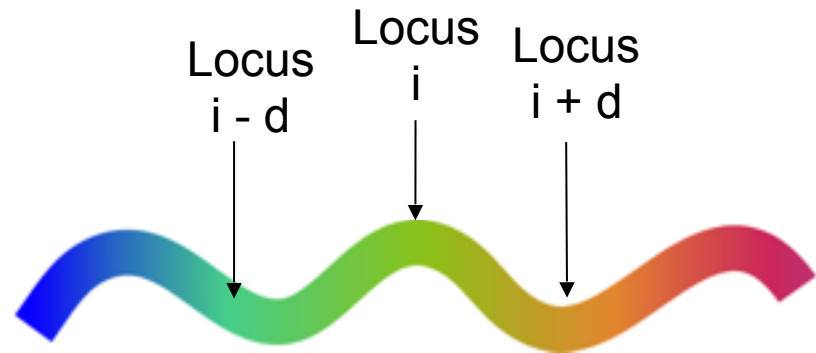
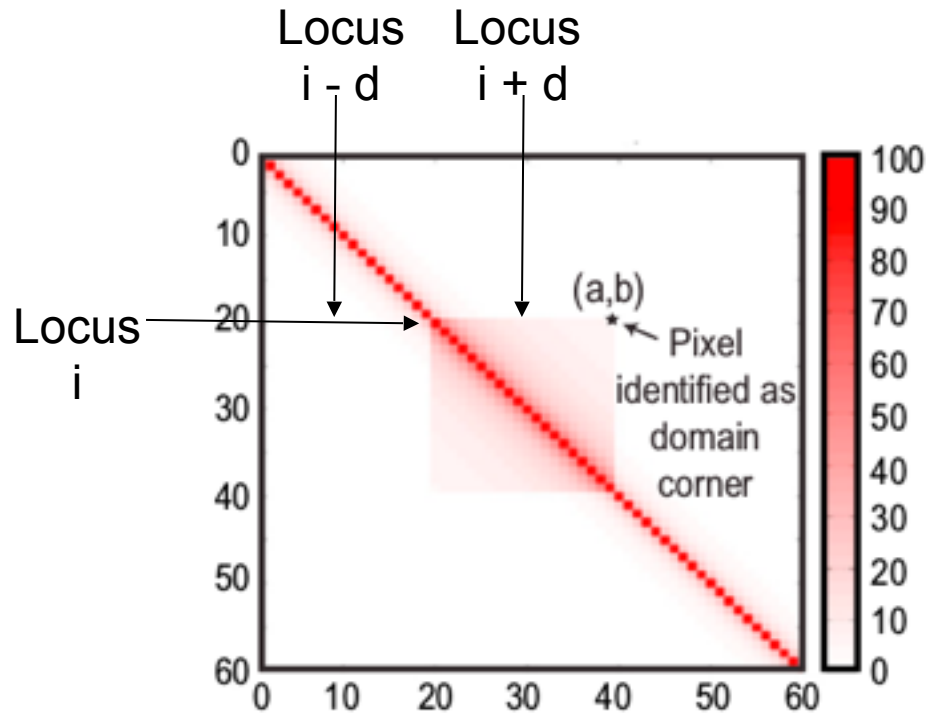
Elucidating molecular mechanisms of domain formation requires robust detection of domain boundaries

“Juicer” and “Juicebox” provide an automated pipeline for analyzing HiC sequencing data



# An “arrowhead” matrix algorithm enables more robust detection of contact domain boundaries

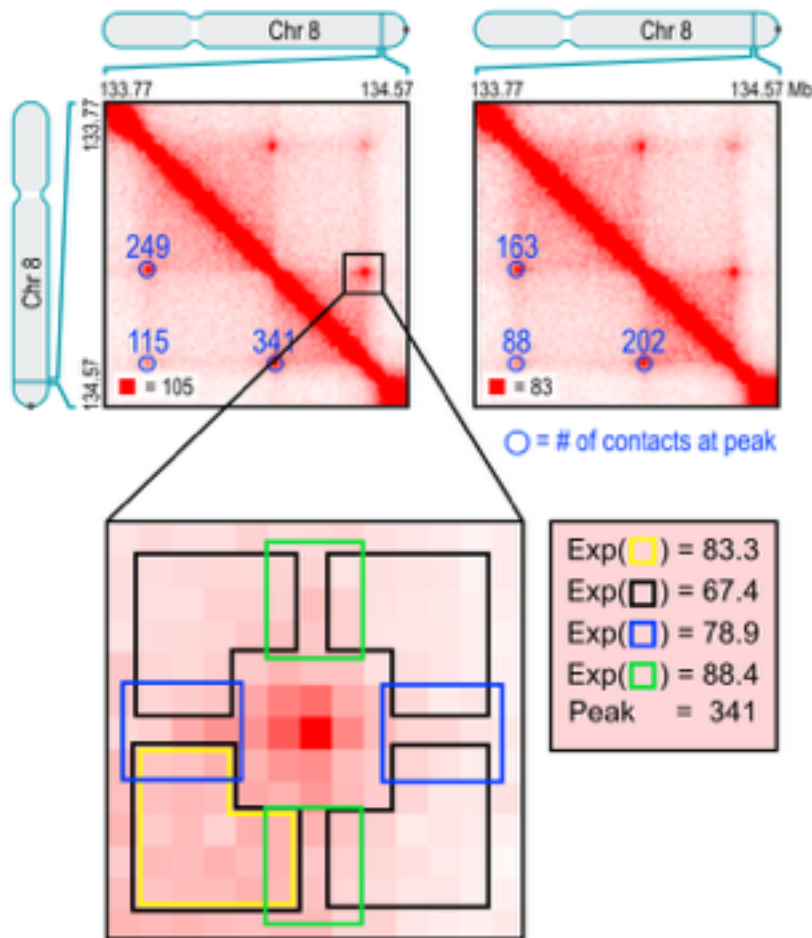
The arrowhead matrix algorithm identifies relative enrichment or depletion of contacts with loci an equal distance ( $d$ ) upstream and downstream of a reference locus



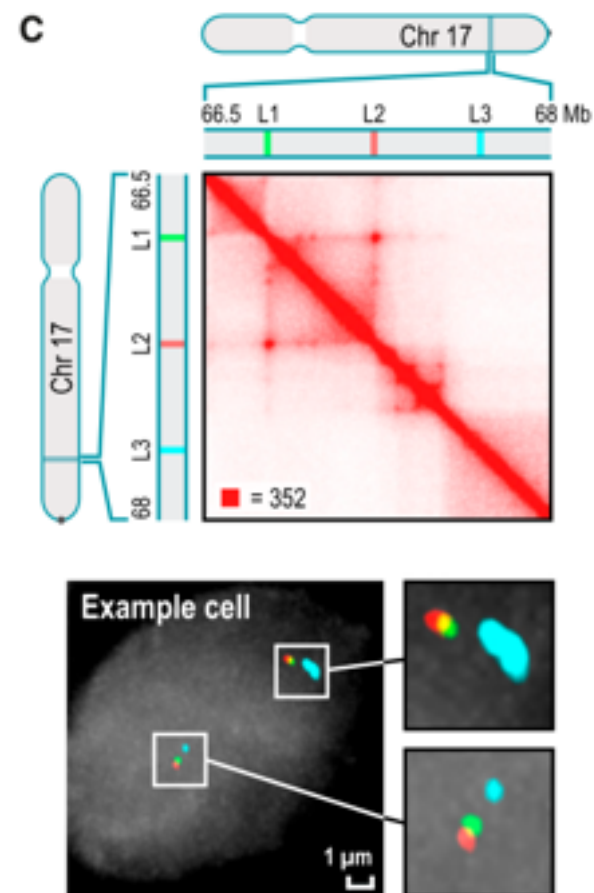
The arrow head matrix is **negative** if locus  $i + d$  is **enriched** in contacts with locus  $i$  compared to  $i - d$ , **positive** if it is **deficient**, and **zero** if it is **neither**

# “HICCUP” local background analysis reveals peaks in contact maps that may demarcate chromatin loops

Local background analysis identifies regions within contact domains that exhibit unusually high contact probabilities



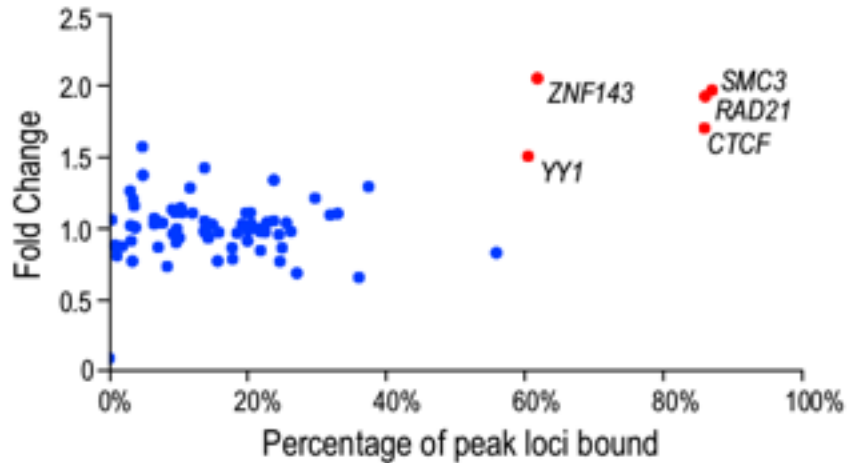
Fluorescence *in situ* hybridization imaging recapitulates contact of loci identified by peak analysis



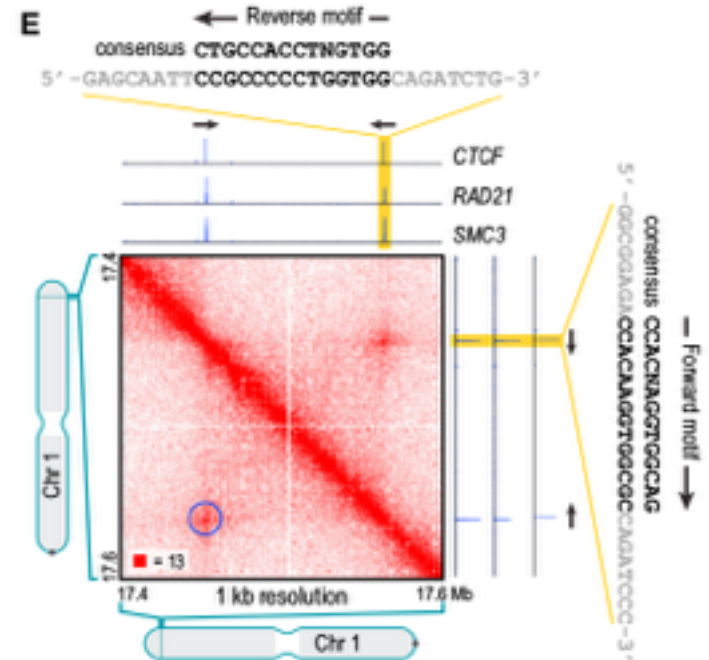


# Protein-mediated loops frequently demarcate contact domain boundaries

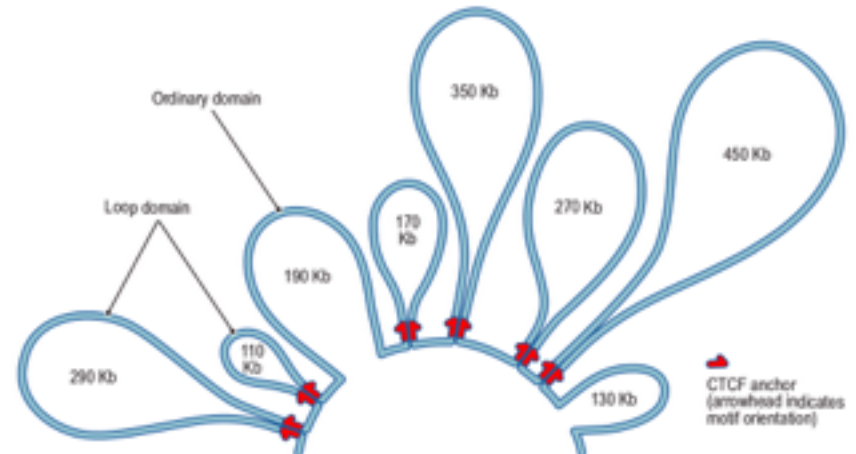
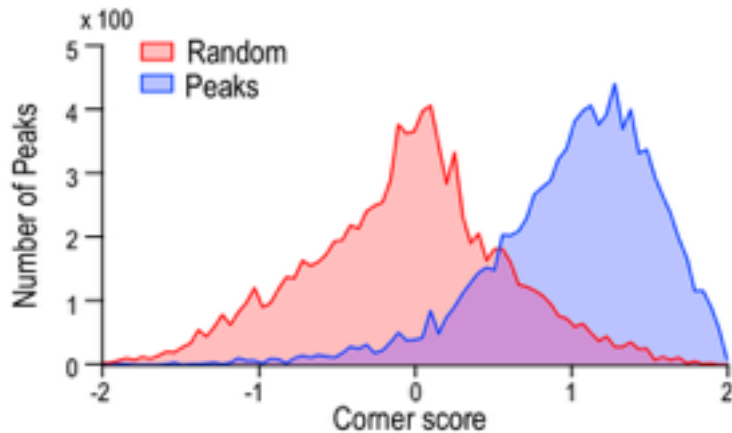
The majority of loops recruit SMC3, RAD21, and CTCF DNA binding proteins



The majority of loops bind CTCF protein pairs in a convergent orientation



Loops frequently demarcate domain boundaries ("corners")





# Summary: *in situ* HiC reveals hierarchical folding in the human genome

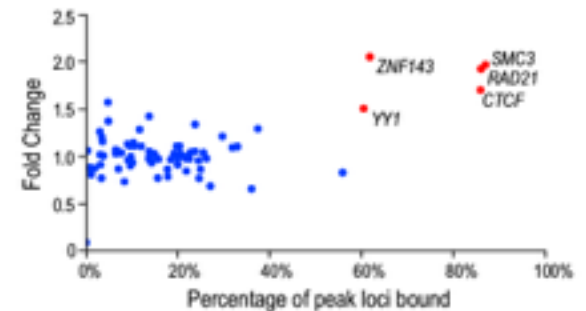
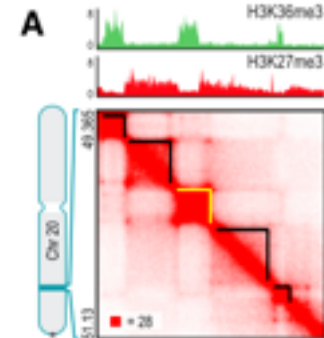
Long-range contact patterns organize the genome into sub-compartments with distinct histone modification patterns

Sub-compartments contain contact domains with internally correlated epigenetic marks

Contact domains are organized by protein mediated loops that frequently demarcate domain boundaries and regulate gene activation

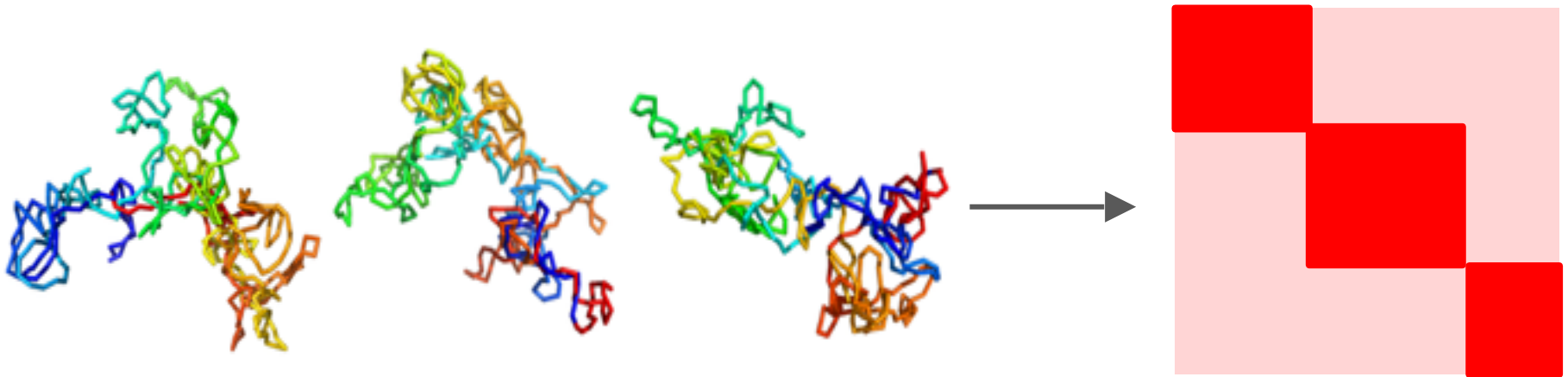


	H3K36me3	H3K27me3	H3K9me3	H3K27me1	H3K27me3	H3A.Z	H3K27ac	H3K4me1	H3K4me2	H3K4me3	H3K79me2	H3K9ac	LaminA/C	NADs	RepG1	RepS1	RepS2	RepS3	RepS4	RepG2
A1	3.5	1.1	1.1	1.4	1.8	3.6	7.8	2.6	4.6	4.5	11.5	7.1	0.7	0.1	0.4	3.1	0.5	0.1	0.2	1.0
A2	2.6	1.0	1.4	1.1	1.8	2.7	4.7	2.1	3.3	2.5	4.3	3.1	0.7	0.4	1.8	2.9	2.0	0.5	0.2	0.7
B1	1.0	1.5	1.1	1.2	1.8	0.9	0.9	1.0	0.9	0.9	1.0	1.0	1.1	1.0	1.3	1.8	2.5	2.1	0.4	0.5
B2	0.9	0.8	1.0	0.8	1.1	0.7	0.6	0.5	0.5	0.8	0.8	0.8	1.7	4.5	0.5	0.1	0.4	1.8	3.7	3.7
B3	0.9	0.9	0.8	0.9	1.0	0.8	0.6	0.5	0.5	0.8	0.9	0.9	1.6	0.0	0.5	0.1	0.4	1.8	3.6	3.3
B4	1.5	0.8	0.8	0.9	2.2	5.3	7.0	1.2	4.5	4.6	6.8	8.5	1.0	2.8	1.5	2.1	2.0	1.8	0.5	0.7

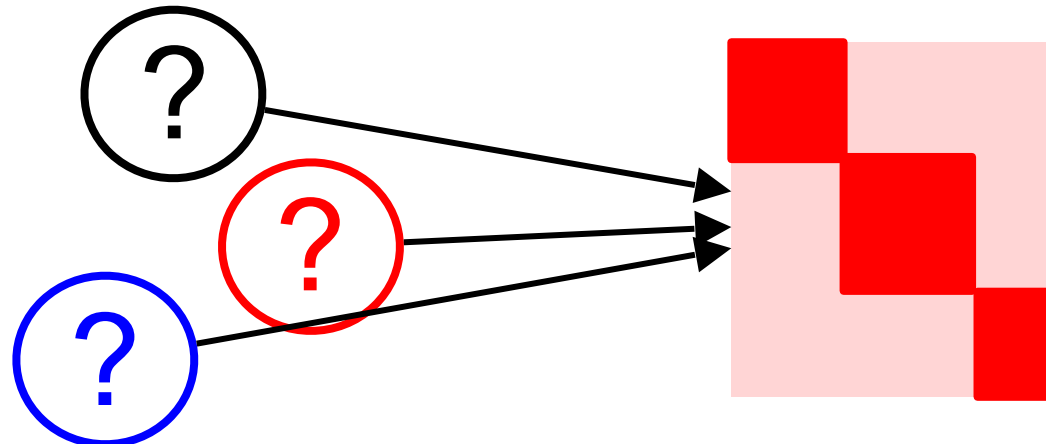


# Critiques

Contact maps from cell populations represent averages over **conformational ensembles**, which may conceal genomic architecture within individual cells

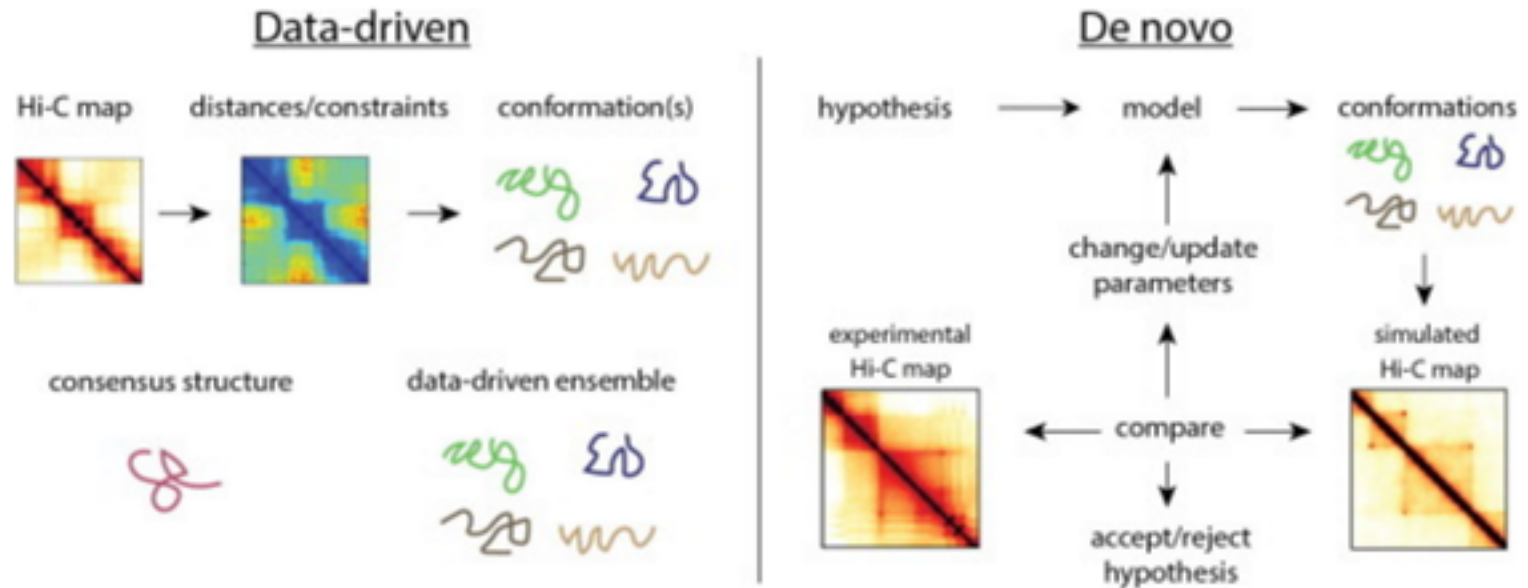


Solving genomic organization from contact maps is highly under-determined: many organizations can produce similar contact probabilities



# Outlook on chromosome conformation analysis: a union between biology, data science, and soft matter physics

Polymer physics modeling combined with HiC data may provide new insights into the emergent physics that governs chromosome organization



# Genome architectures revealed by tethered chromosome conformation capture and population-based modeling

Reza Kalhor<sup>1,2</sup>, Harianto Tjong<sup>1</sup>, Nimanthi Jayathilaka<sup>1,2</sup>, Frank Alber<sup>1</sup> & Lin Chen<sup>1,3,4</sup>

# A quick overview

**Problem:** Low signal to noise ratios in Hi-C compromise ability to detect interactions between chromosomes

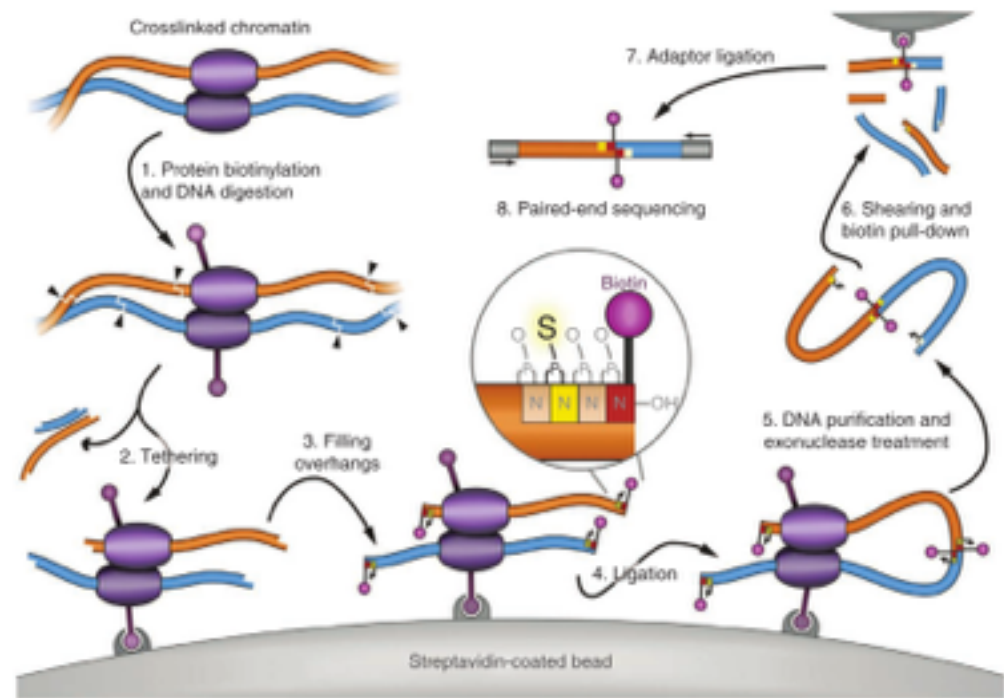
**Solution:** Develop a new conformational capture experiment with reduced signal to noise

**Result:** Can use low-noise data to model genome architectures of cell population

# Tethered conformational capture (TCC)

Immobilization in TCC improves signal to noise ratio

- Can wash away DNA that is not crosslinked
- Immobilization prevents unwanted ligations

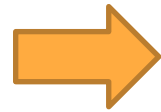
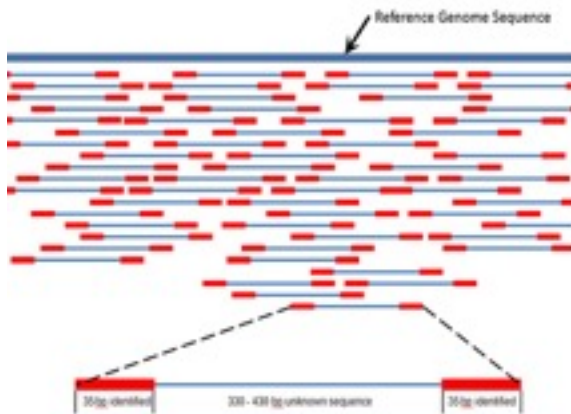


Reza Kalhor *et al*, Nature Biotechnology (2011).

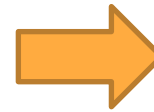


# Contact frequency maps can be created from Hi-C reads

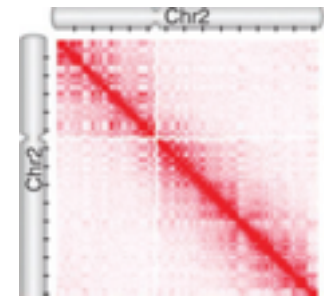
Align reads to human genome



Ligation Matrix



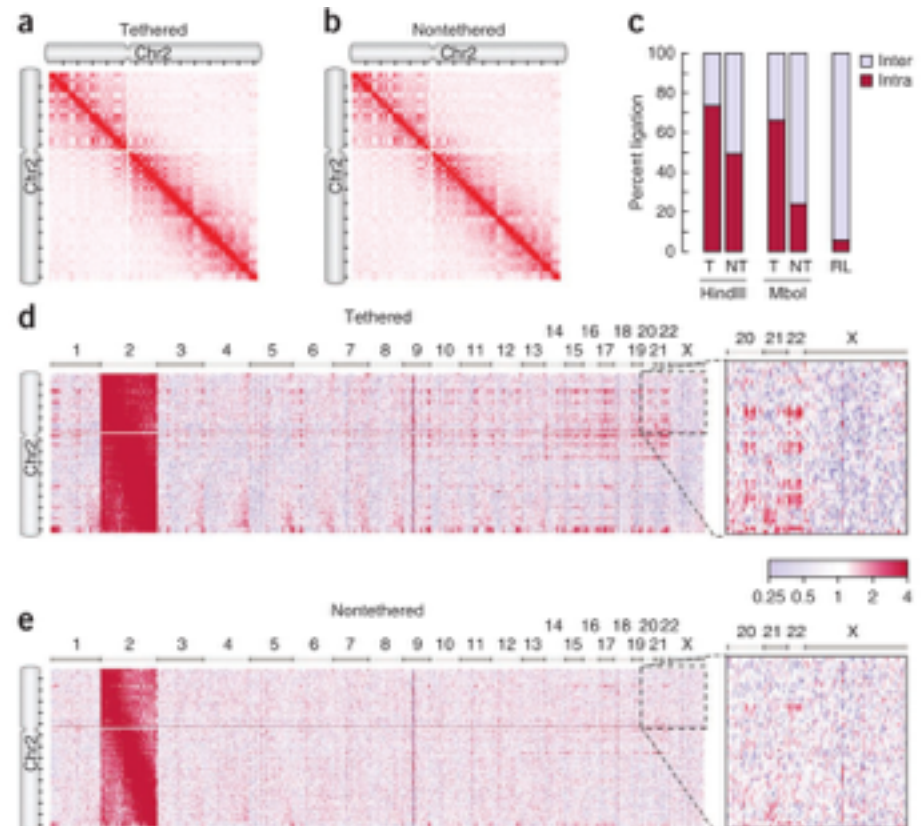
Contact frequency map



A contact frequency map represents the frequency of interactions between regions of chromosomes

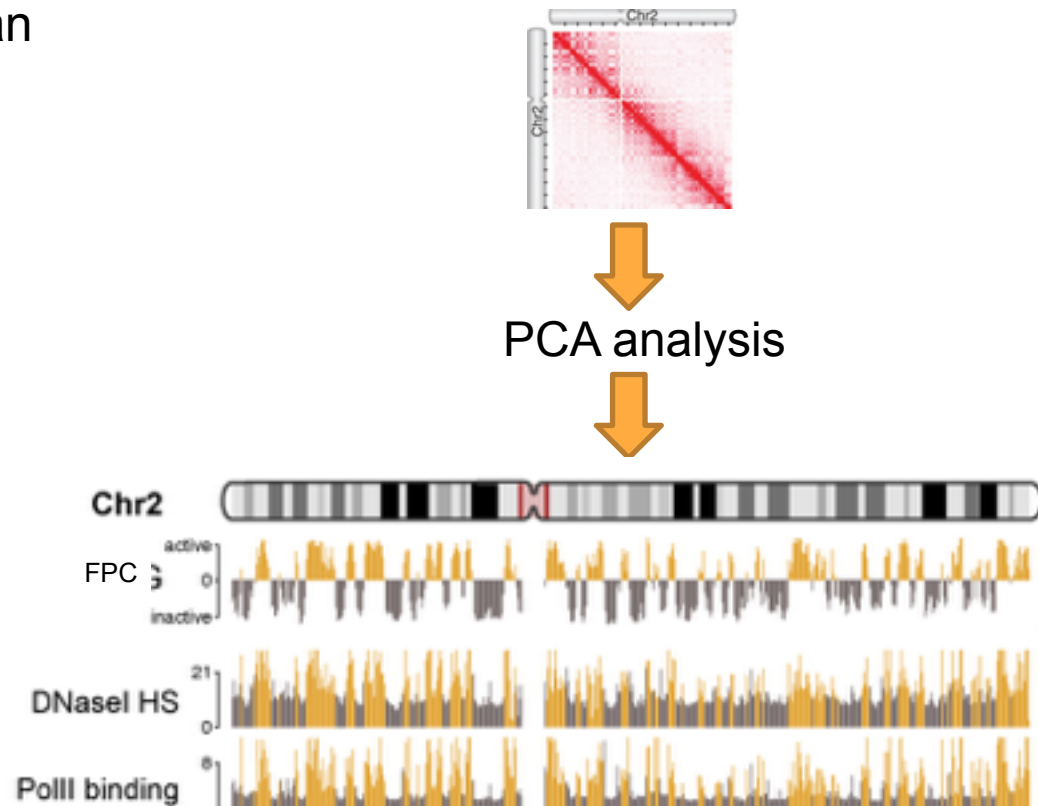
# Improved signal to noise ratio in tethered libraries

- Good correlation between TCC and Hi-C (**a,b**)
- Less interchromosomal contacts in TCC (**c**)
- Can also compute genome wide enrichment maps (**d,e**)



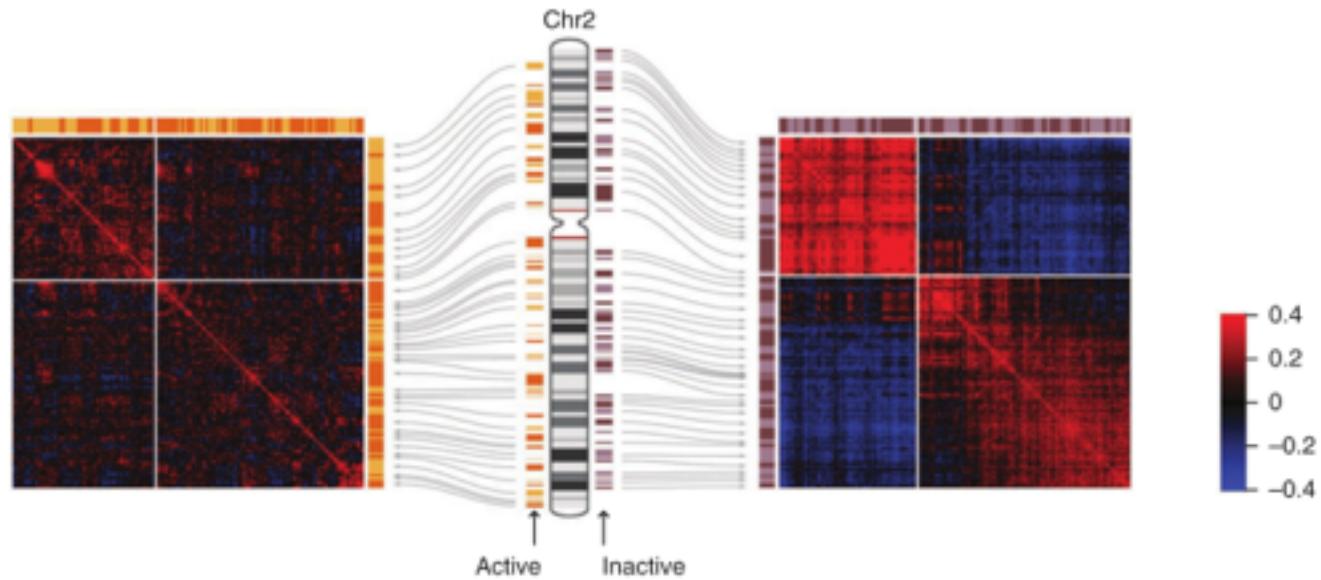
# Principal component analysis can be used to determine active vs inactive regions

- First principal component (FPC) can classify chromosome into active or inactive regions
- Active regions have high gene expression, high DNase sensitivity
- Inactive class have low gene expression, low DNase sensitivity
- Confirmed experimentally



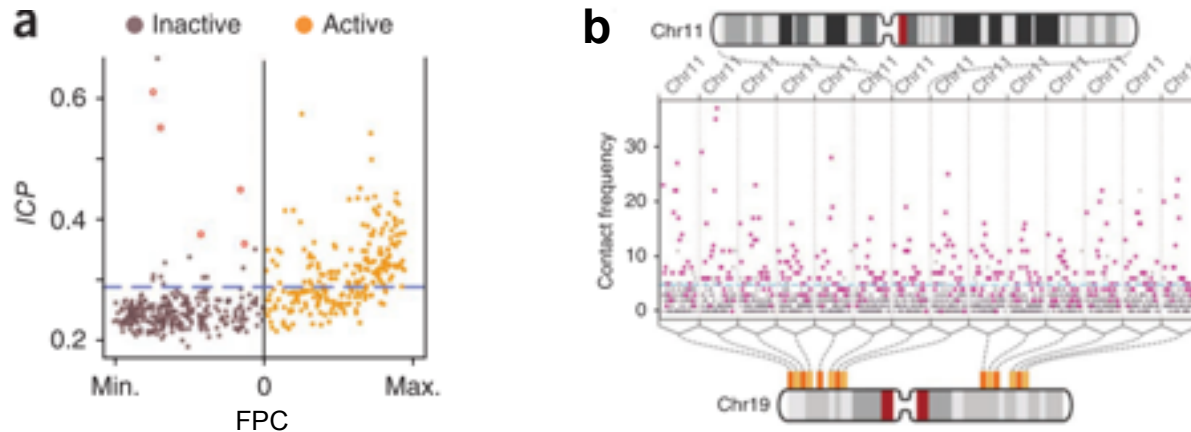
Reza Kalhor *et al*, Nature Biotechnology (2011).

# Intrachromosomal interactions can be separated into active-active and inactive-inactive maps



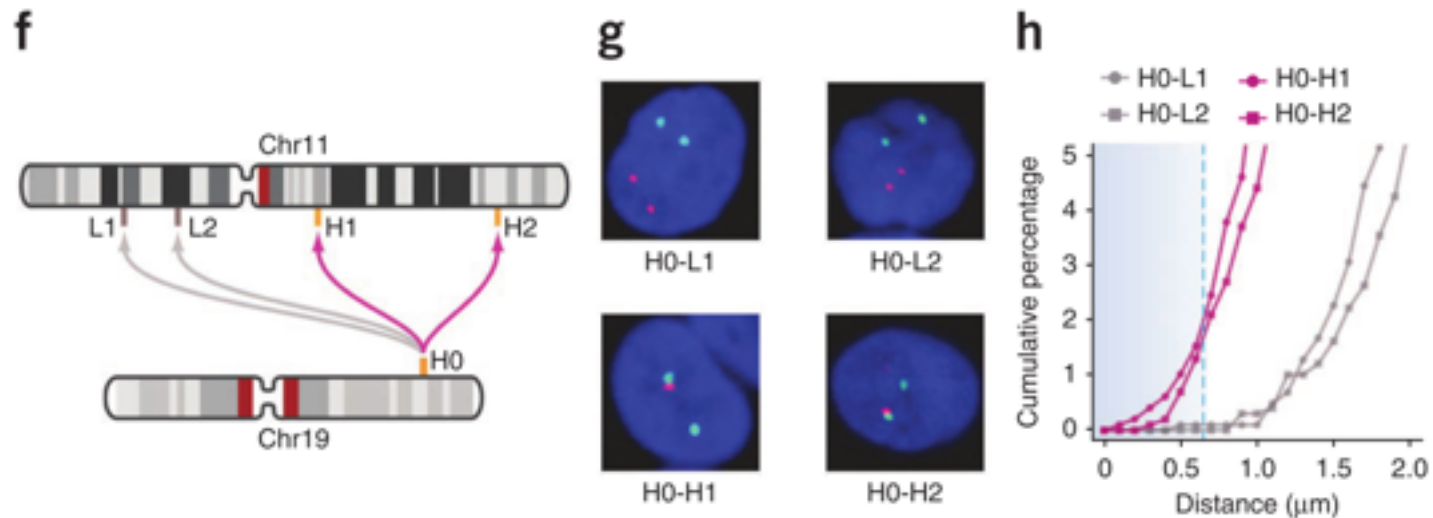
Results suggest that in larger chromosomes inactive regions from opposing chromosomes are largely inaccessible to each other

# Interchromosomal contacts are mediated by the active class and interact indiscriminately



- Interchromosomal contact probability (ICP) represents propensity of a region to form contacts with other chromosomes
- Low contact frequencies indicate that formation of interchromosomal contacts largely governed by spatial accessibility (b)

# 3-D fluorescence *in situ* hybridization agrees with TCC interchromosomal contacts



Active region of chromosome 19 interacts more with active regions on chromosome 11 than with inactive regions on chromosome 11

# Can 3D genome structures be computed from the interchromosomal contact data?

Wide range of frequencies indicate some contacts may only be present in small fraction of cells -> must generate population of structures

To find these structures, they formed an optimization problem

1. Represent structure of chromosome at appropriate resolution
2. Create a scoring function that uses data
3. A method for optimizing the scoring function

# Modeling the 3D organization of the genome

## Represent chromosome

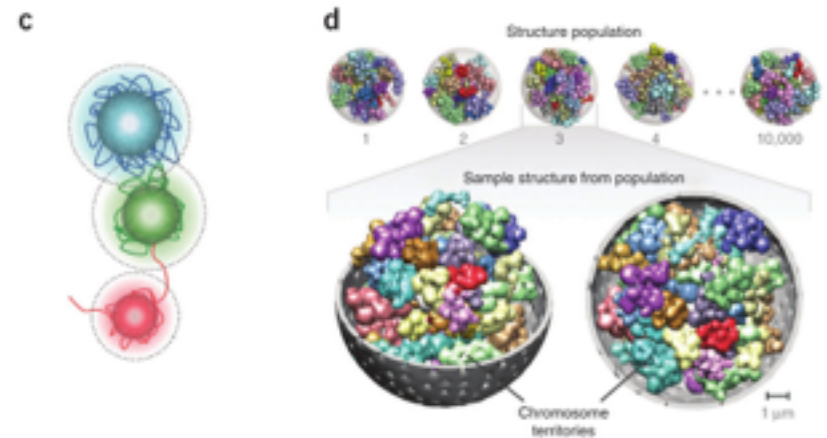
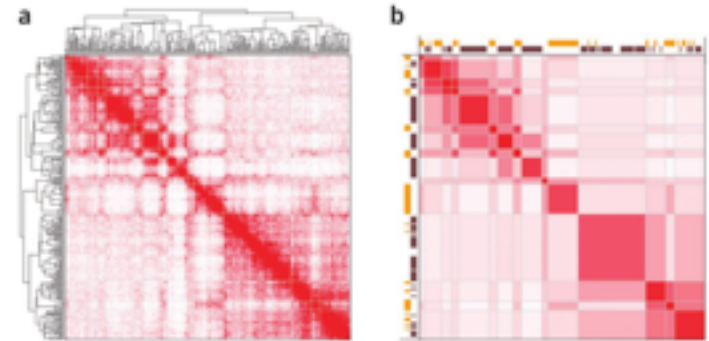
- Divided all chromosomes into 428 blocks (a,b)
- Each block represented as a sphere with two radii (hard and soft) (c)

## Scoring function

- Nuclear volume restraints
- Excluded volume restraints
- Contact restraints

## Optimization

- Start from 10,000 random positions and iterate until score=0
- Utilized Integrated Modeling Platform to solve optimization problem

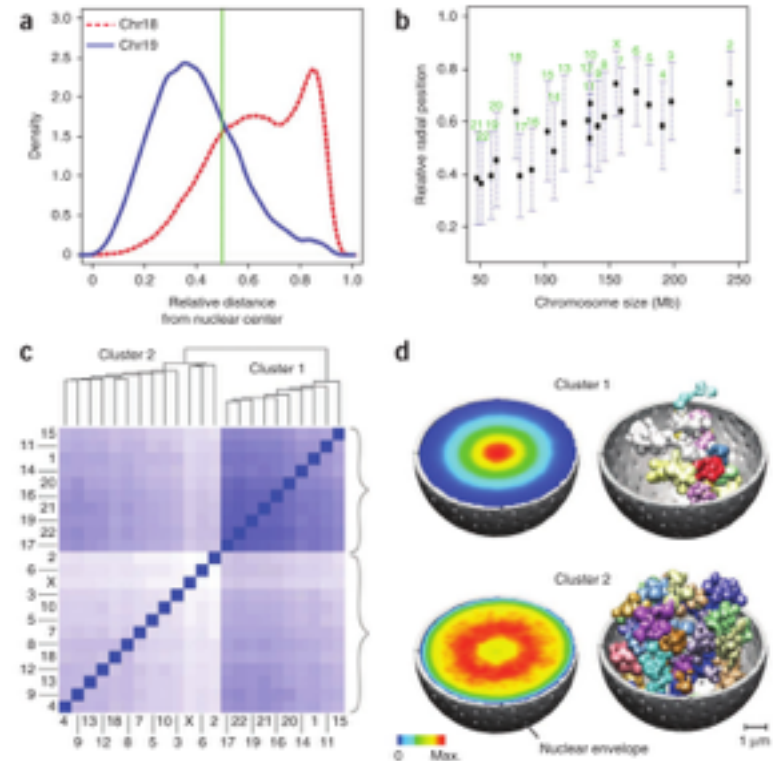


$$S(X_1, \dots, X_M) = \sum_{m=1}^M \sum_{l=1}^{2N} u_{lm}^{\text{mac}} + \sum_{m=1}^M \sum_{l=1}^{(2N-1)} \sum_{j>l}^{2N} u_{ljm}^{\text{exc}} + \sum_{m=1}^M \sum_{l=1}^{N-1} \sum_{j=l+1}^N w_{ljm} u_{ljm}^{\text{con}} = 0$$



# Analysis of computed genome structures

- Large degree of structural variation, contacts only present in small fraction of cells
- On average, only 21% of contacts are shared between any two structures in population
- However radial positions clearly defined and agree with experimental data



Reza Kalhor *et al*, Nature Biotechnology (2011).

# A critique of the paper

## Strengths

- Good job correcting artifacts through computation
- Thorough documentation of computation (80 page supplemental)

## Weaknesses

- More analysis of sources of heterogeneity
- No comparison of genome architecture with Hi-C data
- Weak analysis of results. How will we use this?
- Has not replaced Hi-C as standard for measuring genomic contacts

## Future work

- Examine differences between different cell types
- Incorporate experimental data
- Single cell