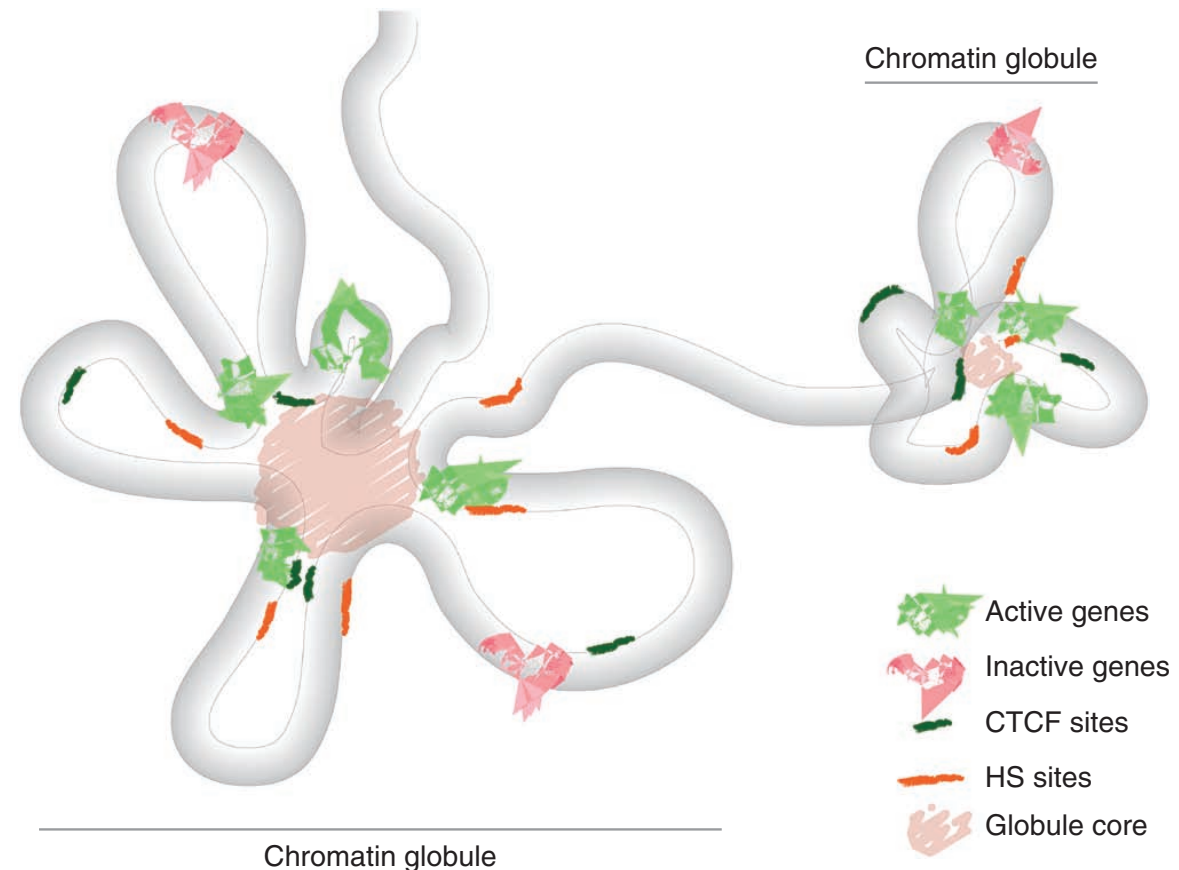
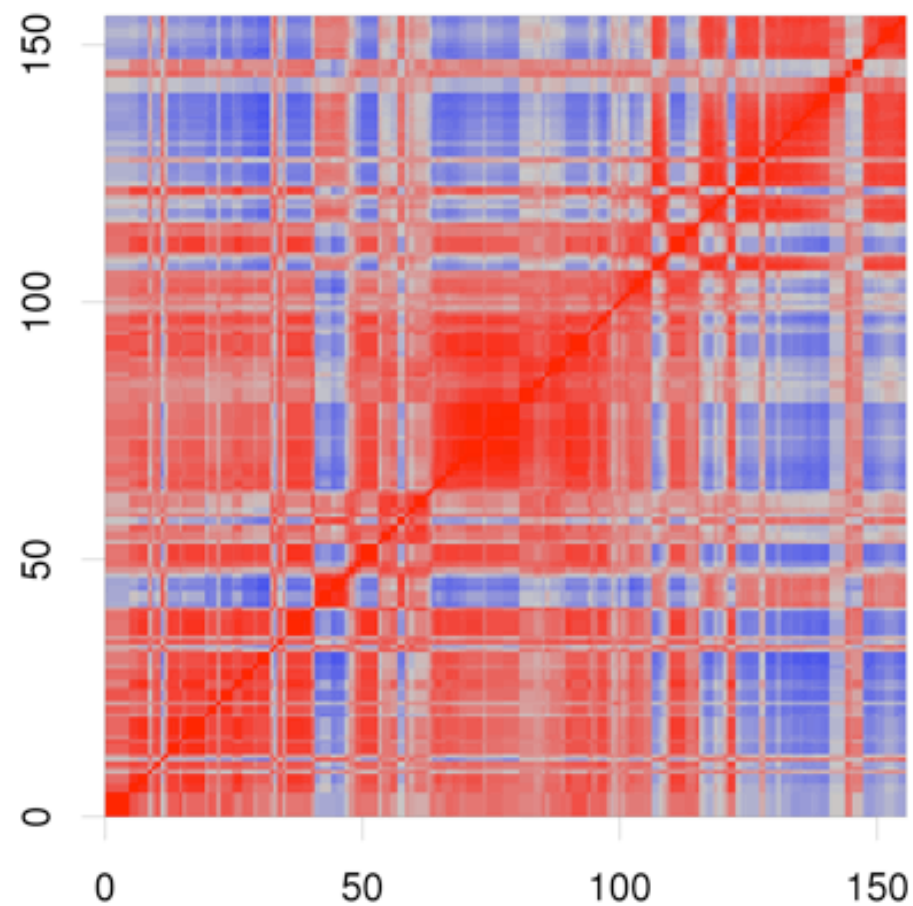


Hi-throughput Chromatin Conformation capture -- Hi-C



Bau 2010

CSAMA 2015, Brixen

18. 06. 2015.

Aleksandra Pekowska

aleksandra.pekowska@embl.de

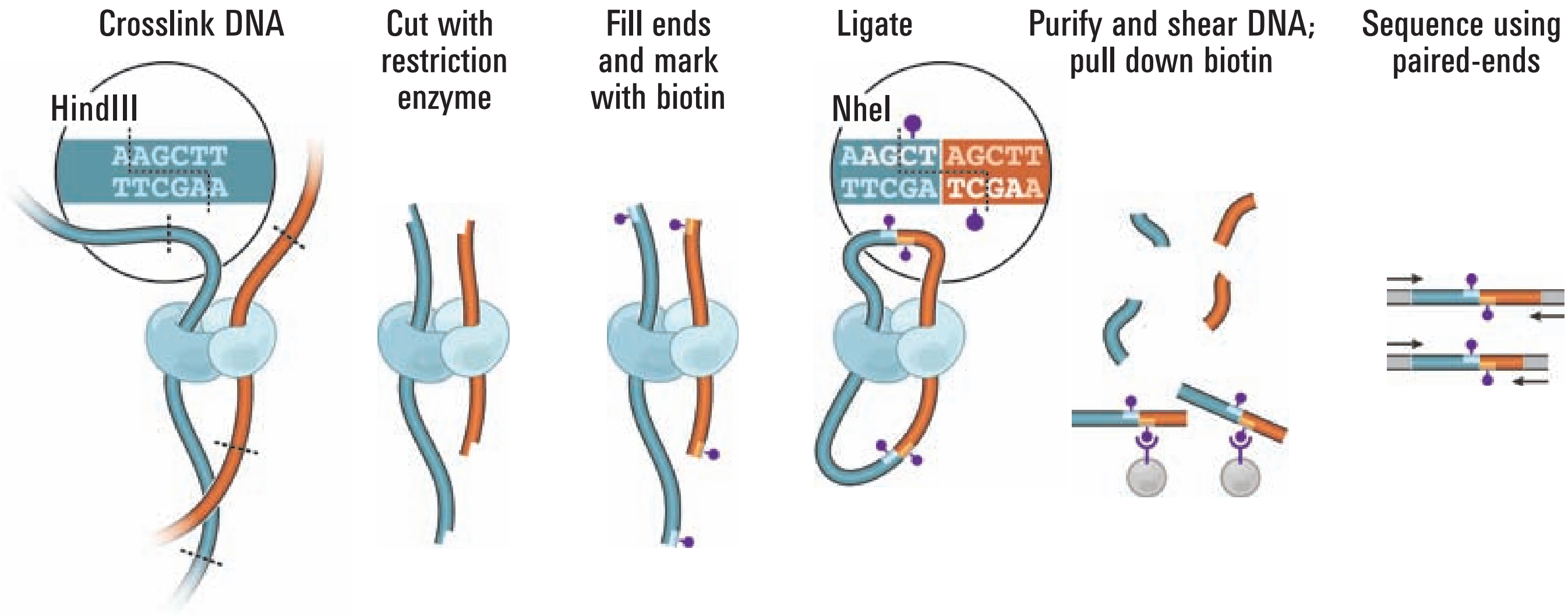
Outline of the lecture

Purpose: introduce basic steps and key considerations in Hi-C analysis

- 1. The Hi-C/TCC method**
- 2. What can we measure with Hi-C?**
- 3. Study design**
- 4. Hi-C analysis workflow:**
 - a. Preprocessing
 - b. Quality controls
 - c. Normalization
 - d. Chromosome-wide analysis
 - e. Identification of local structures - TADs
 - f. Identification of significant interactions

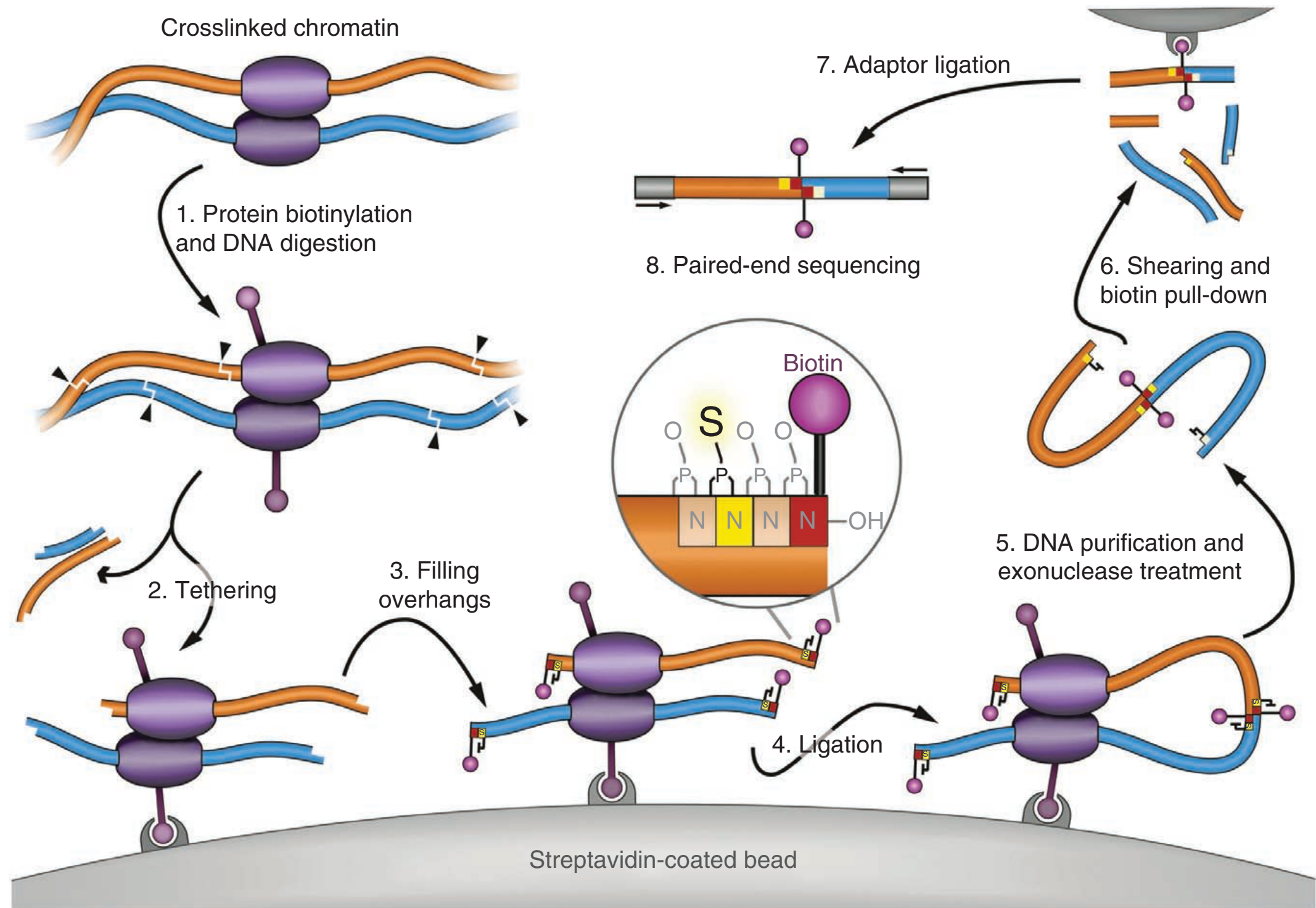
Hi-C and derivatives

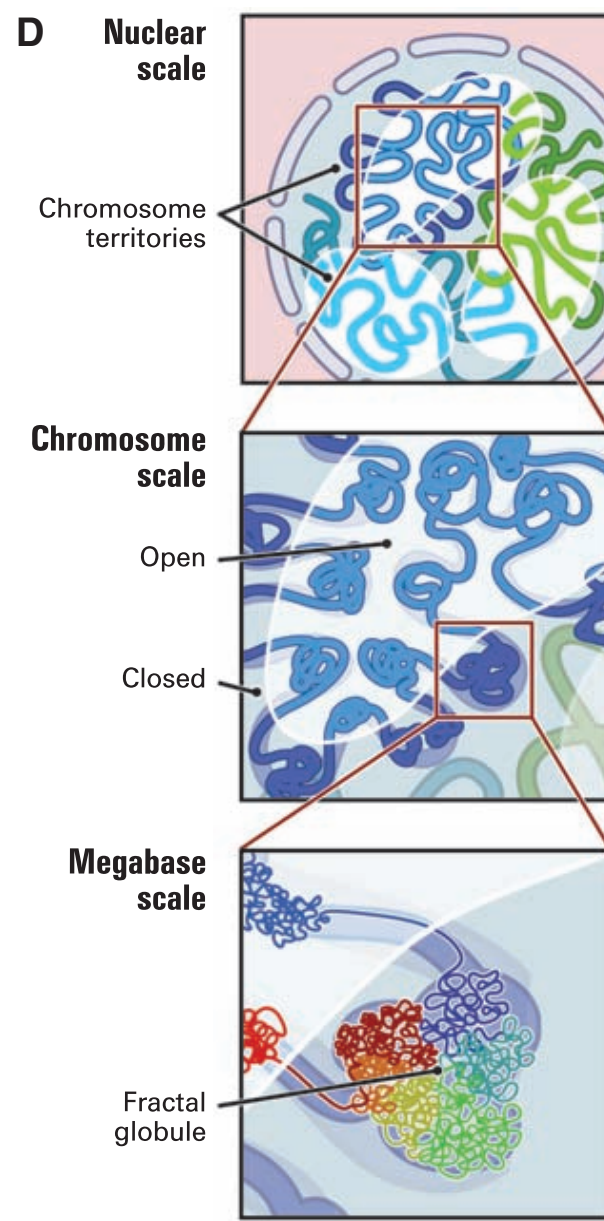
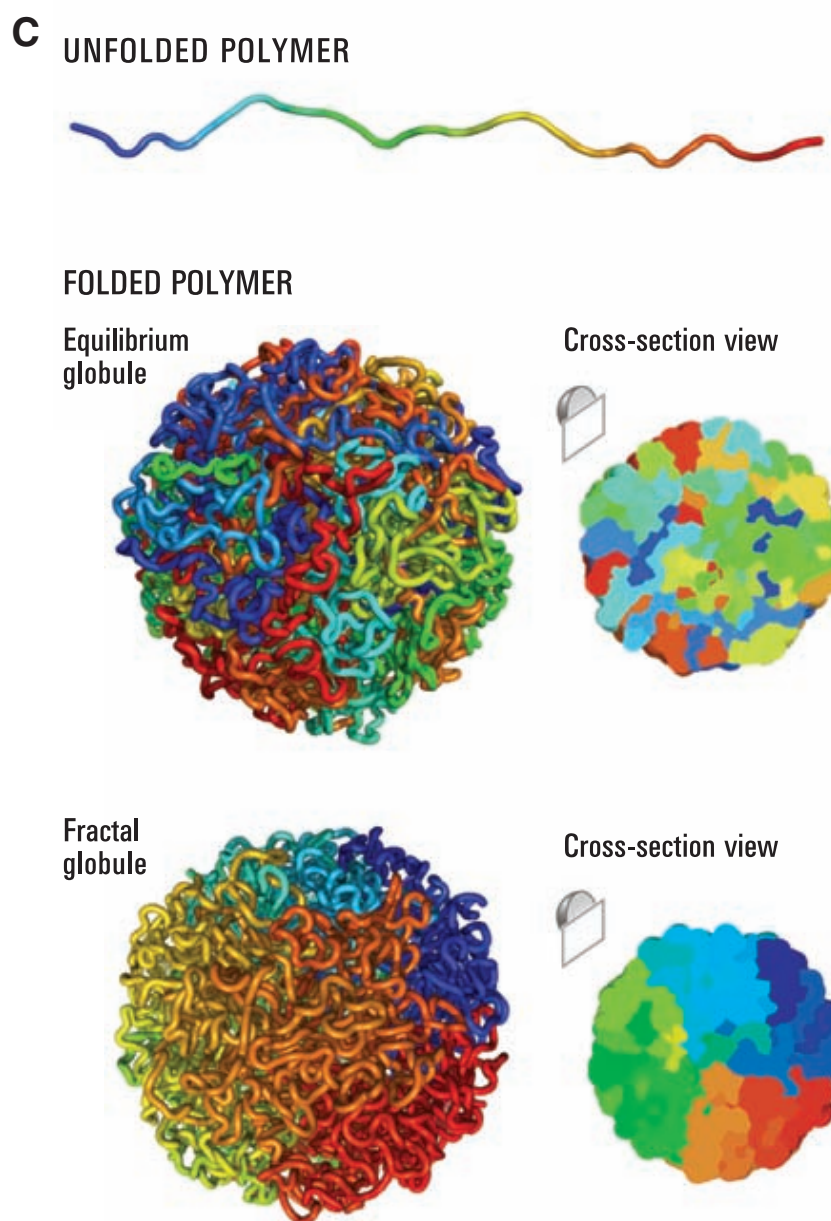
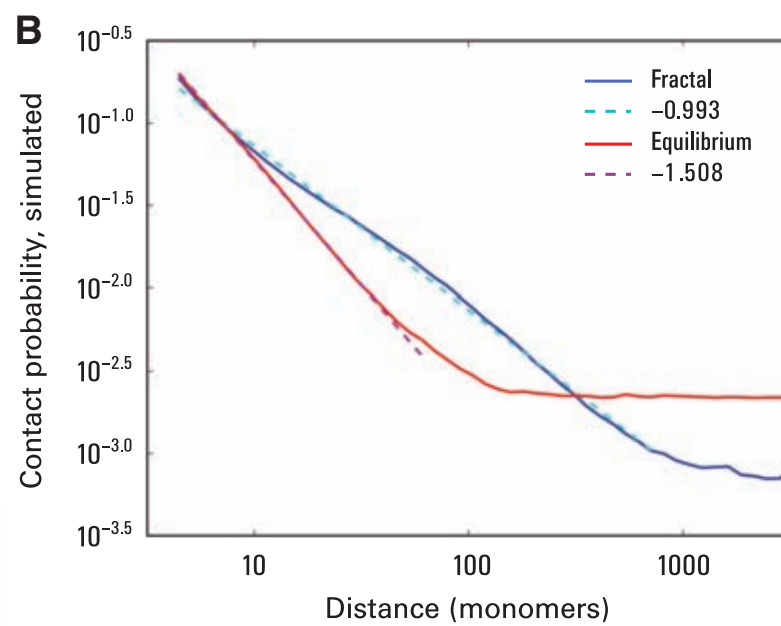
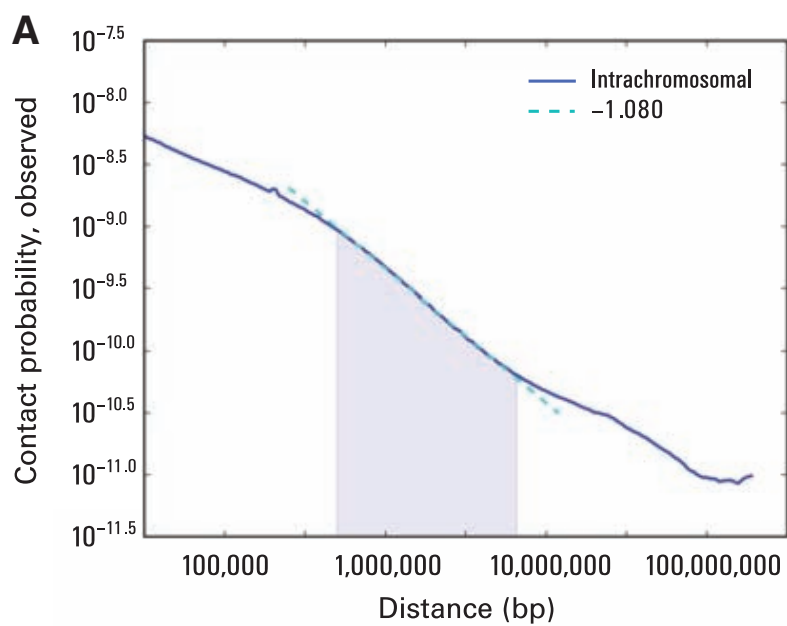
A



Lieberman-Aiden 2009

TCC - tethered chromatin conformation capture



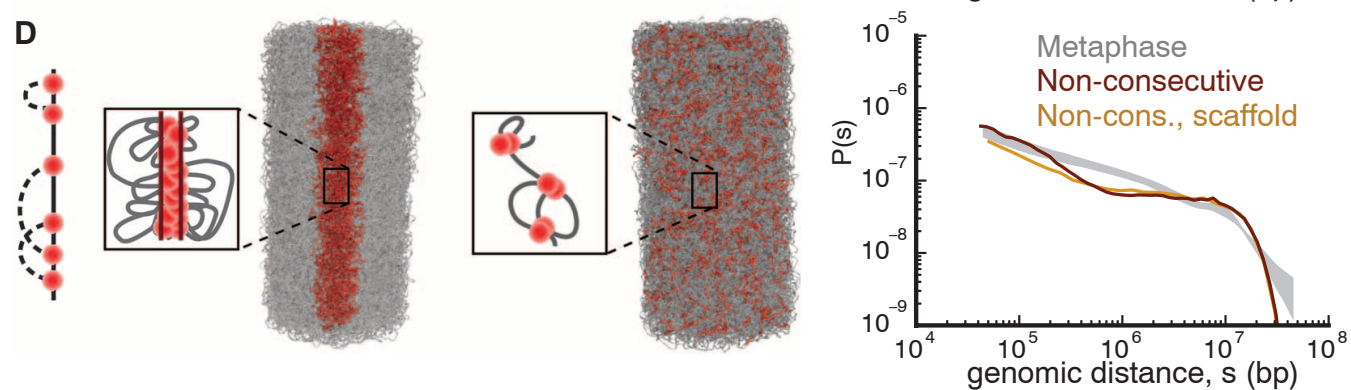
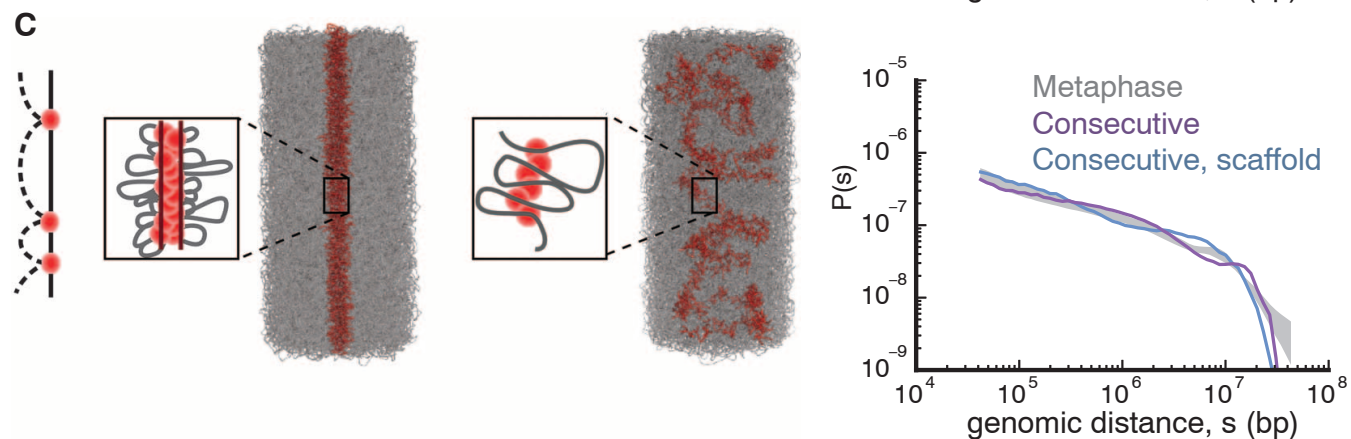
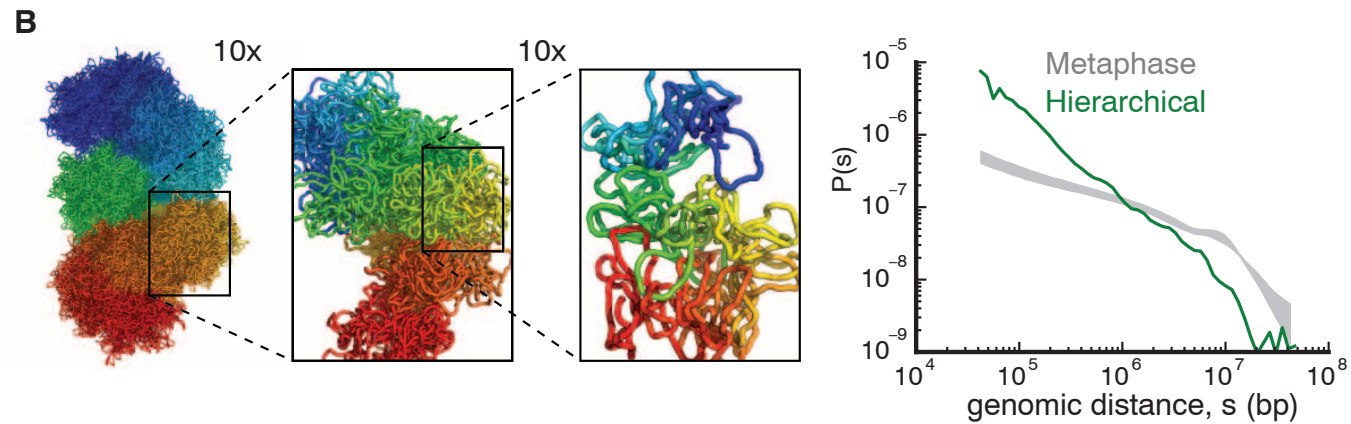
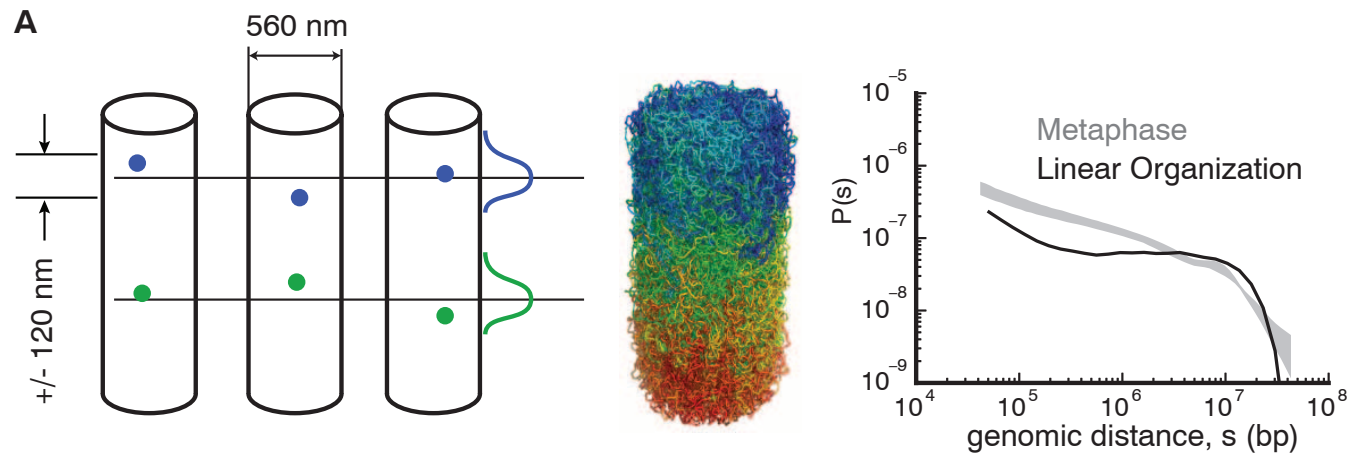


Hi-C methods - what can we learn from them?

Polymer biophysics

General chromatin structure in the interphase

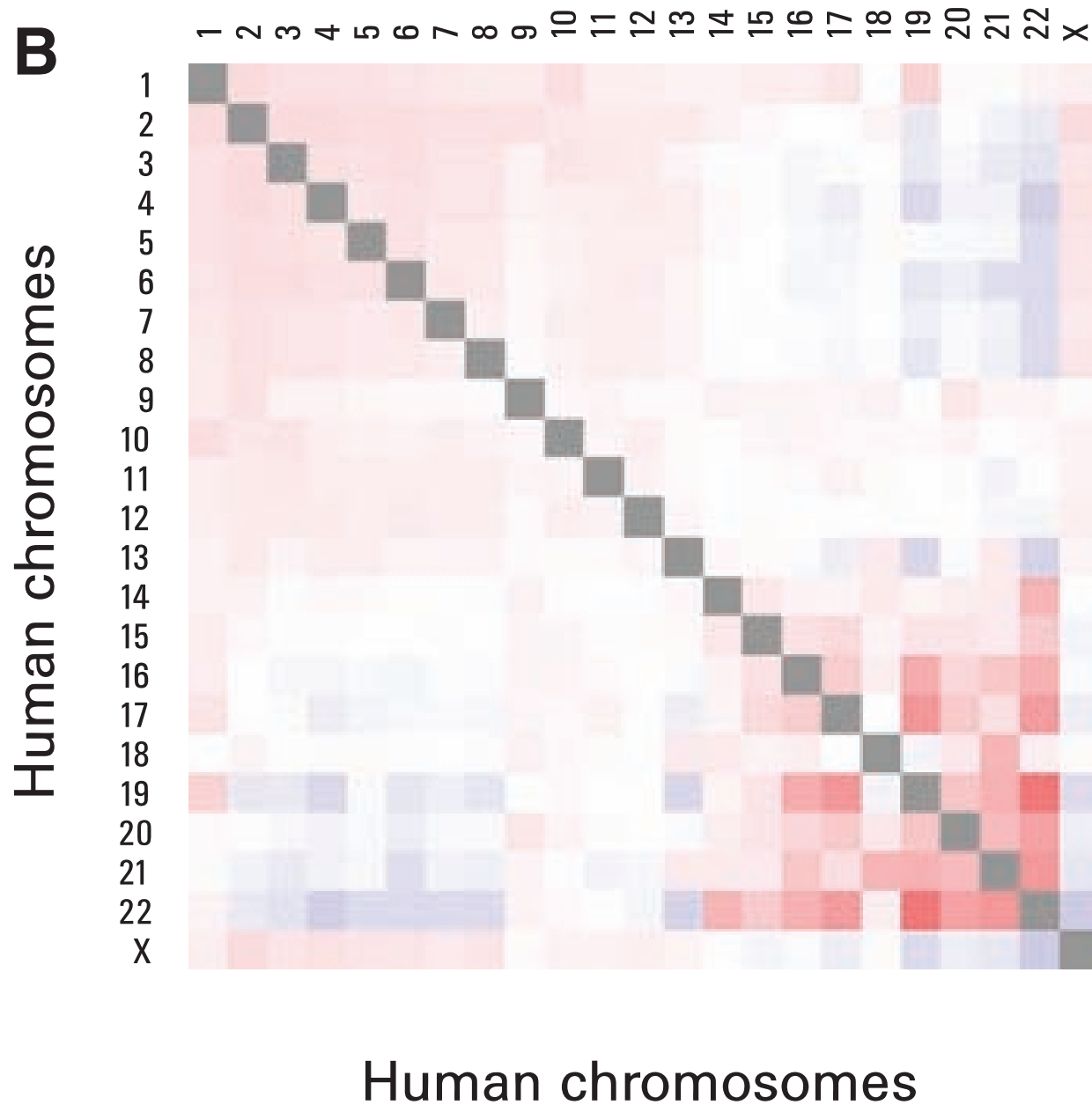
Hi-C methods - what can we learn from them?



Polymer biophysics

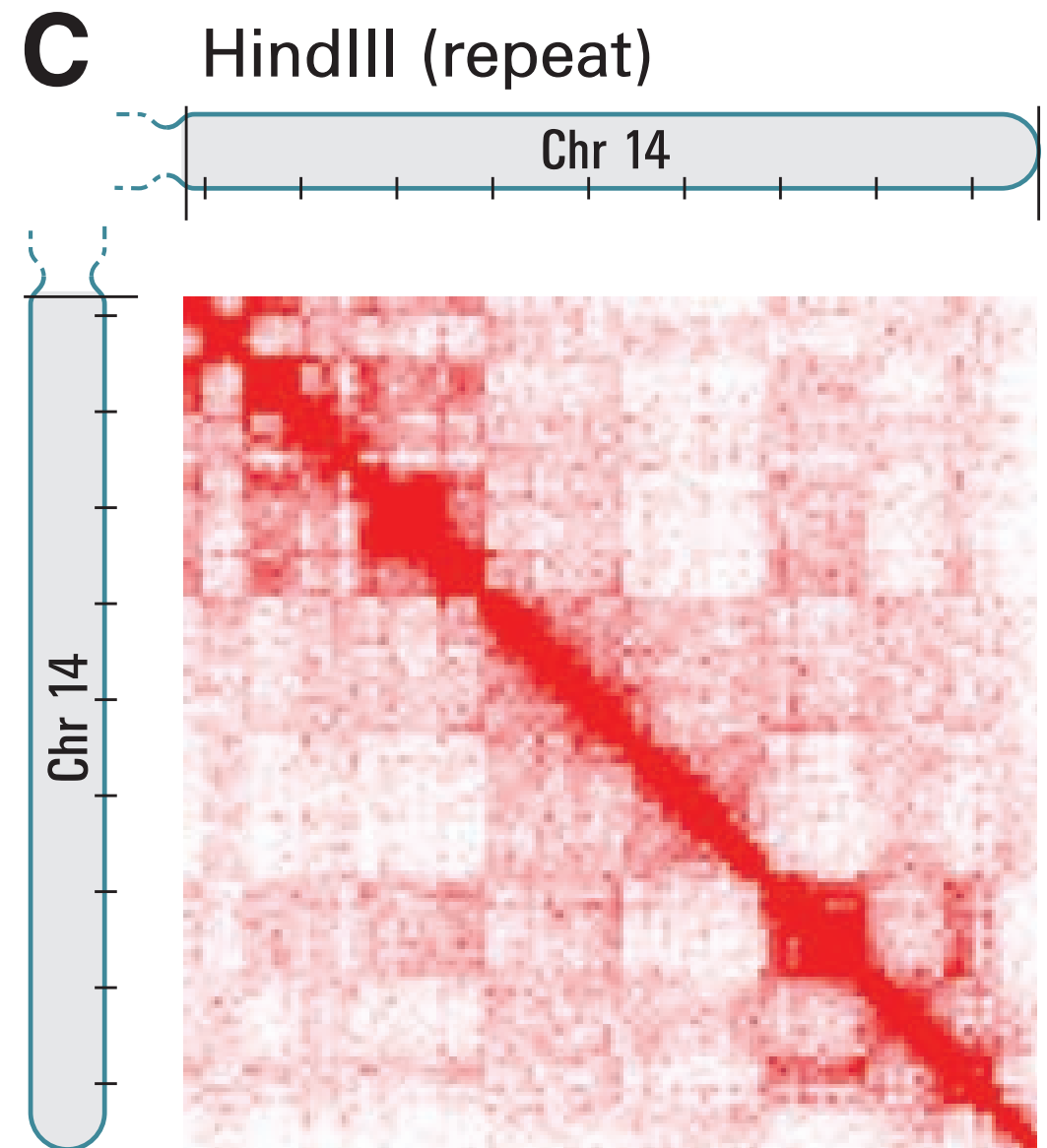
The structure of metaphase chromosomes

Hi-C methods - what can we learn from them?

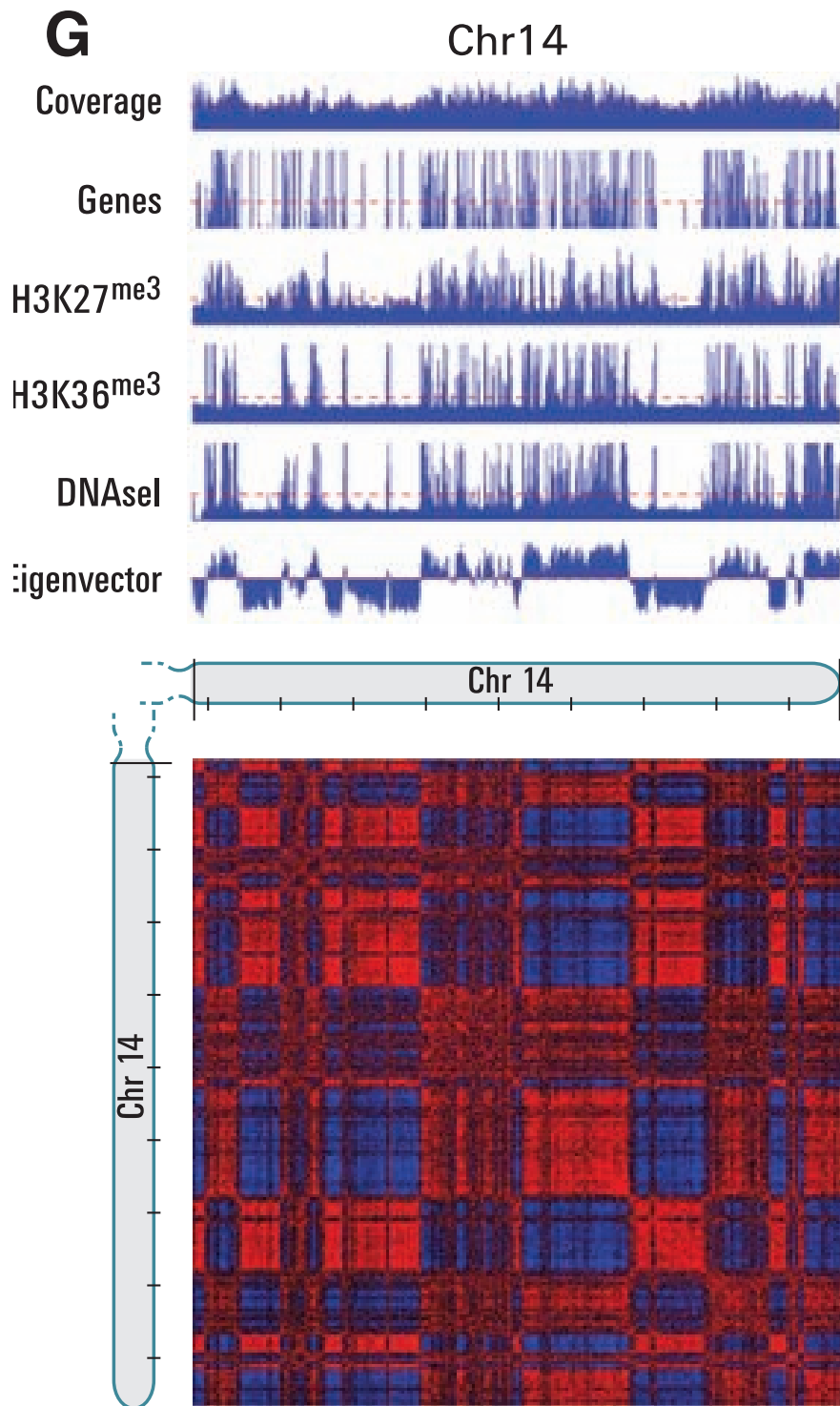


Cross-linking frequencies between sequences genome-wide

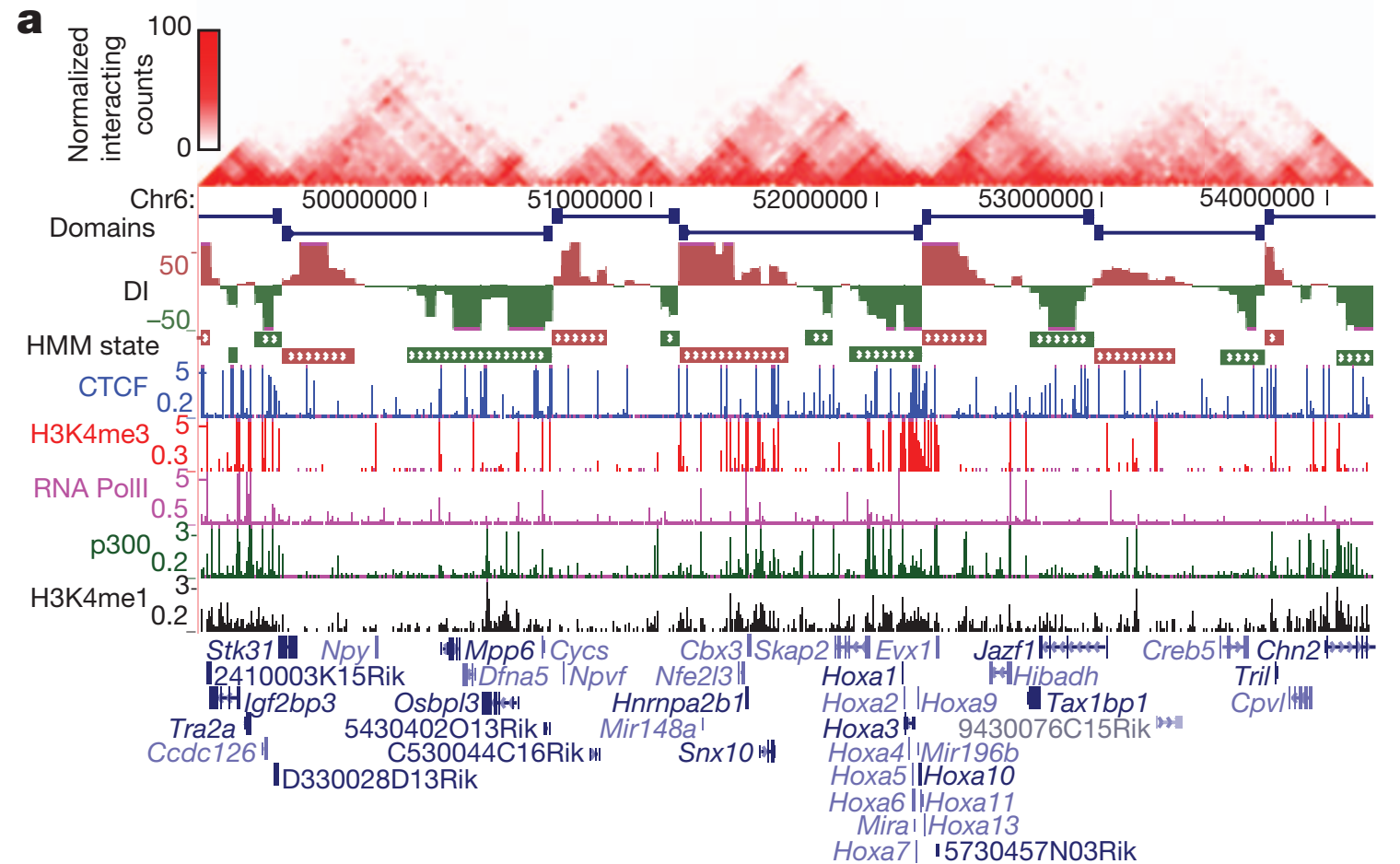
Often referred to as 'interactions'



Hi-C methods - what can we learn from them?

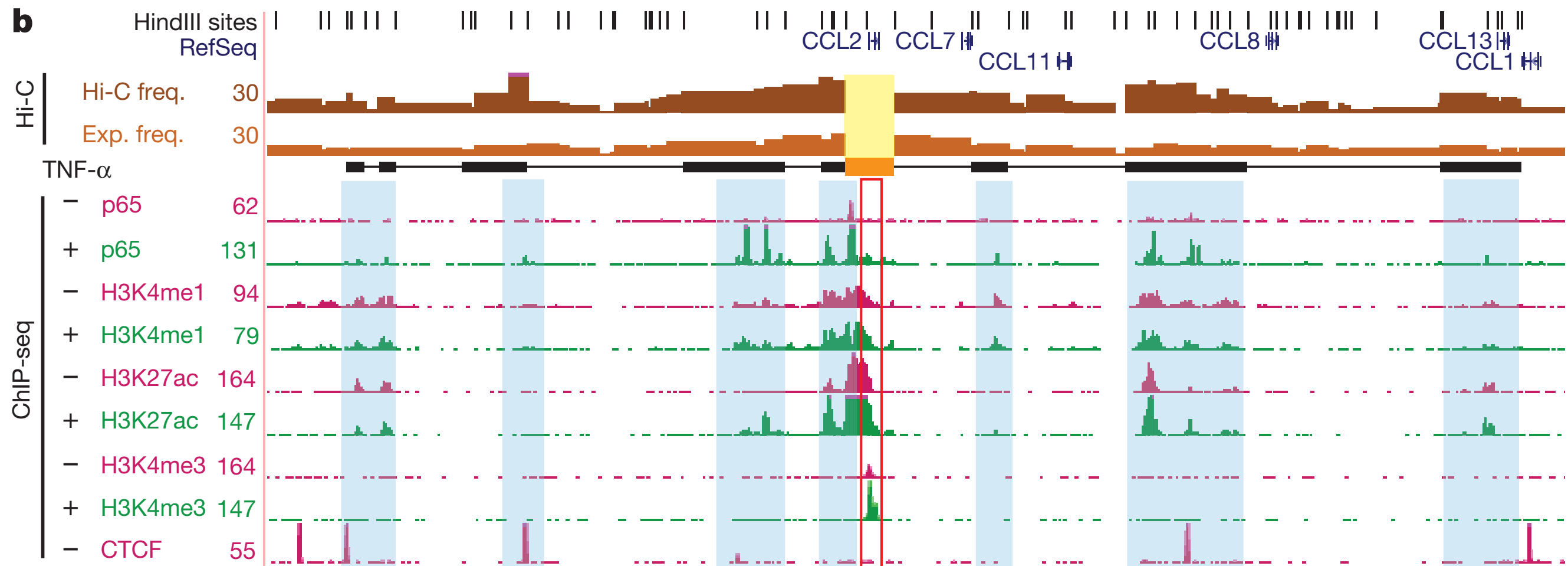


Chromosome compartmentalization



Dixon 2012

Hi-C methods - what can we learn from them?



Jin 2013

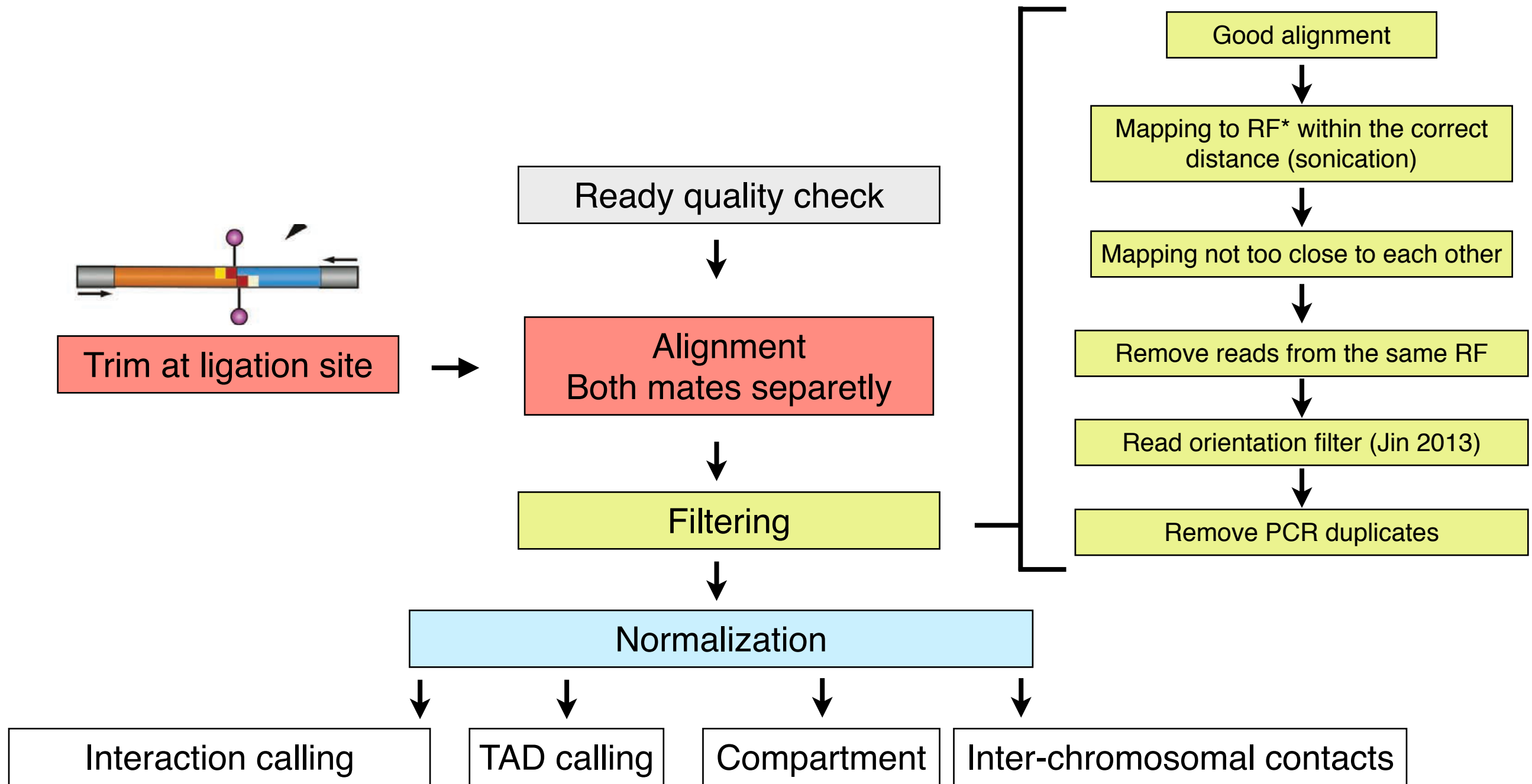
Cis-regulatory element interactions

Study design - sequencing depth 'personal observations'

1 Mb resolution, mammalian genome 1 lane of Hi-Seq per replicate should allow for comparative analysis of inter-chromosomal interactions
(yield ~ 70M usable reads)

The same sequencing depth should allow for attempts in comparative analysis at 10 kb bin level (including 'local' interactions only - up to 1Mb)

Analysis workflow



We can (should) perform normalization

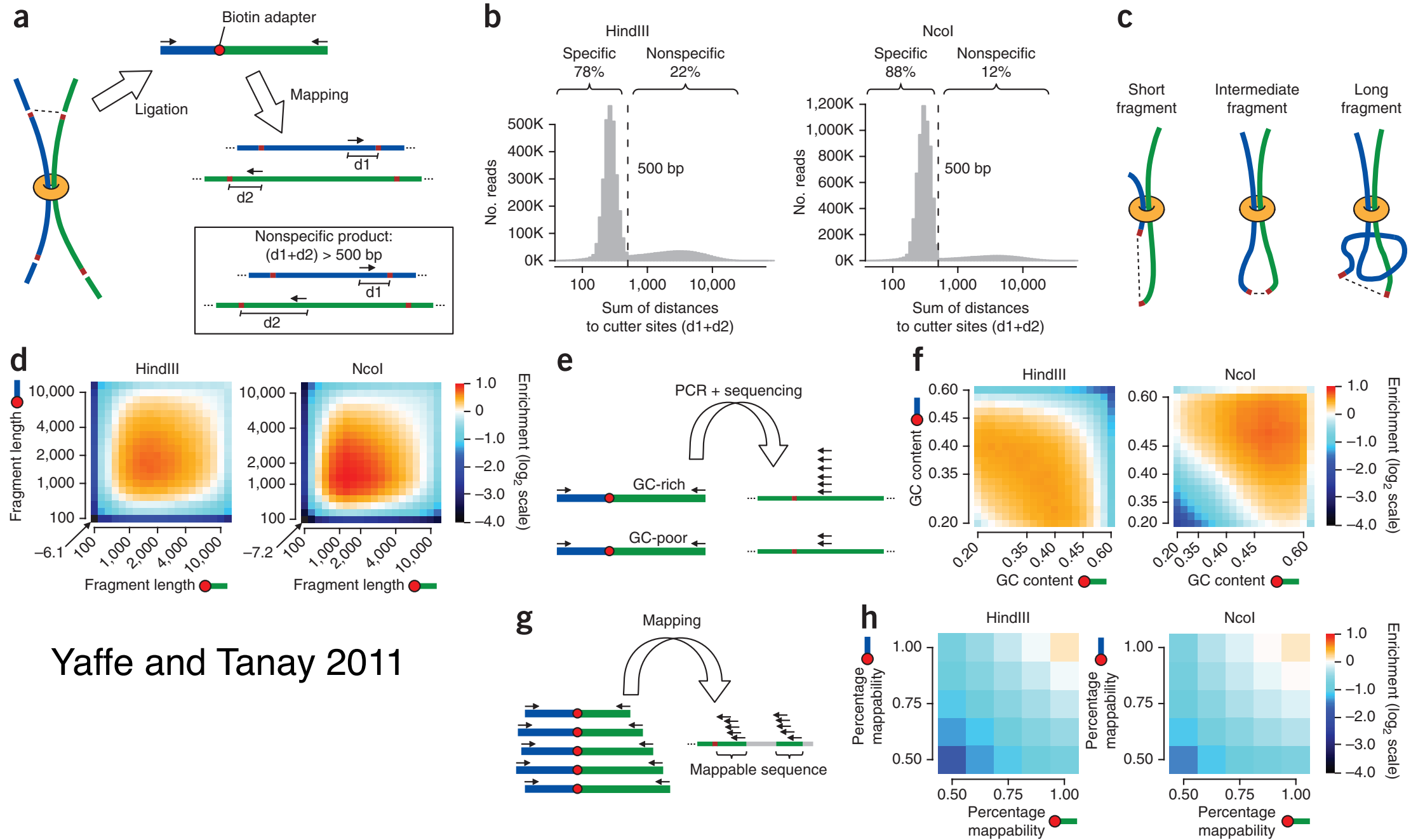
Excellent news: we are genome-wide in this assay!

Restriction Fragments

	0	30	10	15	12	15	10	3	30	6	3	1	0	0	0	0	0	0	0
		0	29	19	3	10	0	1	50	1	3	4	2	0	0	0	0	0	0
			0	1	0	0	0	1	69	0	0	0	0	0	0	0	0	0	0
				0	2	2	3	1	12	0	0	0	0	0	0	0	0	0	0
					0	12	34	20	89	9	0	0	0	0	3	1	1	8	8
						0	32	10	56	0	0	1	1	1	0	0	0	0	0
							0	45	89	0	0	0	0	0	0	0	0	0	0
								0	99	45	30	12	3	1	0	0	0	0	0
									0	60	60	12	67	56	20	13	50	29	30
										0	12	13	4	3	3	3	1	1	0
											0	5	6	2	3	1	1	1	0
												0	13	20	15	0	0	0	0
													0	34	16	2	3	1	0
														0	19	4	1	0	0
															0	2	1	0	1
																0	1	0	0
																	0	3	1
																		0	3
																			0
																			0

Restriction Fragments

1st approach



Yaffe and Tanay 2011

1. Identify sources of biases: RF length, mapability and CG content
2. Normalize

2nd approach - first step

1. Do not try to identify sources of biases but learn their effect from data (coverage)
2. Normalize for the coverage

number of reads
between segments i
and j

Total number of reads

normalized ligation
frequency between
segments i and j

$$f_{ij} = \frac{c_{ij} \left(\sum_{k=1}^{K-1} \sum_{l=k+1}^K c_{kl} \right)}{\left(\sum_{k=1}^K c_{ik} \right) \left(\sum_{k=1}^K c_{kj} \right)}$$

Total number of for
segment i

Total number of for
segment j

2nd approach 'ICE' - complete

1. Do not try to identify sources of biases but learn their effect from data (coverage)
2. Normalize for the coverage in an iterative fashion

Observed

Biases

$$O_{ij} = B_i B_j T_{ij} \text{ True}$$

'relative contact probabilities'

$$\sum_{i=1, |i-j|>1}^N T_{ij} = 1$$

* diagonal and 1st off-diagonal are removed
additional filtering required

How does it work algorithmically?

“We start by creating a working copy of the matrix O_{ij} , denoted W_{ij} as the iterative process gradually changes this matrix to T_{ij} .

We initialize the iterative procedure by setting each element of the vector of total biases B to 1. We begin each iteration by calculating the coverage

$$S_i = \sum_j W_{ij}$$

Next, additional biases ΔB_i are calculated by renormalizing S_i to have the unit mean

$$\Delta B_i = S_i / \text{mean}(S_i).$$

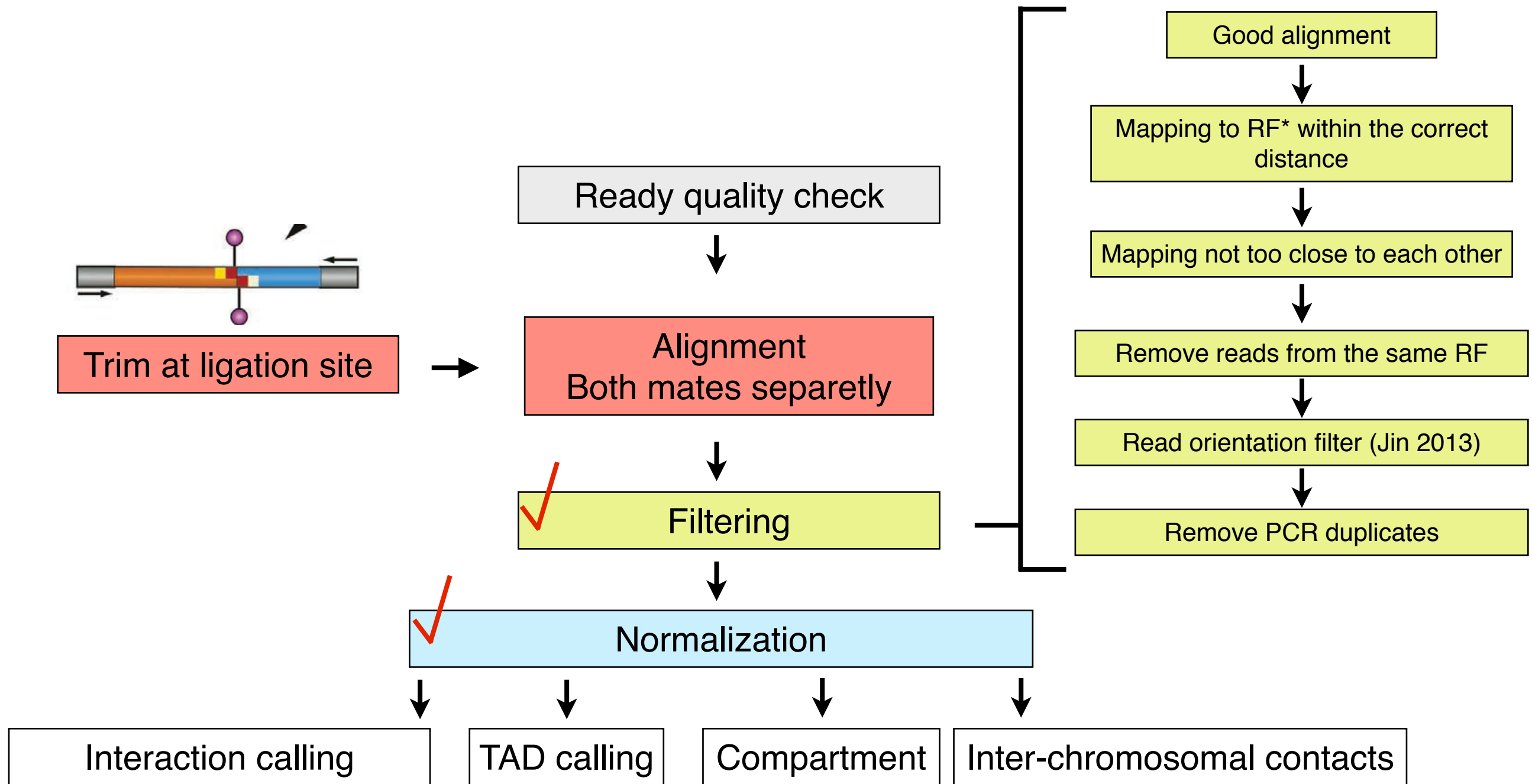
We then

$$W_{ij} / \Delta B_i \Delta B_j \text{ for all } (i,j)$$

and update the total vector of biases by multiplying by the additional biases.

Iterations are repeated until the variance of the additional biases becomes negligible “

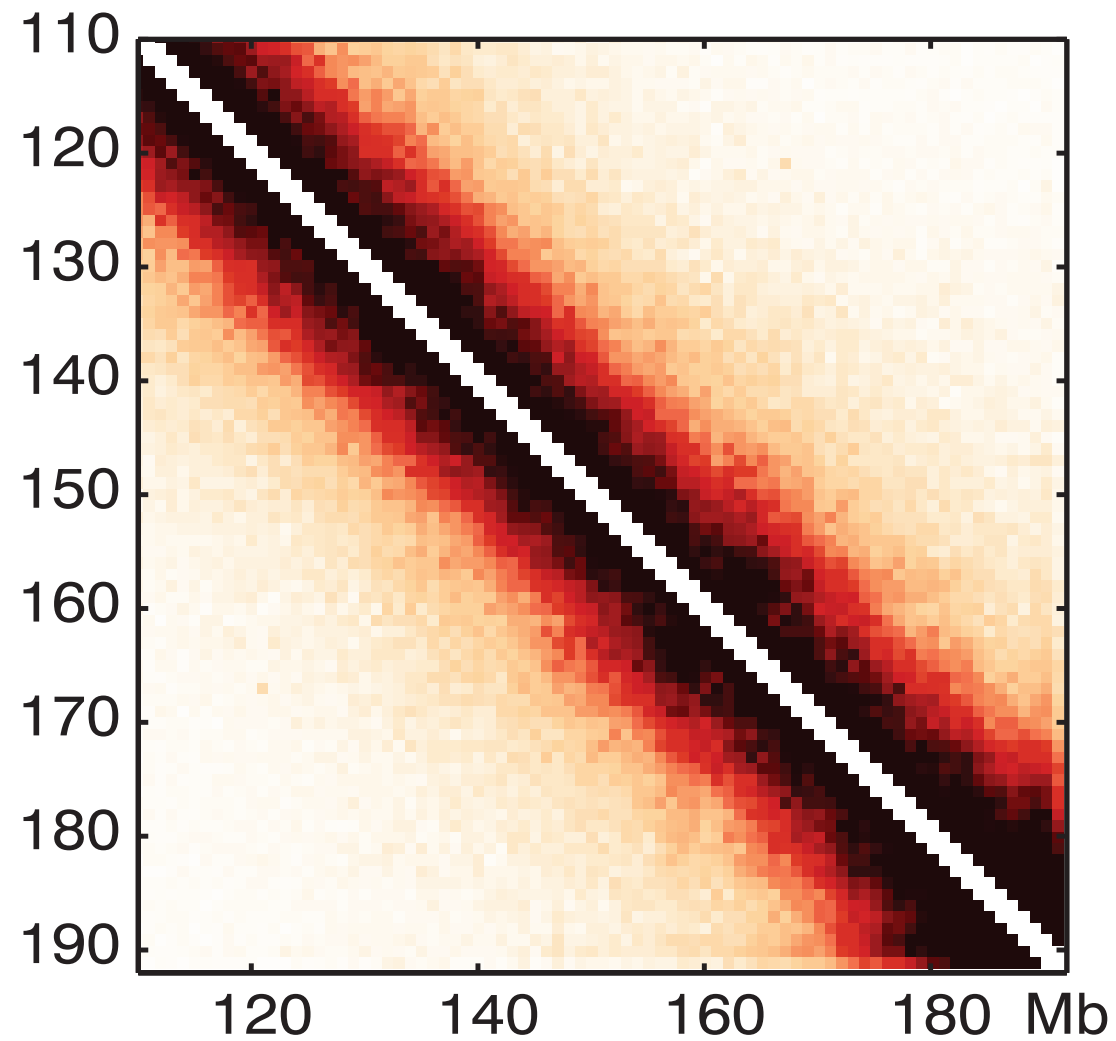
Analysis workflow



Isolation of interactions

Random collisions

Uniform - visualization

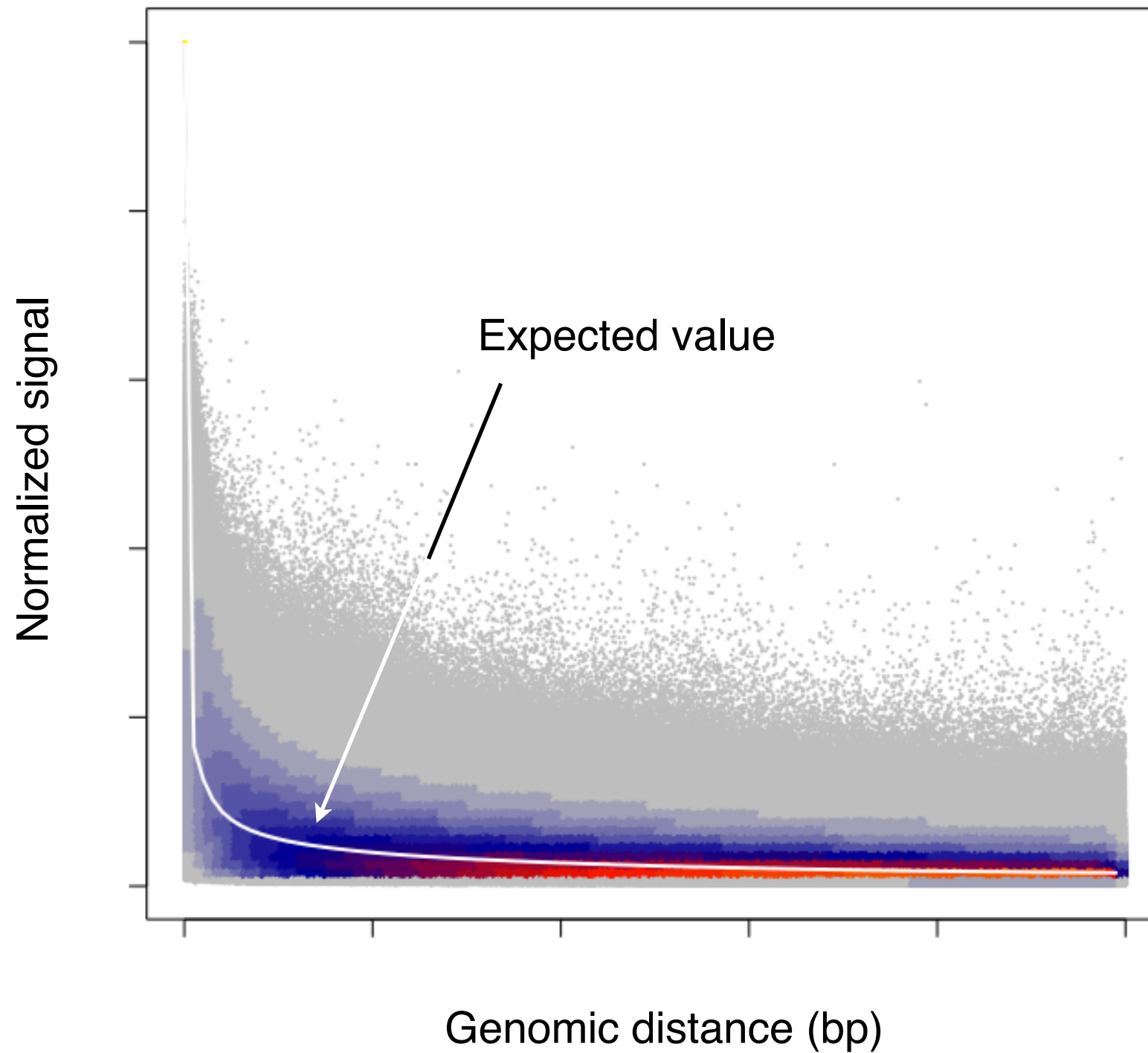


**Interactions occurring just
because chromatin is a
biopolymer and folds**

Naumova 2013

Isolation of interactions

Random collisions - expected interaction strength at a particular distance



Isolation of interactions

LETTER

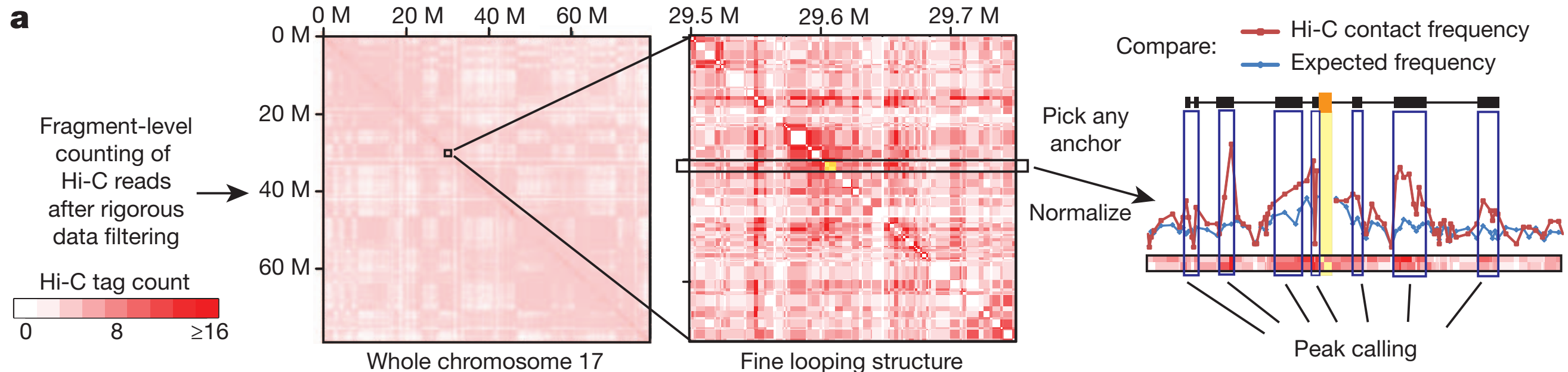
doi:10.1038/nature12644

A high-resolution map of the three-dimensional chromatin interactome in human cells

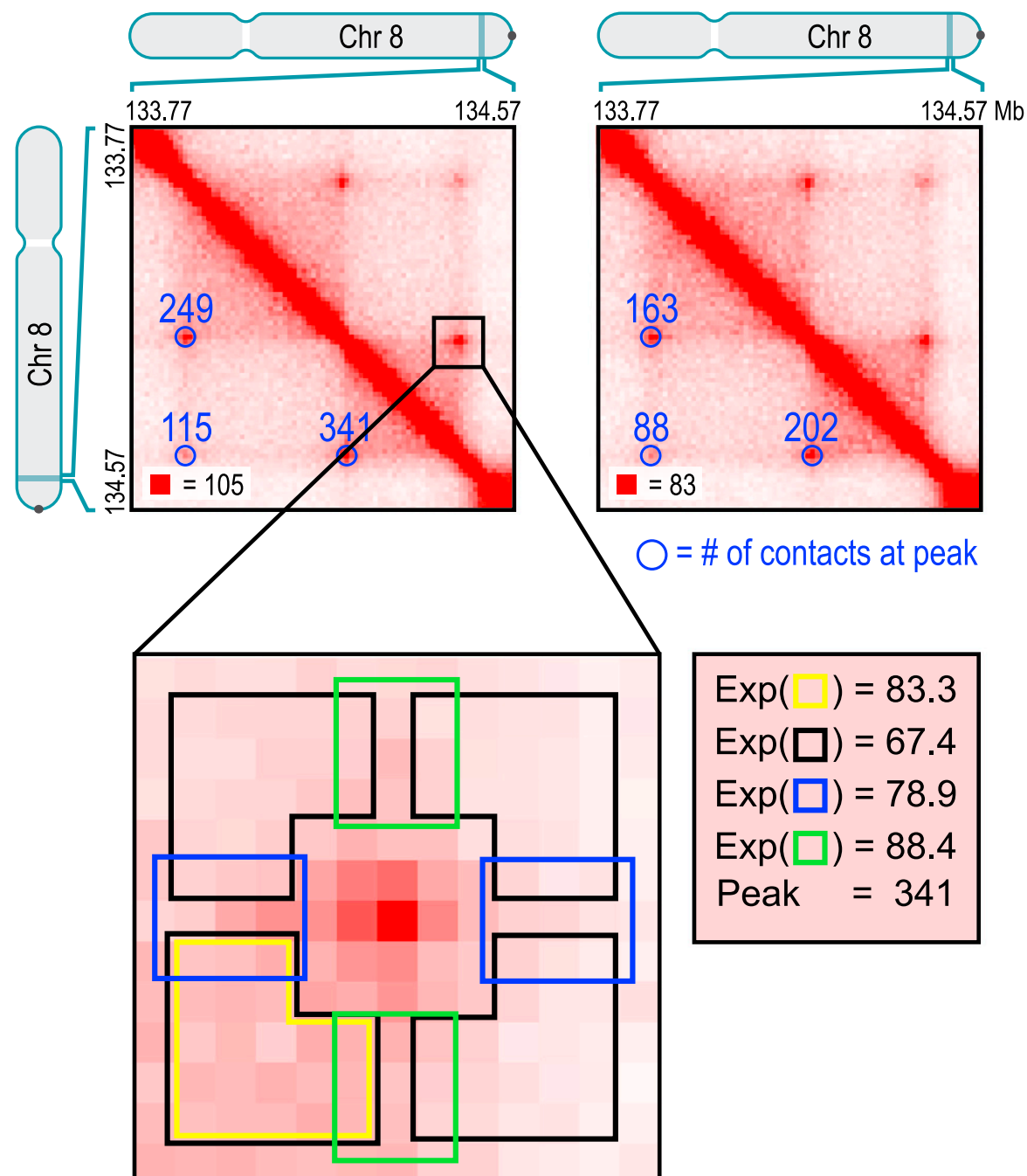
Fulai Jin^{1*}, Yan Li^{1*}, Jesse R. Dixon^{1,2}, Siddarth Selvaraj^{1,3}, Zhen Ye¹, Ah Young Lee¹, Chia-An Yen¹, Anthony D. Schmitt^{1,4}, Celso A. Espinoza¹ & Bing Ren^{1,5}

Use negative binomial to assess for each interaction whether its strength is unexpectedly high given the:

- biases
- distance
- additional signal strength threshold



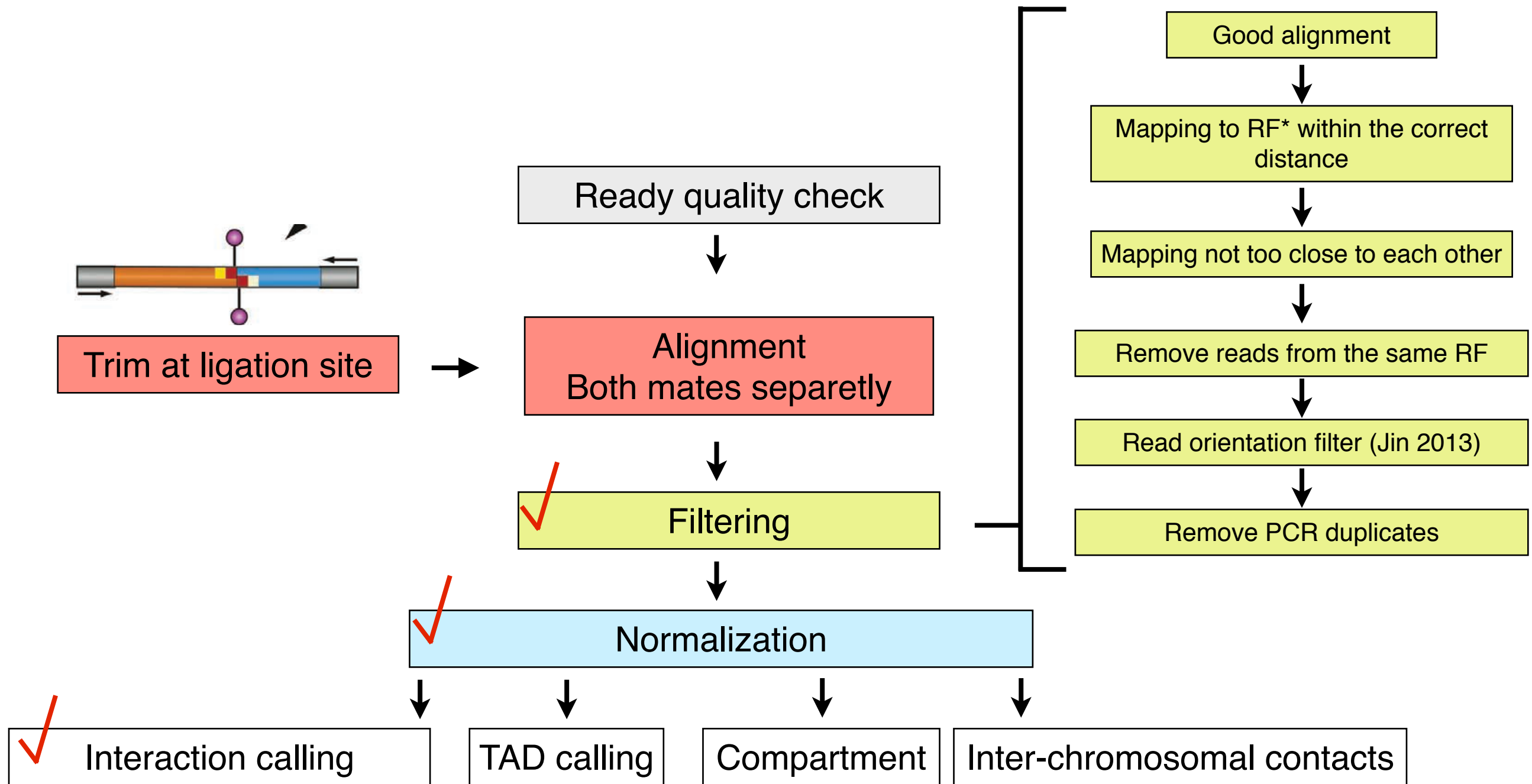
Isolation of interactions



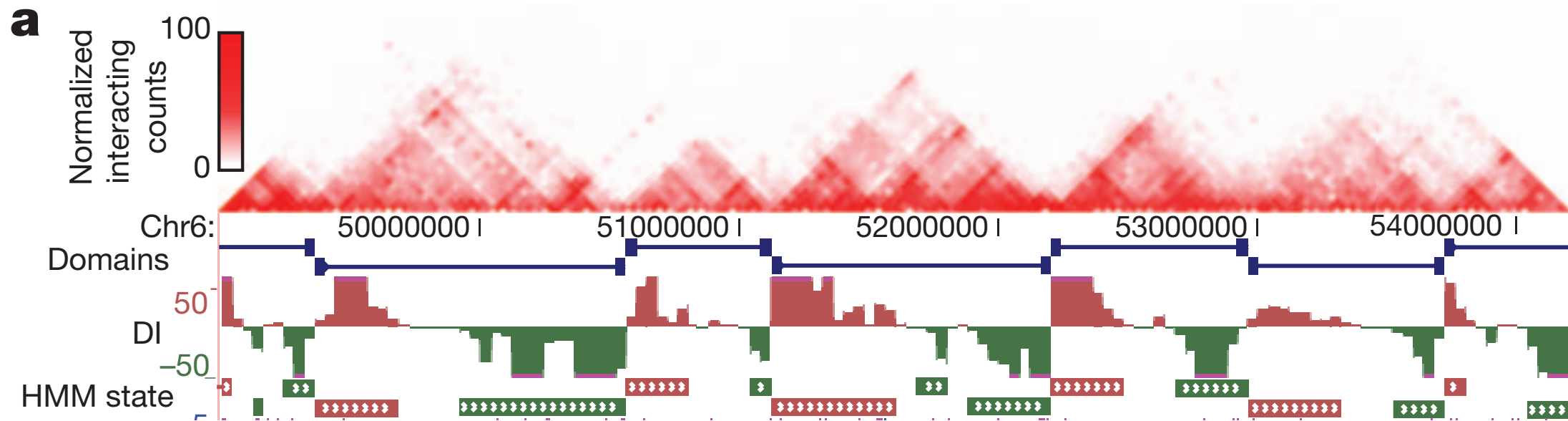
HiCCUPS

Pixels in the middle should have signal 50% higher than the surroundings.

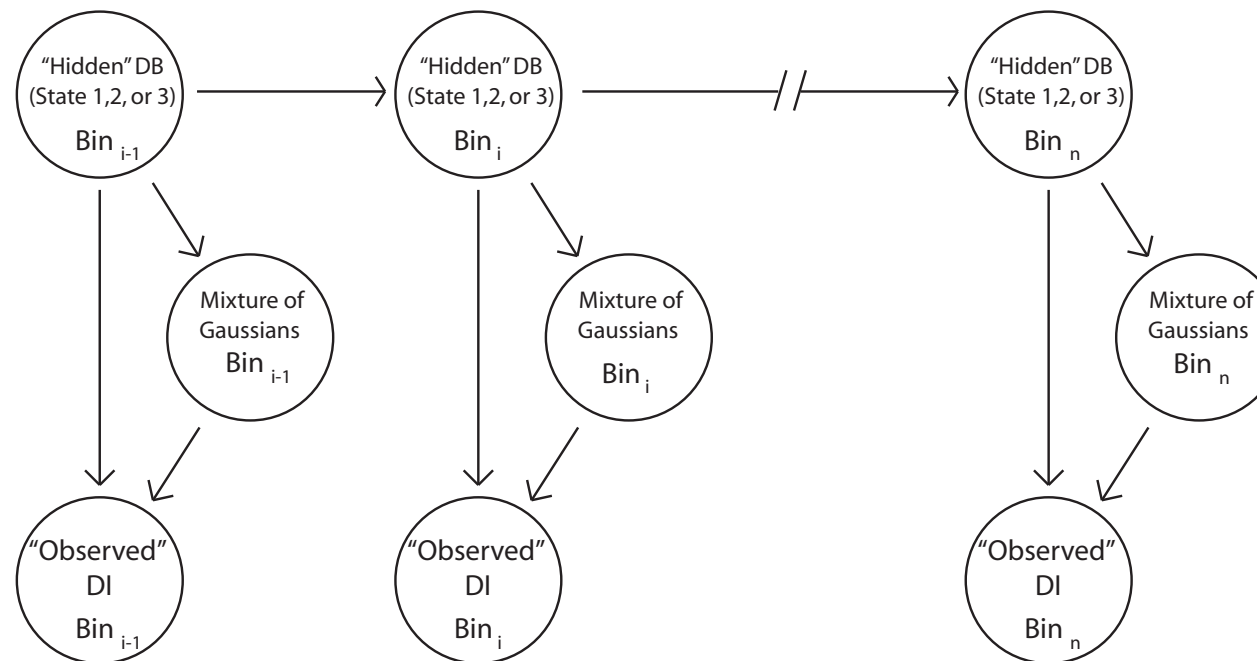
Analysis workflow



Isolation of TADs - directionality index (DI)



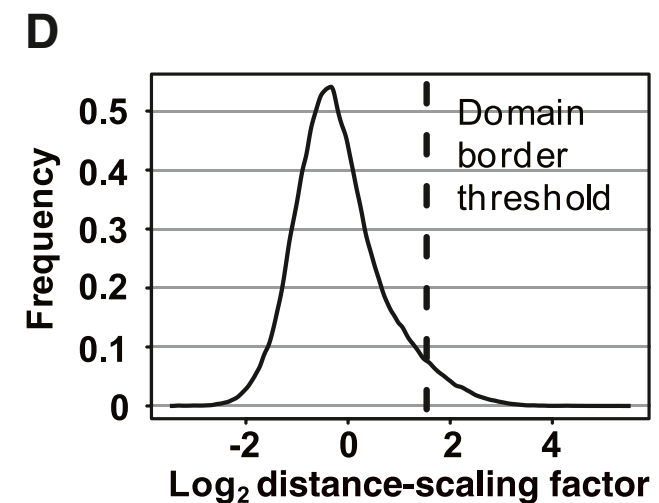
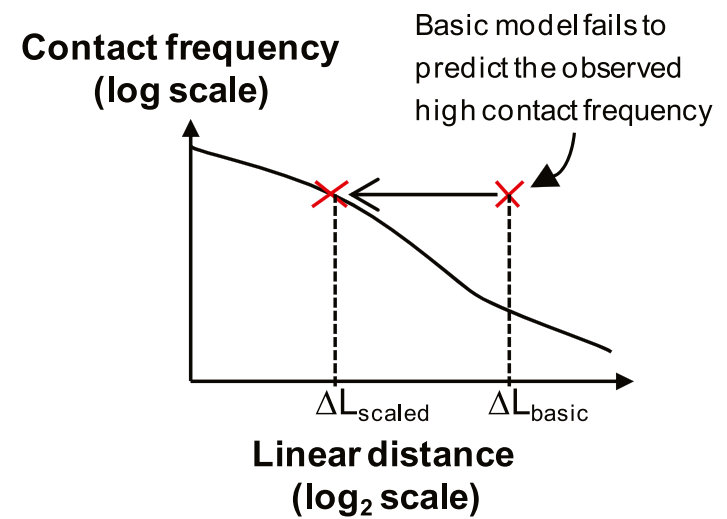
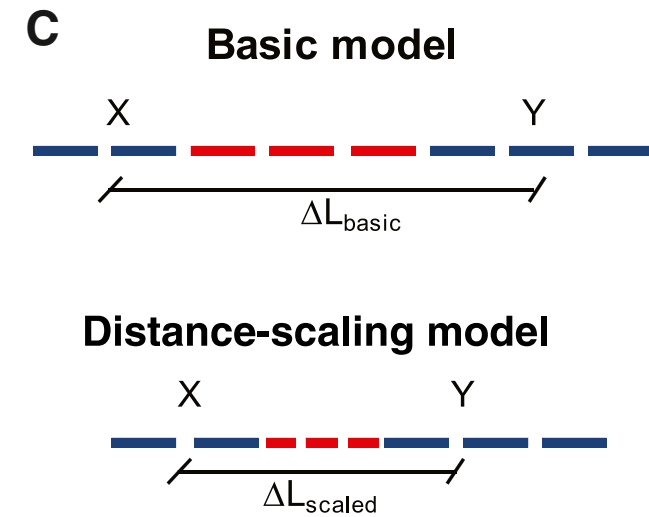
$$DI = \left(\frac{B - A}{|B - A|} \right) \left(\frac{(A - E)^2}{E} + \frac{(B - E)^2}{E} \right)$$



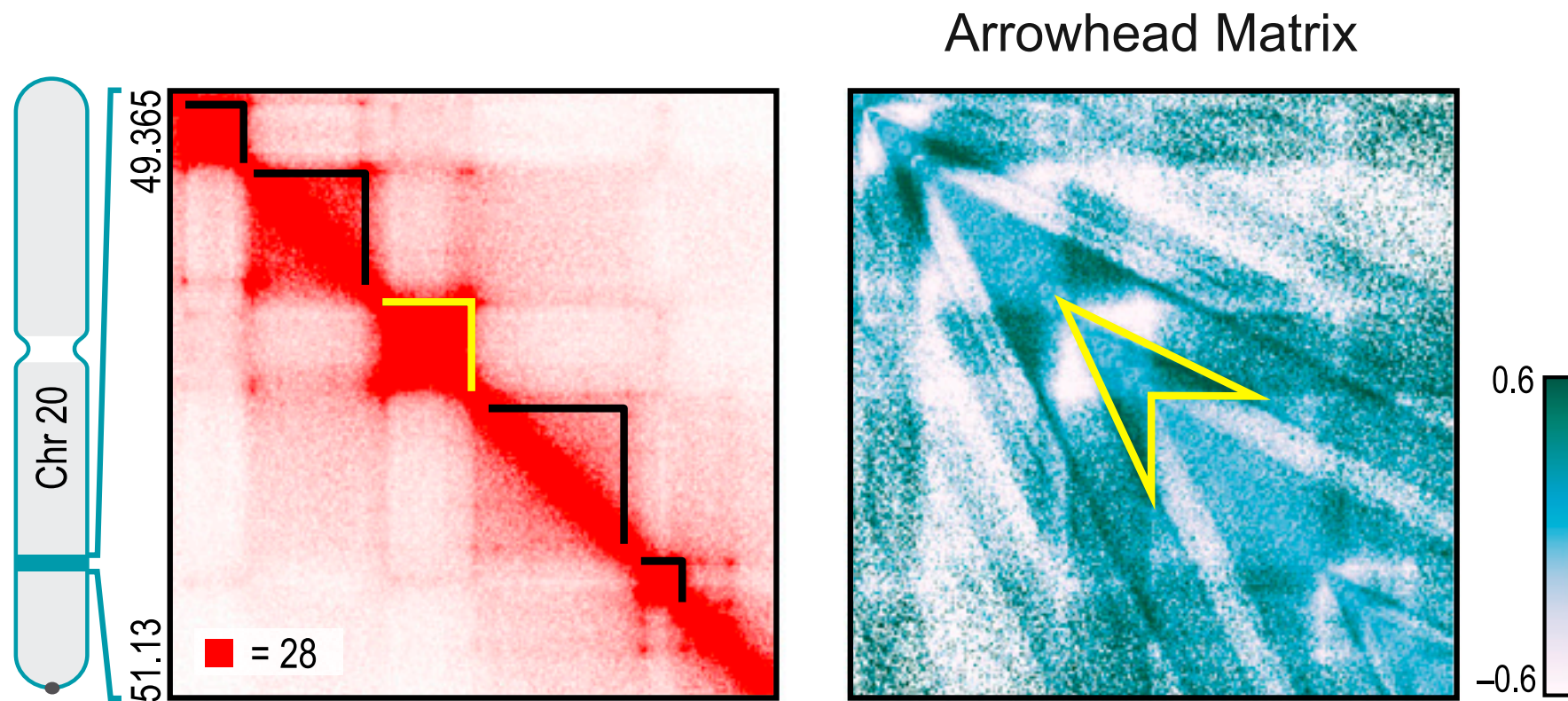
Tom Sexton's and Amos Tanay's method

Particularly convenient for compact genomes

Here *D. melanogaster*



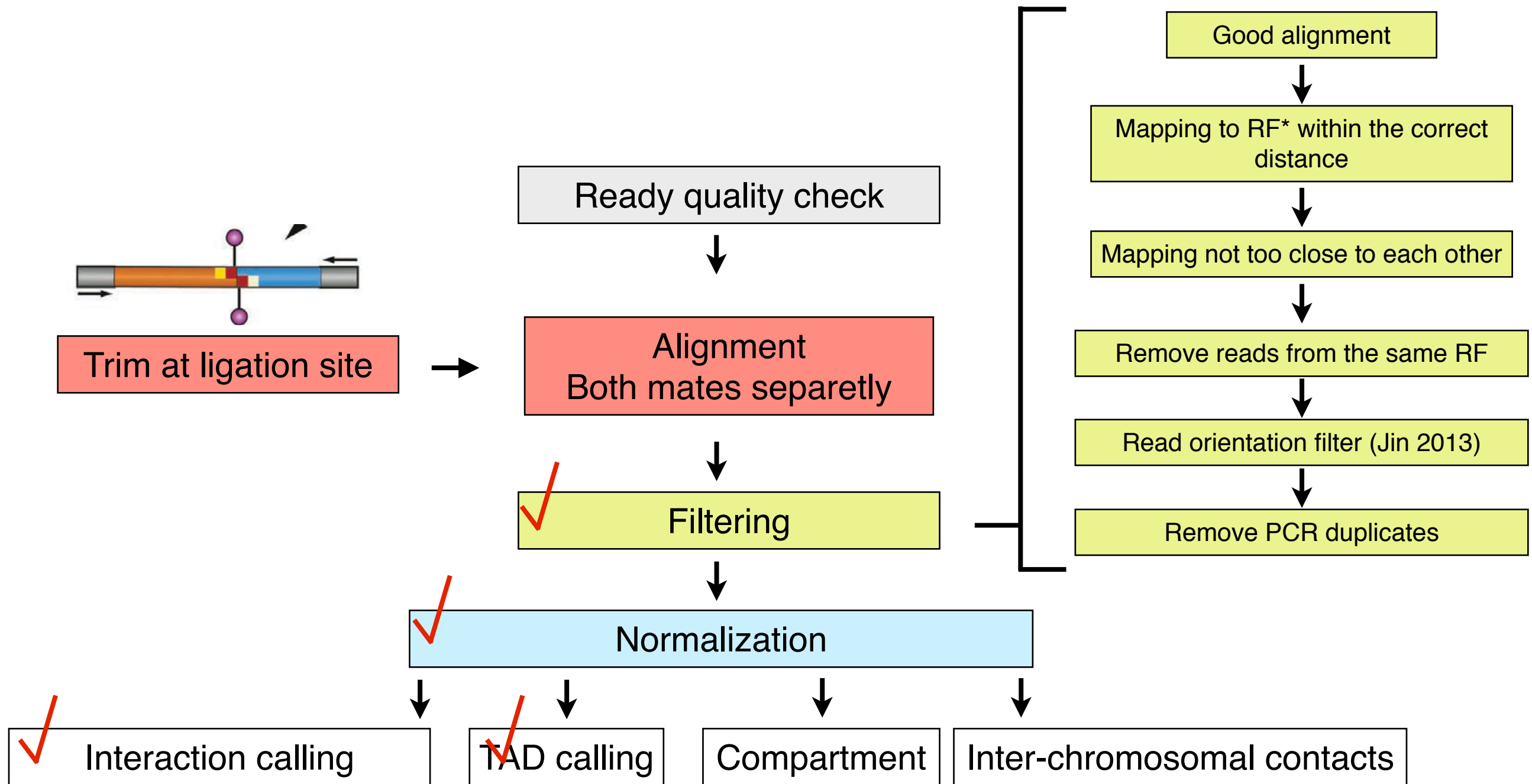
Isolation of TADs - other approaches



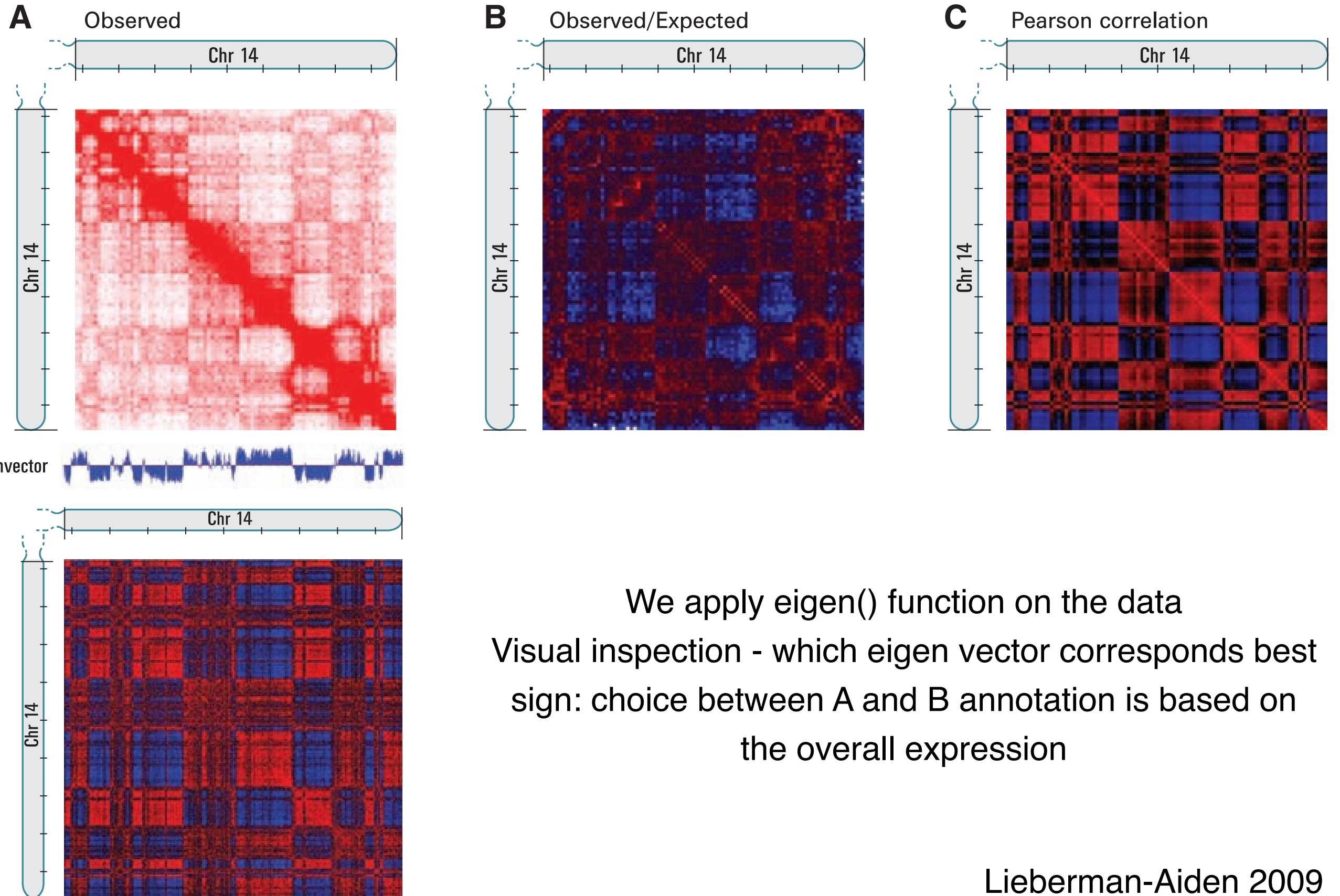
MATRIX -1x(Obs/Exp - 1)

‘A “corner score” matrix, indicating each pixel’s likelihood of lying at the corner of a domain, is efficiently calculated from the arrowhead matrix using dynamic programming.’

Analysis workflow

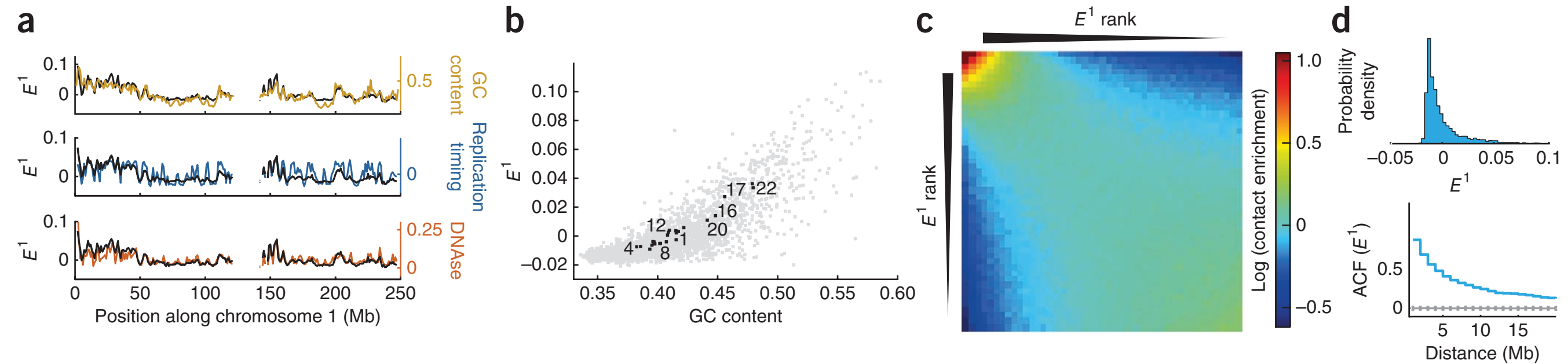


The isolation of compartments



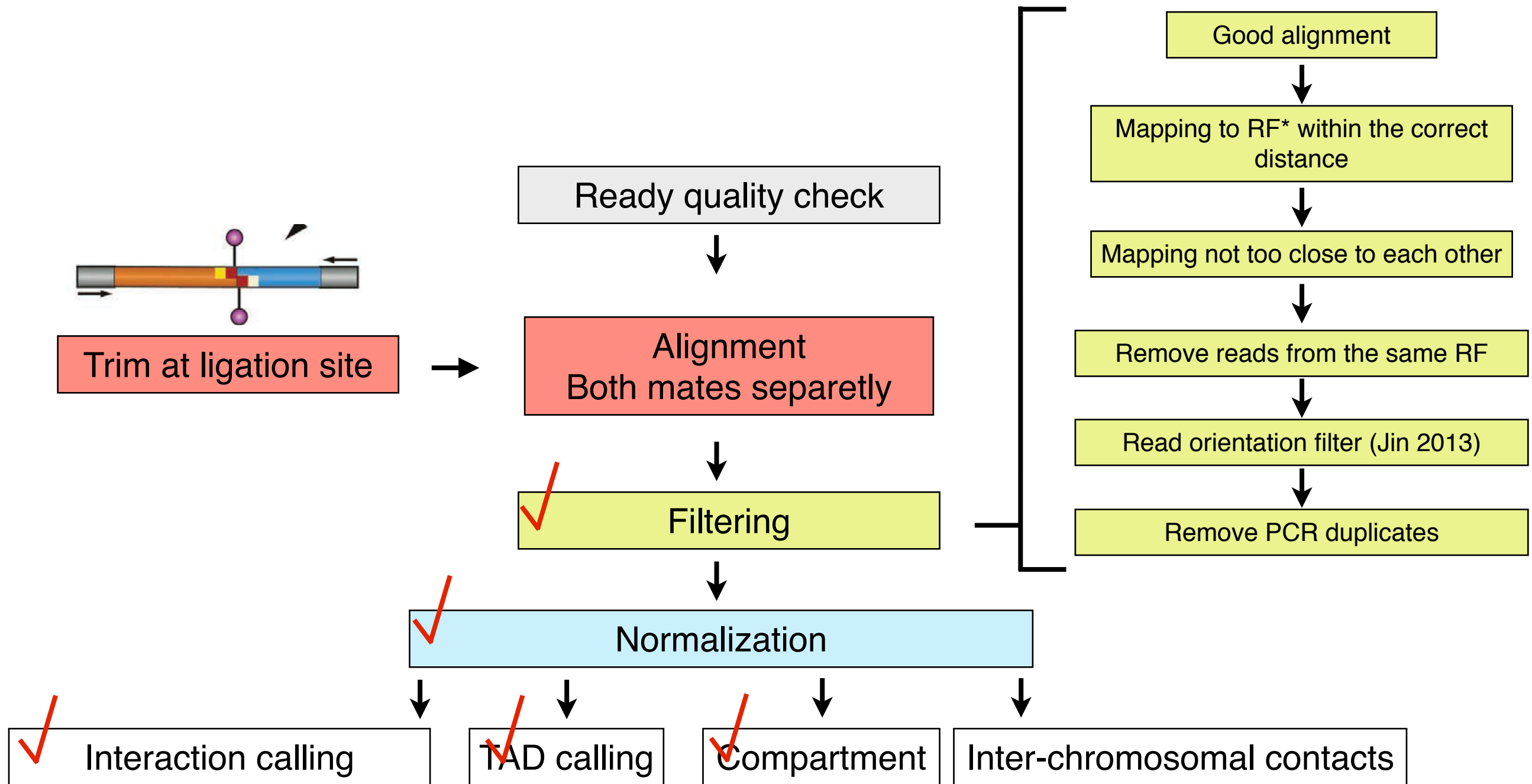
We apply eigen() function on the data
Visual inspection - which eigen vector corresponds best
sign: choice between A and B annotation is based on
the overall expression

Inter-chromosomal interactions and compartments



Taking into account only inter-chromosomal contacts reveals a continuum in E

Analysis workflow



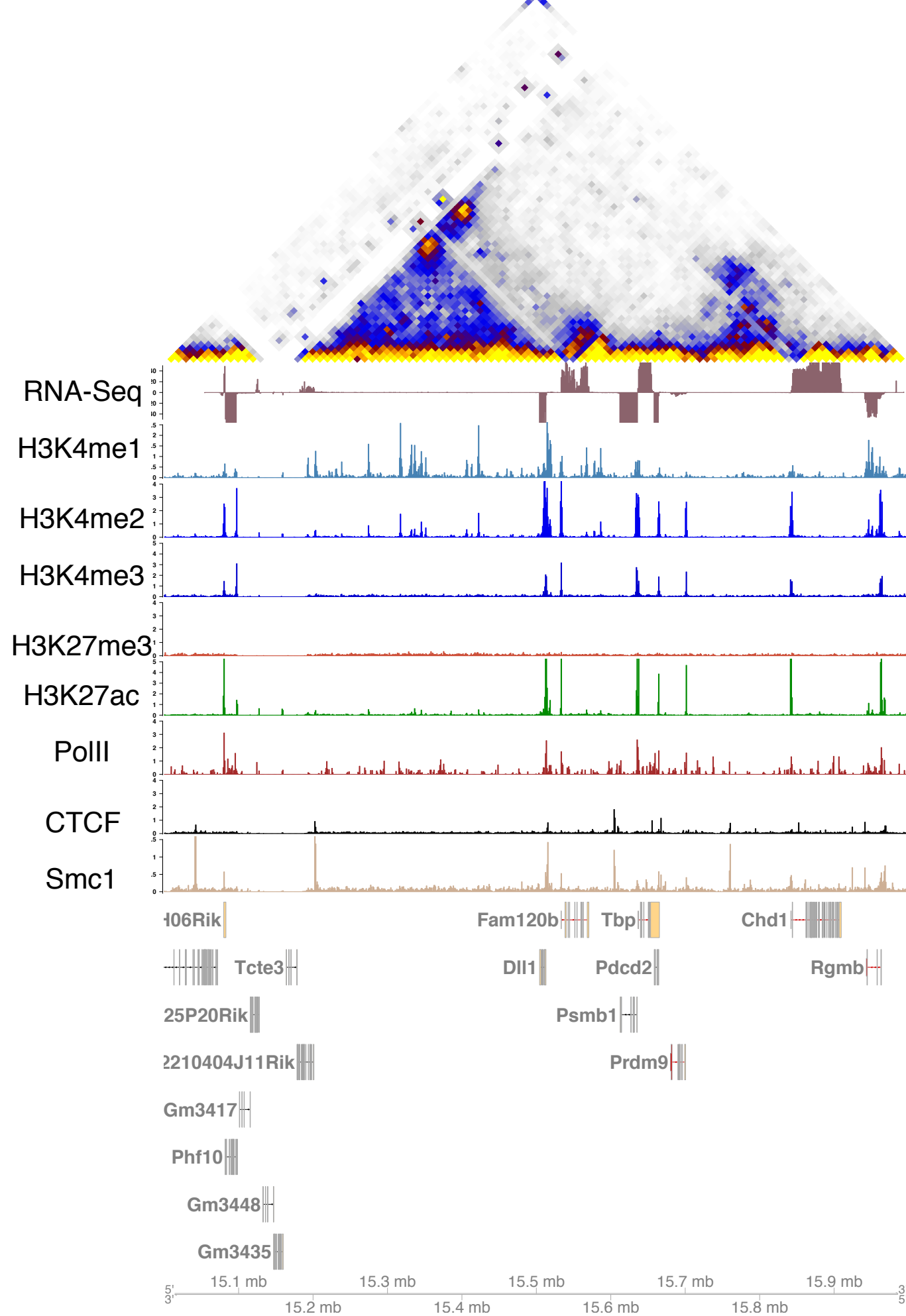
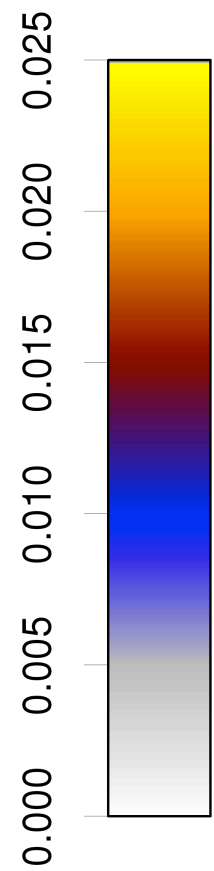
Tools

HiTC - bioconductor package for Hi-C/5C data exploration, quality checks, binning, fitting, visualization

Our tool - Bioconductor package in preparation:

- binning/not
- normalization - ICE and other proportional fitting algorithms (convergence)
- TAD calling
- interaction calling
- compartment analysis
- visualization

Normalized
Interaction
strength



Primer

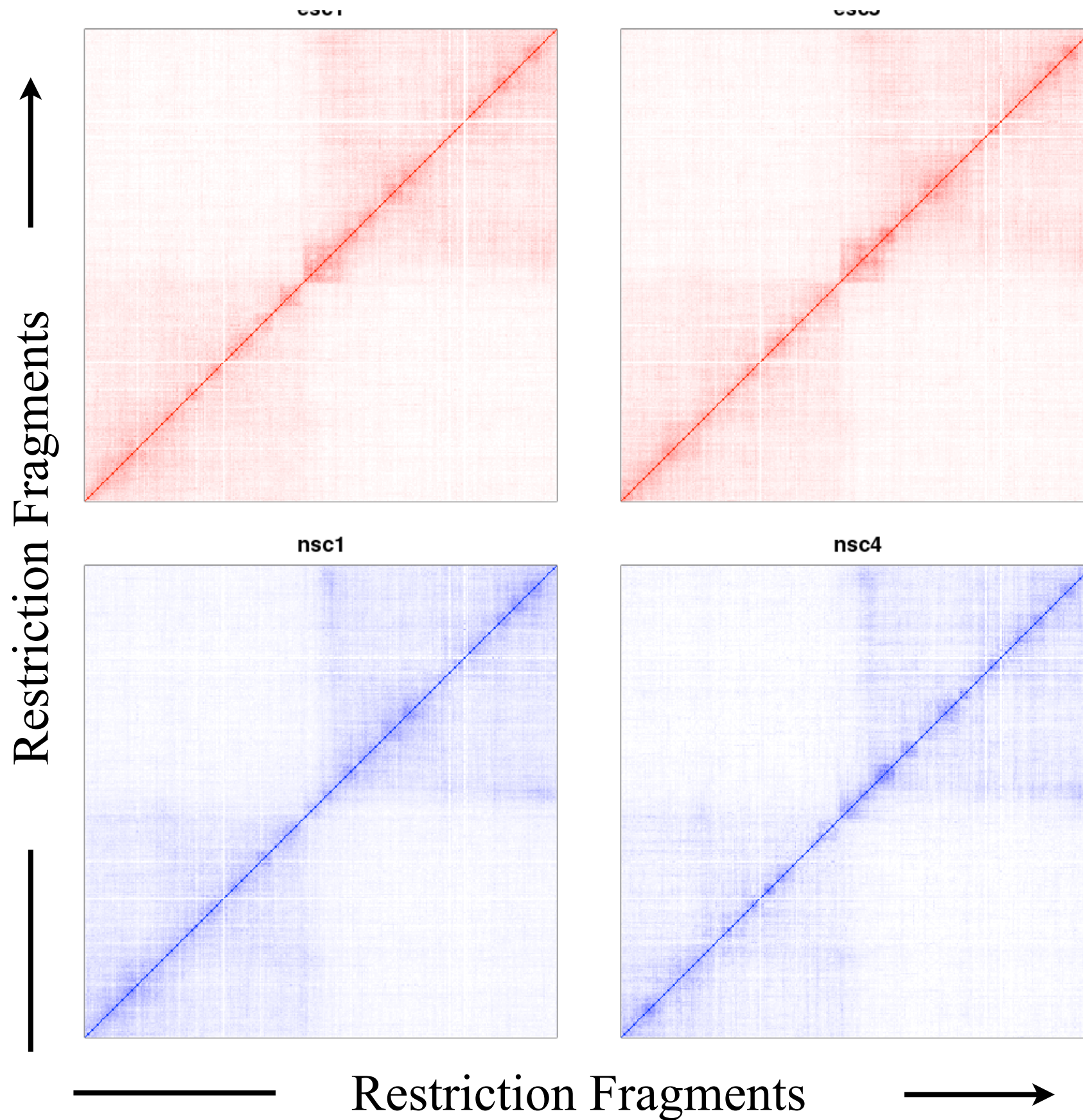
↑ Restriction Fragments

0	30	10	15	12	15	10	3	30	6	3	1	0	0	0	0	0	0	0	0
	0	29	19	3	10	0	1	50	1	3	4	2	0	0	0	0	0	0	0
		0	1	0	0	0	1	69	0	0	0	0	0	0	0	0	0	0	0
			0	2	2	3	1	12	0	0	0	0	0	0	0	0	0	0	0
				0	12	34	20	89	9	0	0	0	0	3	1	1	8	8	0
					0	32	10	56	0	0	1	1	1	0	0	0	0	0	0
						0	45	89	0	0	0	0	0	0	0	0	0	0	0
							0	99	45	30	12	3	1	0	0	0	0	0	0
								0	60	60	12	67	56	20	13	50	29	30	90
									0	12	13	4	3	3	3	1	1	0	1
										0	5	6	2	3	1	1	1	0	0
											0	13	20	15	0	0	0	0	0
												0	34	16	2	3	1	0	0
													0	19	4	1	0	0	0
														0	2	1	0	1	1
															0	1	0	0	0
																0	3	1	0
																	0	3	2
																		0	3
																			0

Restriction Fragments →

$$CS = \frac{\sum_{i=1}^n A_i \times B_i}{\sqrt{\sum_{i=1}^n (A_i)^2} \times \sqrt{\sum_{i=1}^n (B_i)^2}}$$

Cosine similarity

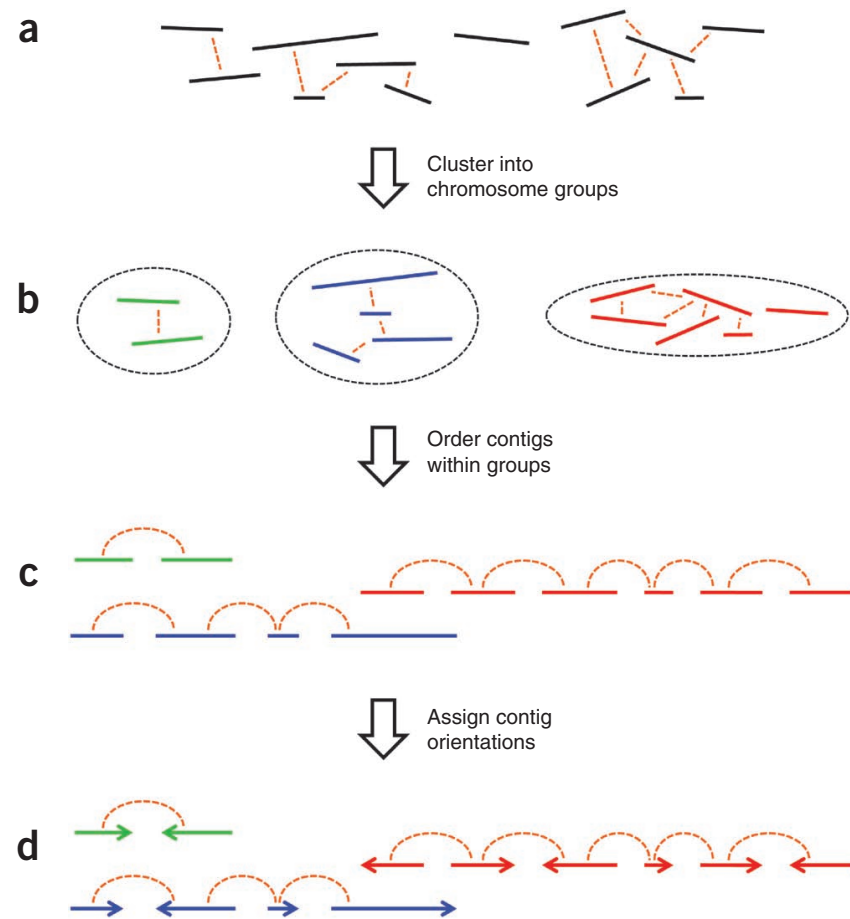


Acknowledgements

Wolfgang Huber
Bernd Klaus
Florian Hahne

Even more exciting use of Hi-C

Genome reassembly

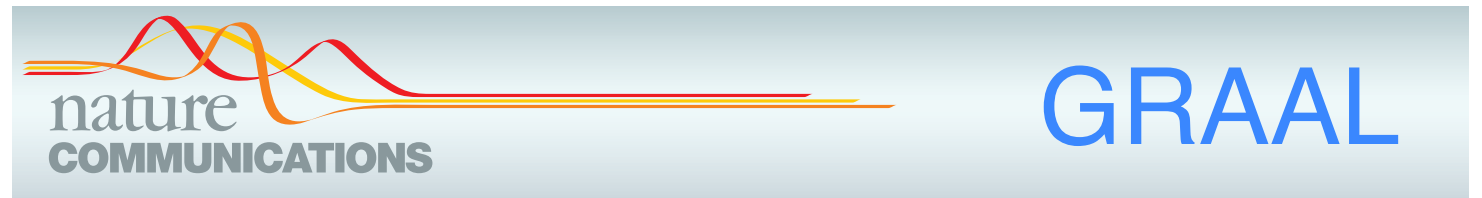


LACHESIS

Burton 2013

Genome scaffolding

Kaplan 2013



ARTICLE

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OPEN

High-quality genome (re)assembly using chromosomal contact data

Hervé Marie-Nelly^{1,2,3,4,5,*}, Martial Marbouty^{1,2,*}, Axel Cournac^{1,2}, Jean-François Flot⁶, Gianni Liti⁷, Dante Poggi Parodi^{5,8}, Sylvie Syan⁹, Nancy Guillén⁹, Antoine Margeot⁸, Christophe Zimmer^{3,4} & Romain Koszul^{1,2}

References (I)

- Ay F, Bailey TL, Noble WS. 2014. Statistical confidence estimation for Hi-C data reveals regulatory chromatin contacts. *Genome Res* **24**: 999–1011.
- Baù D, Sanyal A, Lajoie BR, Capriotti E, Byron M, Lawrence JB, Dekker J, Marti-Renom M a. 2011. The three-dimensional folding of the α -globin gene domain reveals formation of chromatin globules. *Nat Struct Mol Biol* **18**: 107–114.
- Burton JN, Adey A, Patwardhan RP, Qiu R, Kitzman JO, Shendure J. 2013. Chromosome-scale scaffolding of de novo genome assemblies based on chromatin interactions. *Nat Biotechnol* **31**: 1119–25.
- Dixon JR, Selvaraj S, Yue F, Kim A, Li Y, Shen Y, Hu M, Liu JS, Ren B. 2012. Topological domains in mammalian genomes identified by analysis of chromatin interactions. *Nature* **485**: 376–80.
- Hu M, Deng K, Selvaraj S, Qin Z, Ren B, Liu JS. 2012. HiCNorm: Removing biases in Hi-C data via Poisson regression. *Bioinformatics* **28**: 3131–3133.
- Imakaev M, Fudenberg G, Mccord RP, Naumova N, Goloborodko A, Lajoie BR, Dekker J, Mirny LA. 2012. Iterative correction of Hi-C data reveals hallmarks of chromosome organization. *Nat Methods* **9**.
- Jin F, Li Y, Dixon JR, Selvaraj S, Ye Z, Lee AY, Yen C-A, Schmitt AD, Espinoza C a, Ren B. 2013. A high-resolution map of the three-dimensional chromatin interactome in human cells. *Nature* **14**: 290–294.
- Kalhor R, Tjong H, Jayathilaka N, Alber F, Chen L. 2011. Genome architectures revealed by tethered chromosome conformation capture and population-based modeling. *Nat Biotechnol* **30**: 90–98.
- Kaplan N, Dekker J. 2013. High-throughput genome scaffolding from in vivo DNA interaction frequency. *Nat Biotechnol* **31**: 1143–7.

References (II)

- Li W, Gong K, Li Q, Alber F, Zhou XJ. 2014. Hi-Corrector: a fast, scalable and memory-efficient package for normalizing large-scale Hi-C data. *Bioinformatics* **31**: 960–962.
- Lieberman-Aiden E, van Berkum NL, Williams L, Imakaev M, Ragooczy T, Telling A, Amit I, Lajoie BR, Sabo PJ, Dorschner MO, et al. 2009. Comprehensive mapping of long-range interactions reveals folding principles of the human genome. *Science (80-)* **326**: 289–93.
- Marie-Nelly H, Marbouty M, Cournac A, Flot J-F, Liti G, Parodi DP, Syan S, Guillén N, Margeot A, Zimmer C, et al. 2014. High-quality genome (re)assembly using chromosomal contact data. *Nat Commun* **5**: 5695.
- Naumova N, Imakaev M, Fudenberg G, Zhan Y, Lajoie BR, Mirny L a, Dekker J. 2013. Organization of the mitotic chromosome. *Science* **342**: 948–53.
- Rao SSP, Huntley MH, Durand NC, Stamenova EK. 2014. A 3D Map of the Human Genome at Kilobase Resolution Reveals Principles of Chromatin Looping. *Cell* 1–16.
- Selvaraj S, R Dixon J, Bansal V, Ren B. 2013. Whole-genome haplotype reconstruction using proximity-ligation and shotgun sequencing. *Nat Biotechnol* **31**: 1111–8.
- Sexton T, Yaffe E, Kenigsberg E, Bantignies F, Leblanc B, Hoichman M, Parrinello H, Tanay A, Cavalli G. 2012. Three-dimensional folding and functional organization principles of the Drosophila genome. *Cell* **148**: 458–72.
- Yaffe E, Tanay A. 2011. Probabilistic modeling of Hi-C contact maps eliminates systematic biases to characterize global chromosomal architecture. *Nat Genet* **43**: 1059–65.