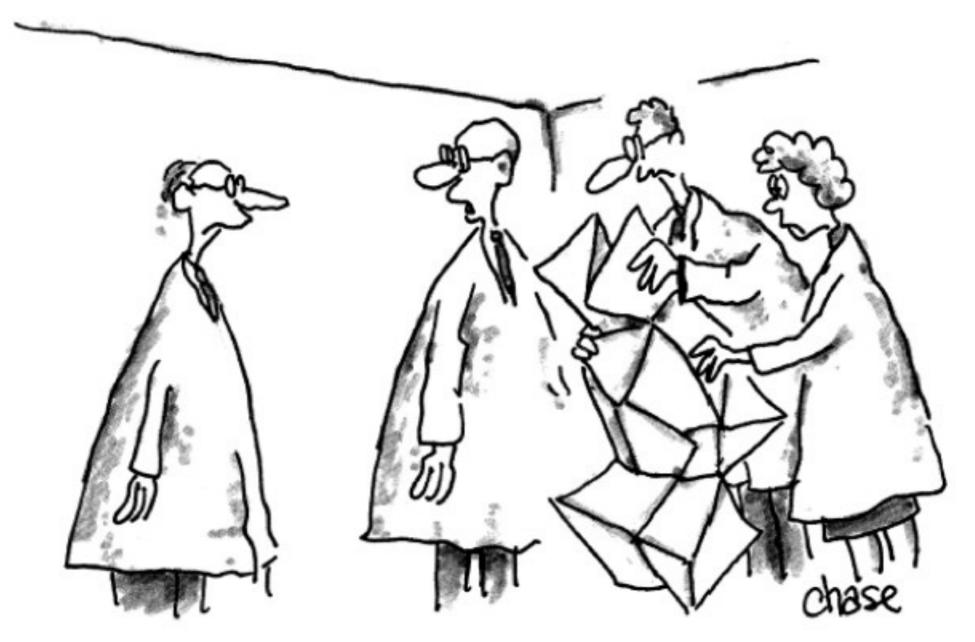
messing around with Hi-C data

Koon-Kiu Yan

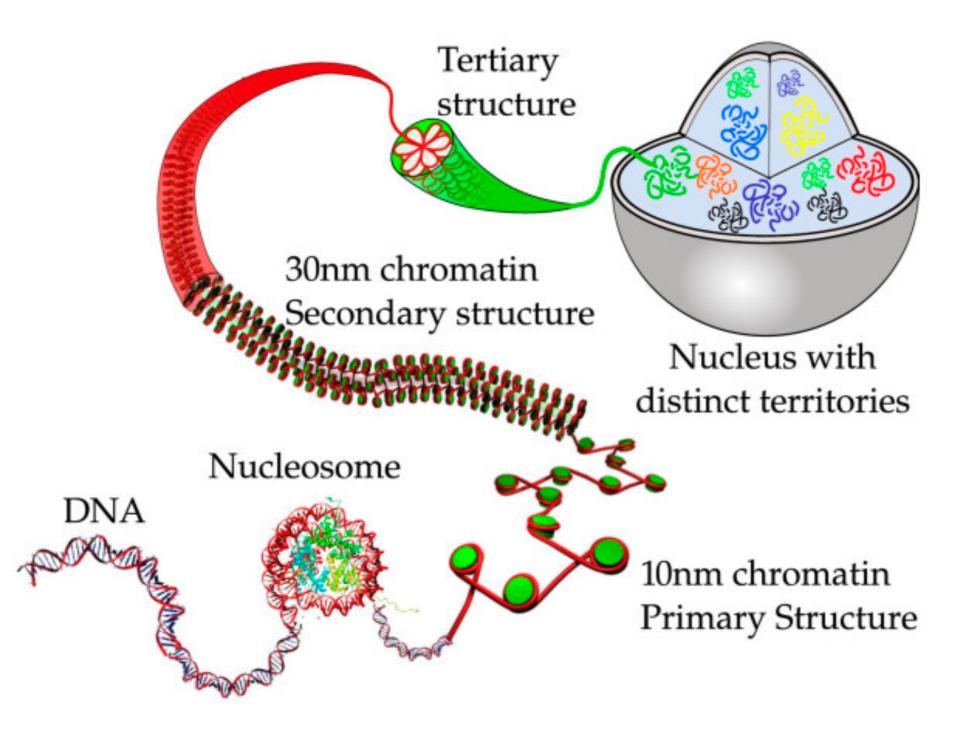
Group Meeting, March 2015

3D organization of genome



"We finished the genome map, now we can't figure out how to fold it."

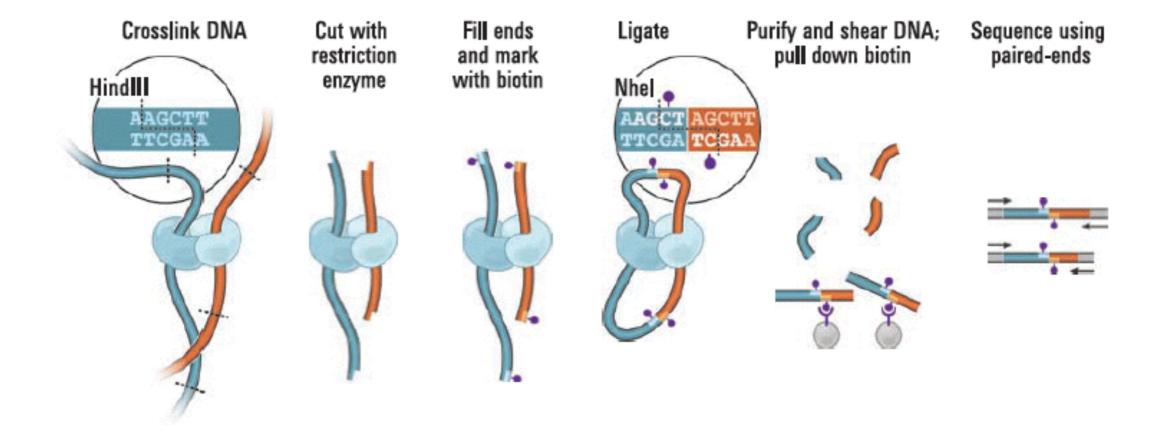
3D organization of genome

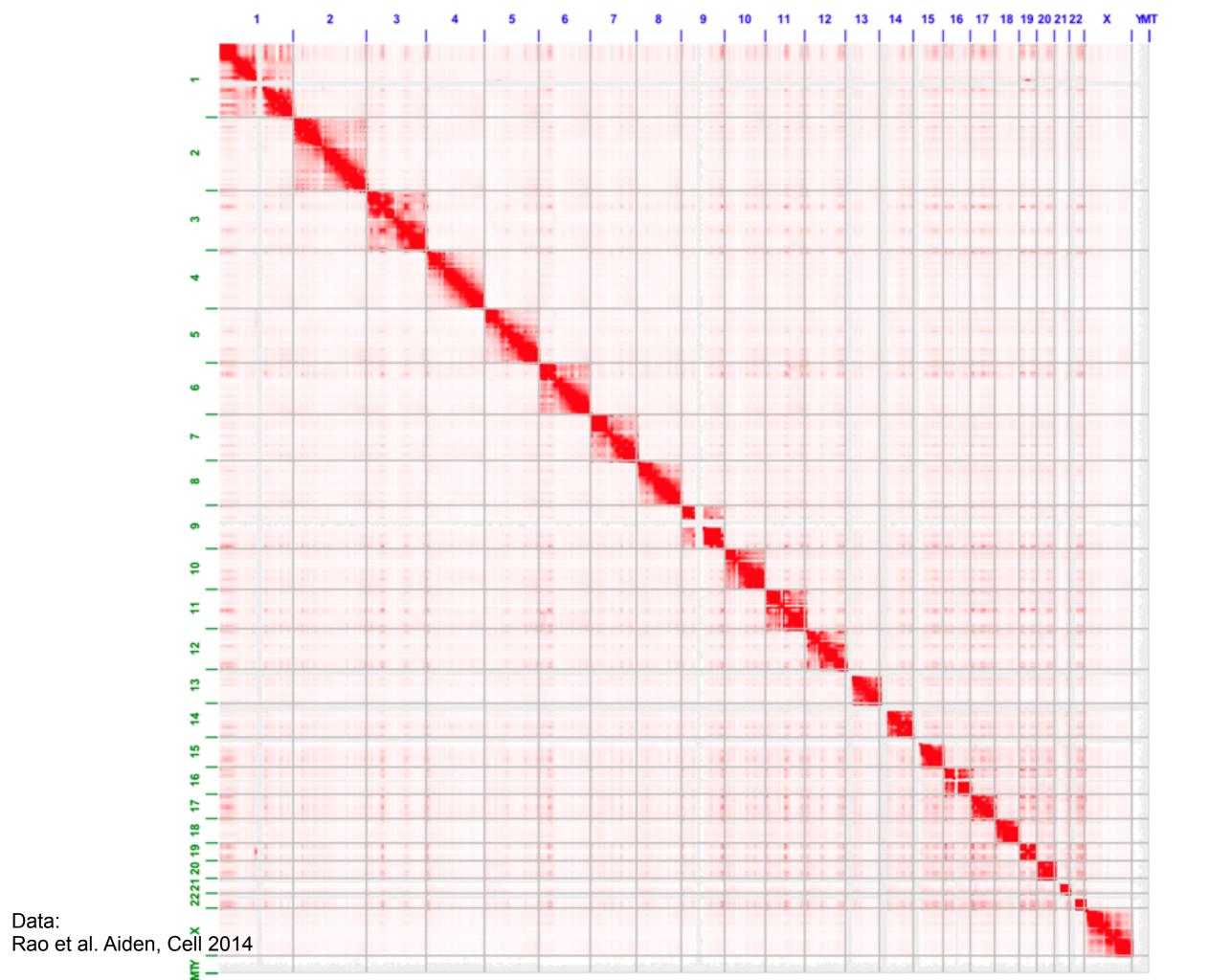


Chromosome conformation capture (3C) and Hi-C

Comprehensive Mapping of Long-Range Interactions Reveals Folding Principles of the Human Genome

Erez Lieberman-Aiden,^{1,2,3,4}* Nynke L. van Berkum,⁵* Louise Williams,¹ Maxim Imakaev,² Tobias Ragoczy,^{6,7} Agnes Telling,^{6,7} Ido Amit,¹ Bryan R. Lajoie,⁵ Peter J. Sabo,⁸ Michael O. Dorschner,⁸ Richard Sandstrom,⁸ Bradley Bernstein,^{1,9} M. A. Bender,¹⁰ Mark Groudine,^{6,7} Andreas Gnirke,¹ John Stamatoyannopoulos,⁸ Leonid A. Mirny,^{2,11} Eric S. Lander,^{1,12,13}† Job Dekker⁵† SCIENCE VOL 326 9 OCTOBER 2009





- Reproducibility and QC metrics in ENCODE
 3D nucleome subgroup
- 2. Identifying topologically associating domains in multiple resolutions

Updates of the ENCODE 3D nucleome subgroup

- Preparation of manuscript for ENCODE guidelines for assessing the quality and the reproducibility of chromosome conformation capture experiments
 - Similar to ENCODE ChIP-seq guidelines (Landt et al. Genome Research 2012)

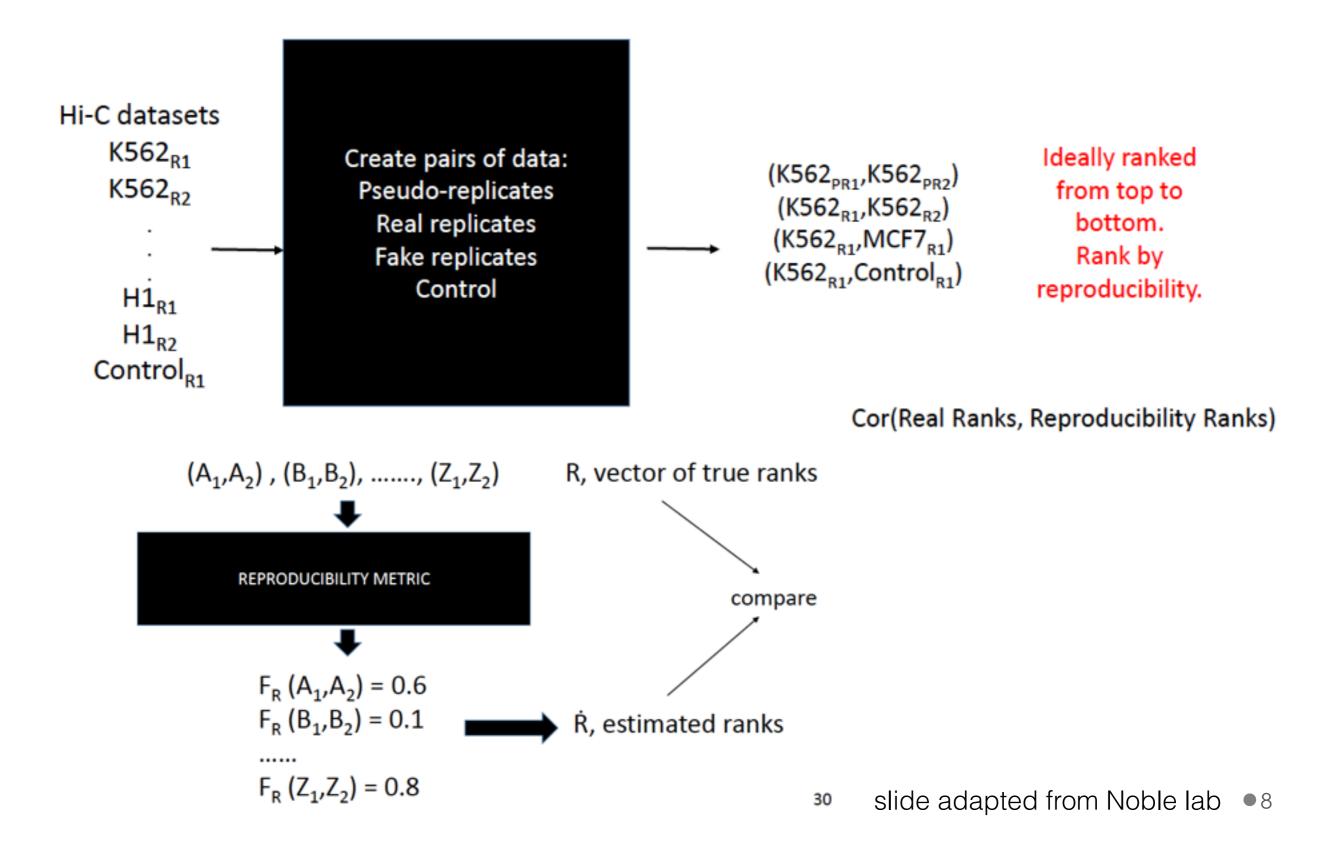




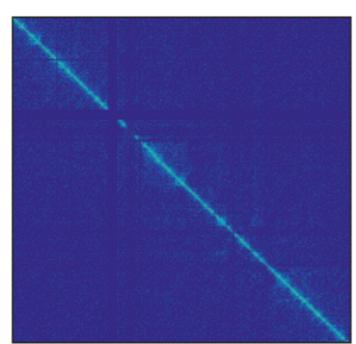
Hi-C data 11 cell types 2 **replicates**

Hi-C data Mouse forebrain Time course 2 **replicates**

Evaluate reproducibility metric

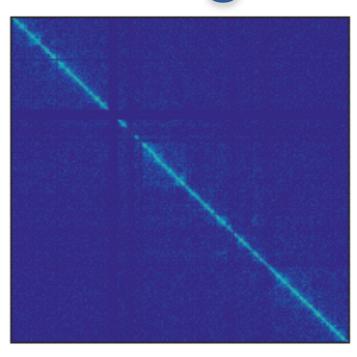


Comparing contact matrices is a technical challenge

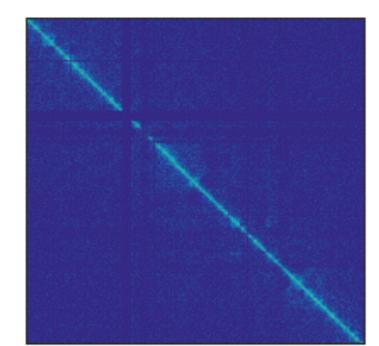


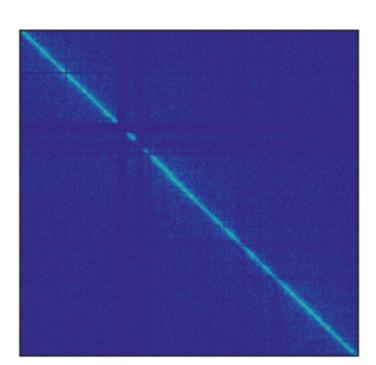
Pair 6

Pair 22



r = 0.95

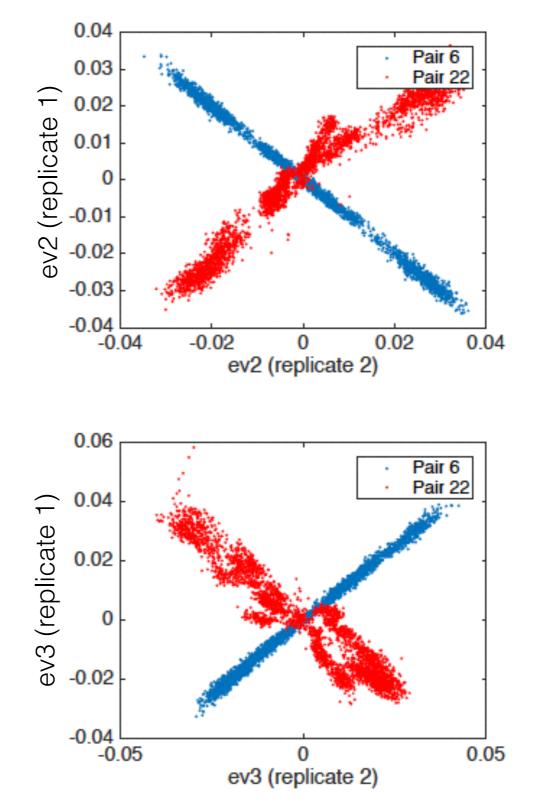




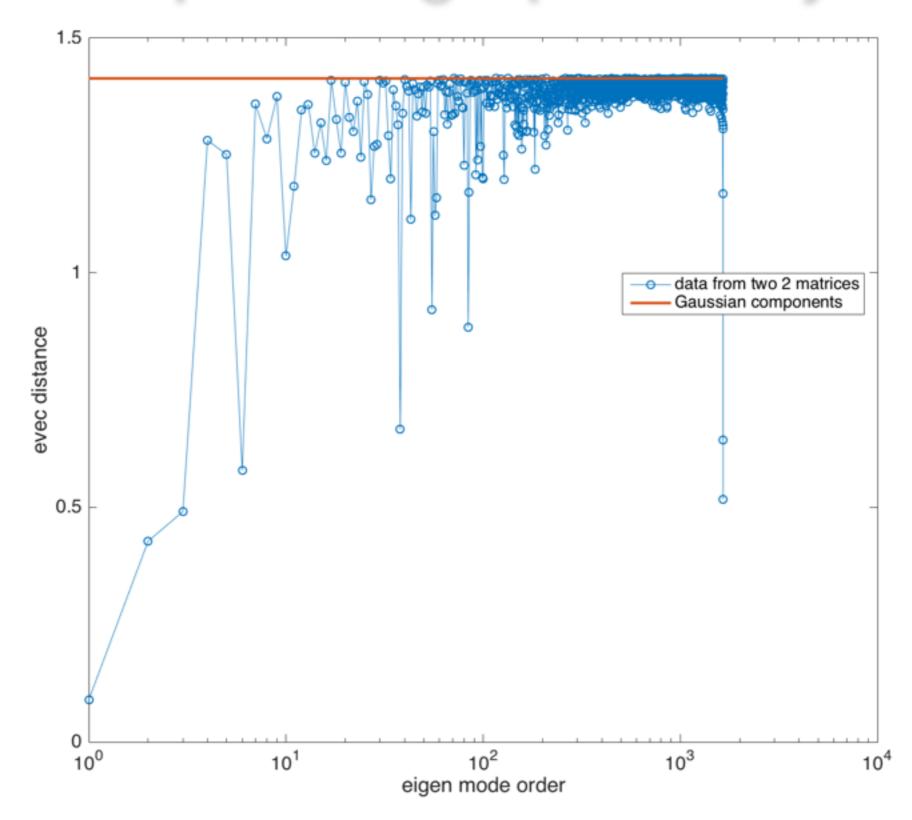
r = 0.70

Quantifying reproducibility using spectral graph theory

- Laplacian L = D A $\mathcal{L} = I - D^{-1/2} A D^{-1/2}$ $0 = \lambda_0 \le \lambda_1 < \lambda_2 < \dots < \lambda_n$
 - leading eigenvectors capture the structures of the graph (dimension reduction)
 - equivalent to eigenmodes of the corresponding random walk on the graph
 - for each pair of leading eigenvectors, calculate the Euclidean distance

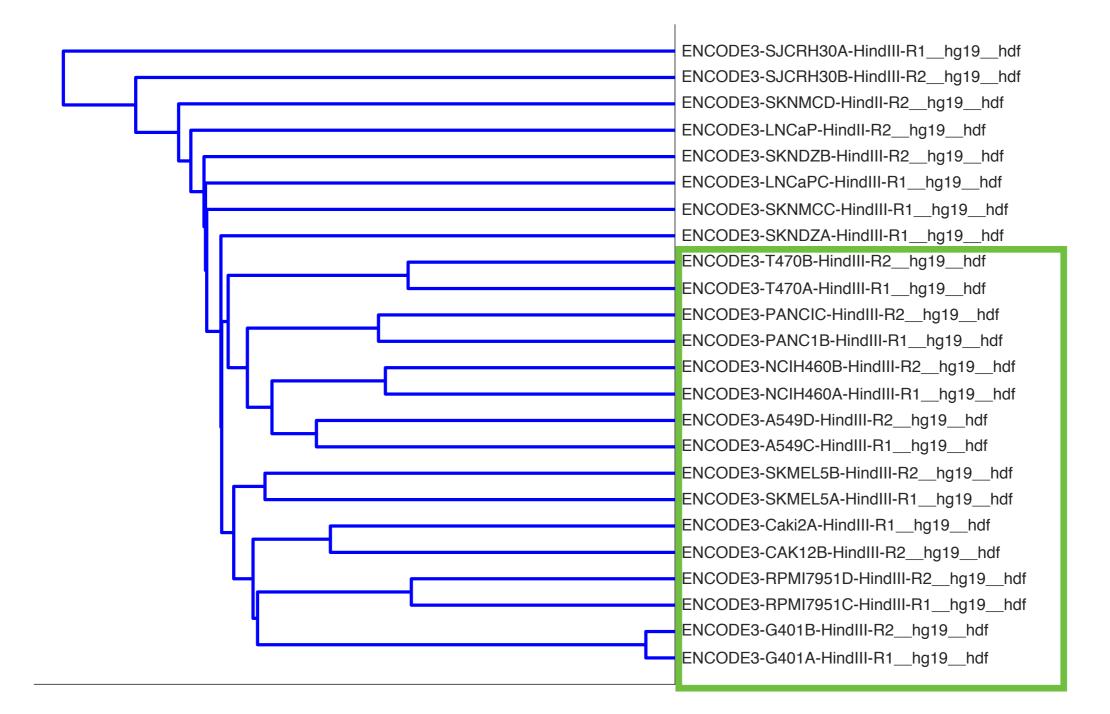


Quantifying reproducibility using spectral graph theory



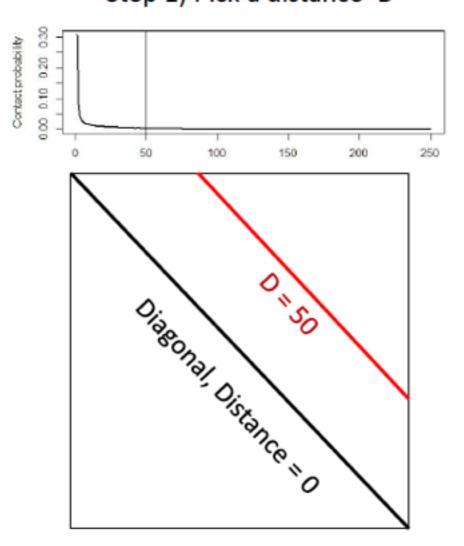
•11

Quantifying reproducibility using spectral graph theory



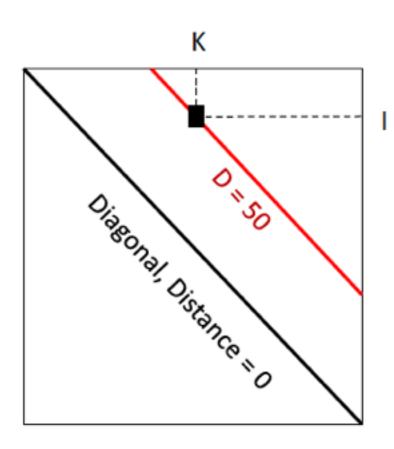
Simulating Noise in Hi-C

Distance effect Step 1) Pick a distance 'D'

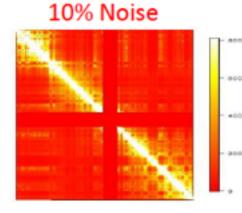


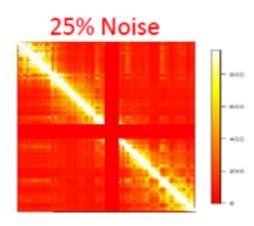
Mappability & GC biases Step 2) Choose M_{IK} at distance 'D' P(I,K) ~ P(I) x P(K)

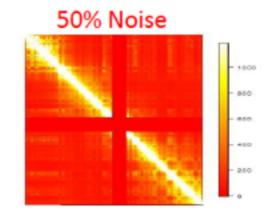
K Diagonal Distance 10 Step 3) Add +1 to chosen bin



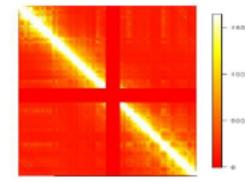
Simulating Noise in Hi-C



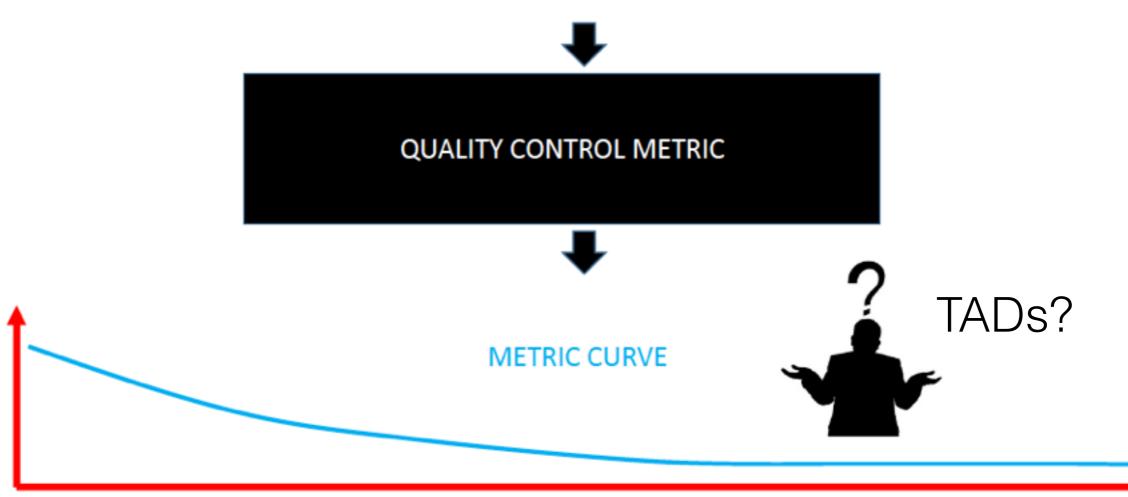




100% Noise



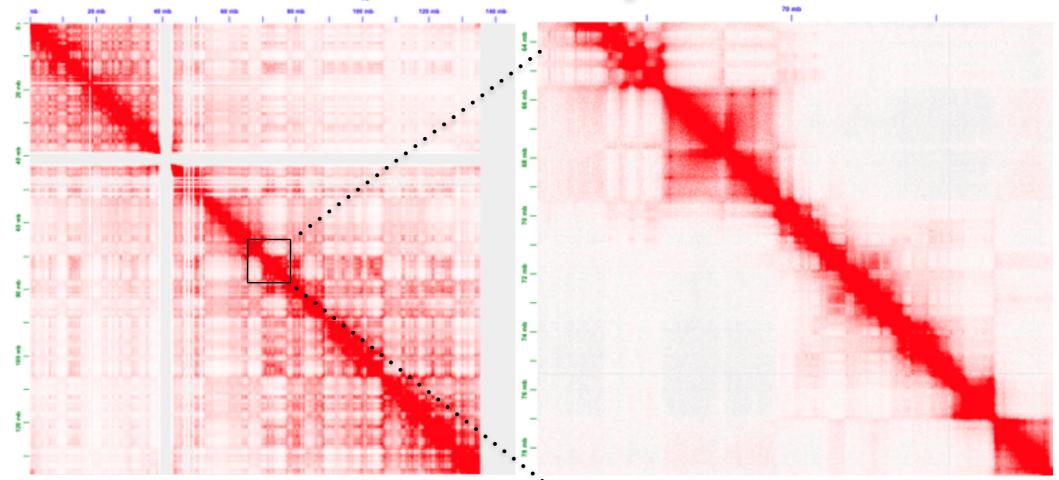
D, D+10%Noise, .. D+100%Noise

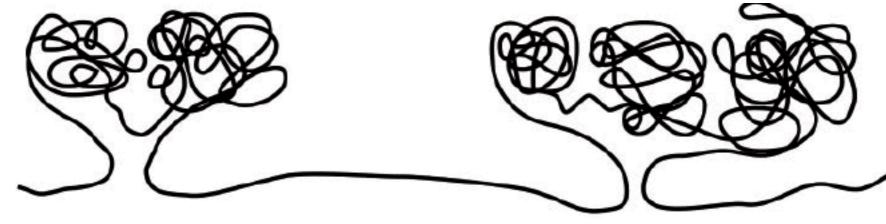


NOISE

- Studies of reproducibility and QC metric in ENCODE 3D nucleome subgroup
- Identifying Topologically associated domains in multiple resolutions

Topologically associating domains (TADs)

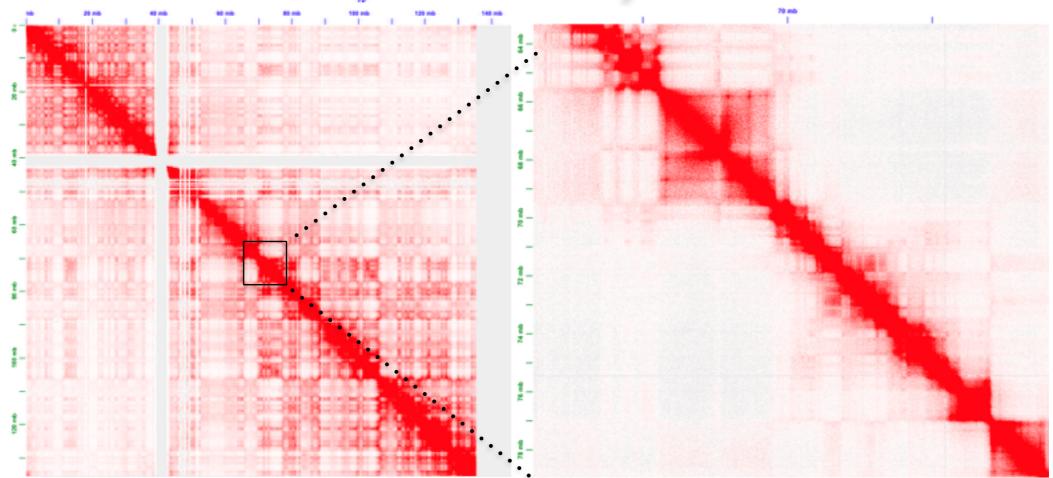




hESC chr1

Weinreb and Raphael, Bioinformatics 2015

Topologically associating domains (TADs)



How could we identify such domains in multiple resolution? Domains resemble network modules

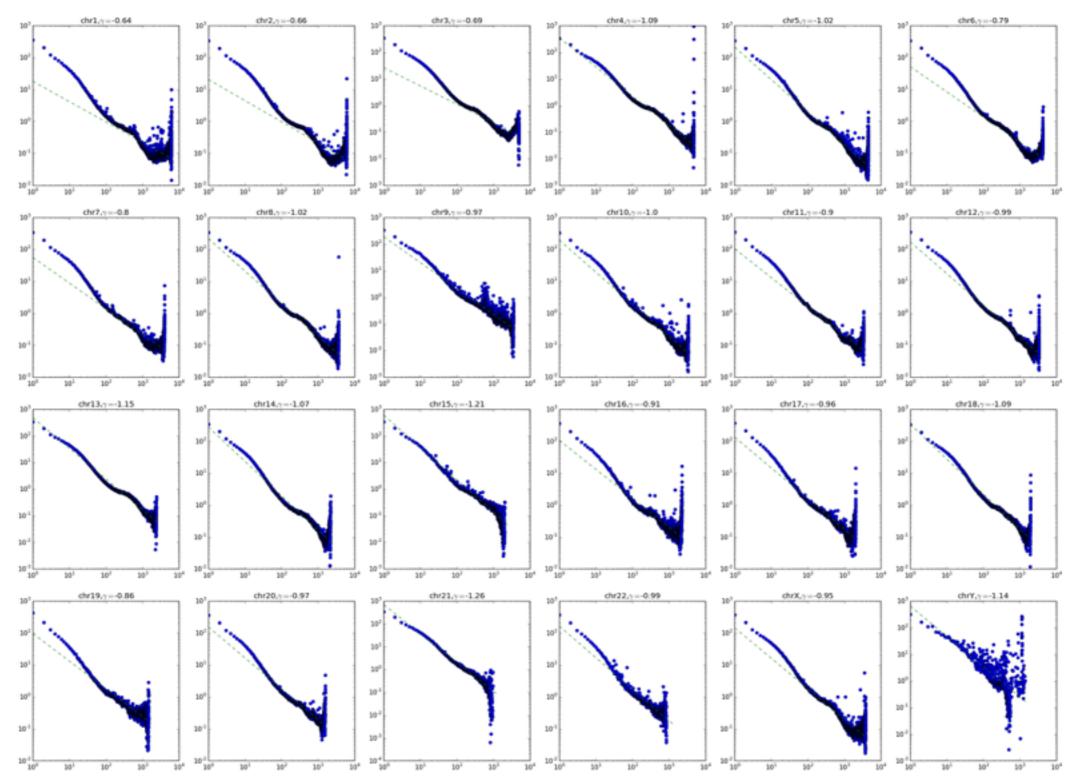
$$Q = \frac{1}{2m} \sum_{i,j} \left(W_{ij} - \frac{k_i k_j}{2m} \right) \delta_{\sigma_i \sigma_j}$$

$$Q = \frac{1}{2N} \sum_{i,j} \left(W_{ij} - \frac{c_i c_j}{2N} \right) \delta_{\sigma_i \sigma_j}$$

$$c_j$$

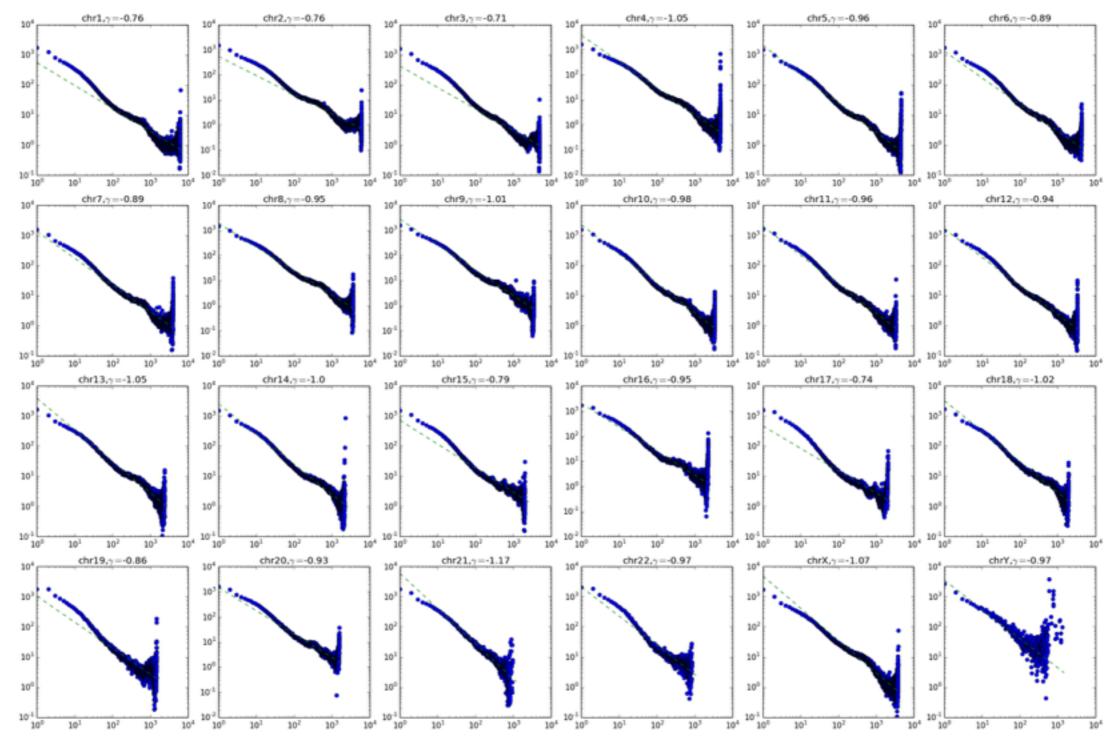
•17

Number of contacts versus genomic distance



Data: Dixon et al. 2012

Number of contacts versus genomic distance



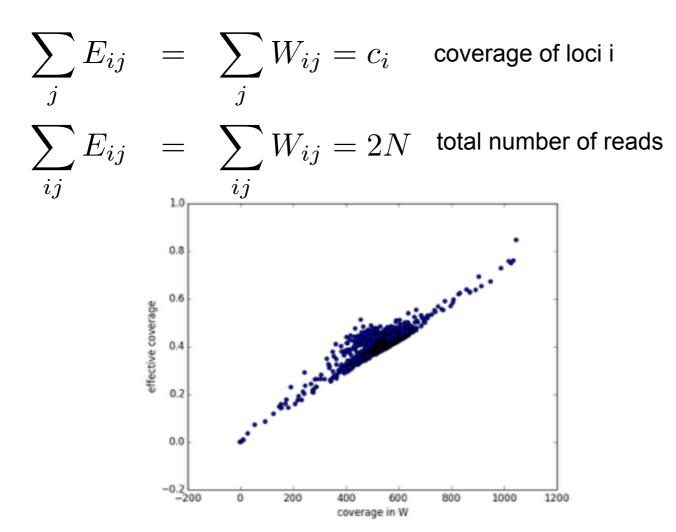
Data: Dekker A549C

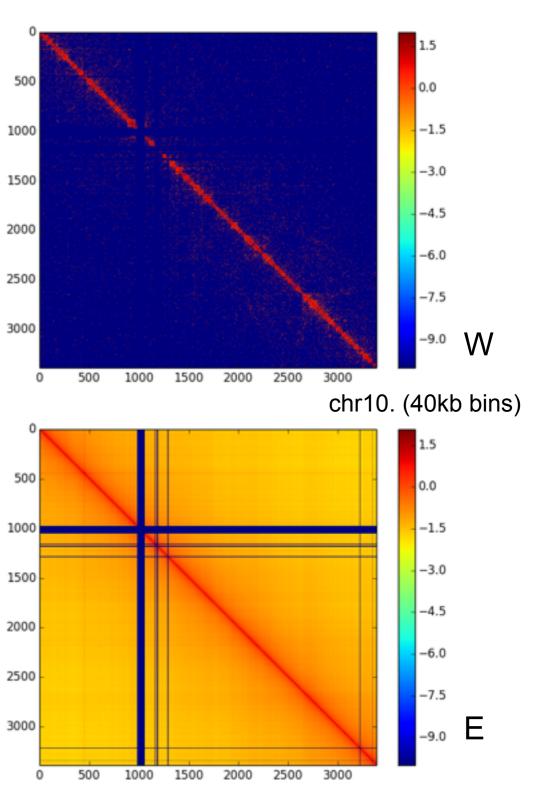
Identifying TADs in multiple resolutions

A null model that takes into account of the genomics distance

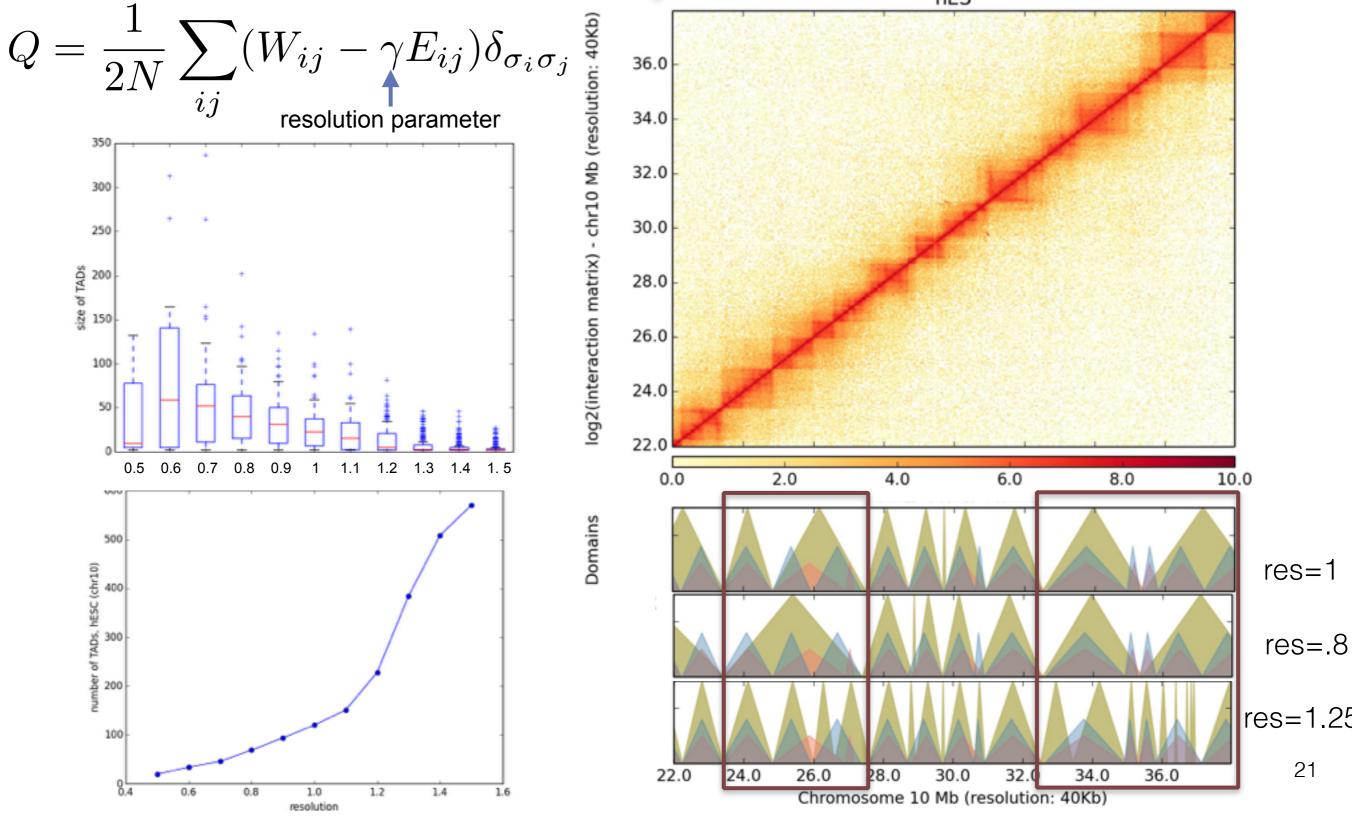
$$E_{ij} = c_i^* c_j^* f(|i-j|)$$

contraints

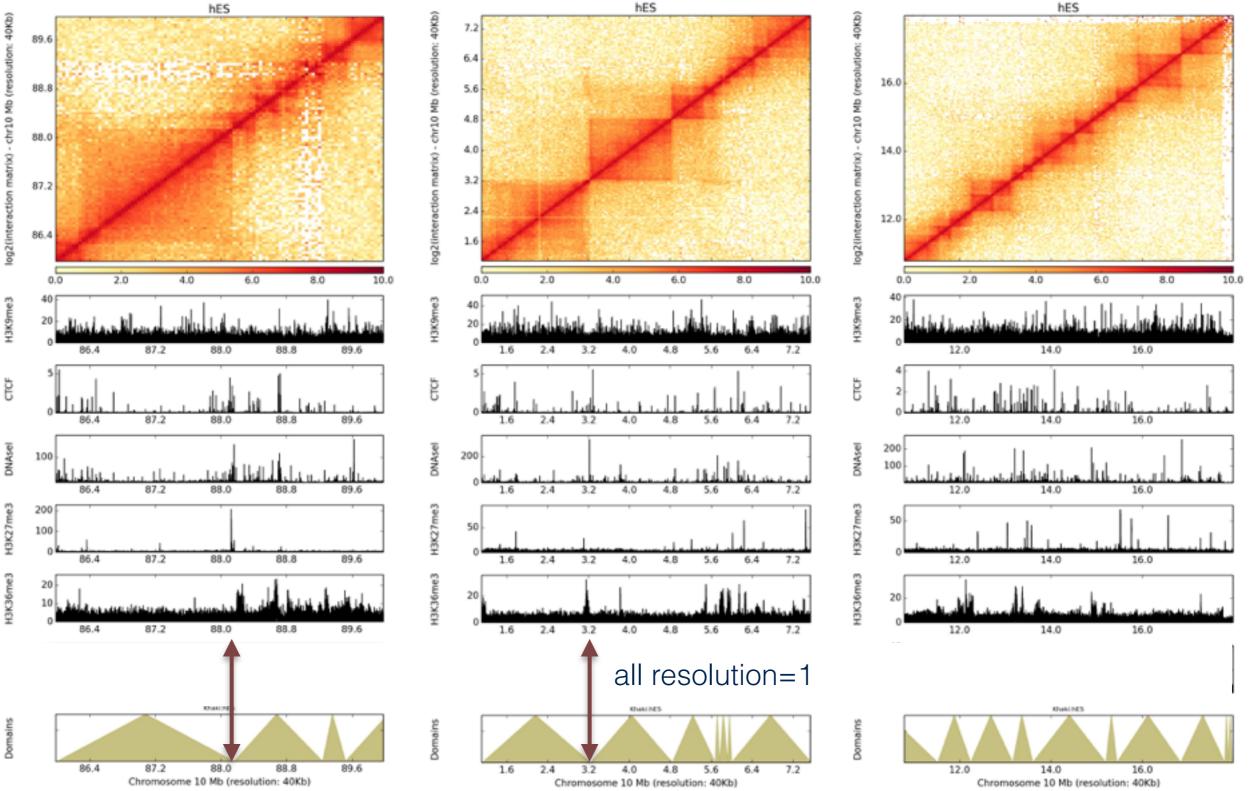




MrTAD Finder: a tool to identify TADs in multiple resolutions

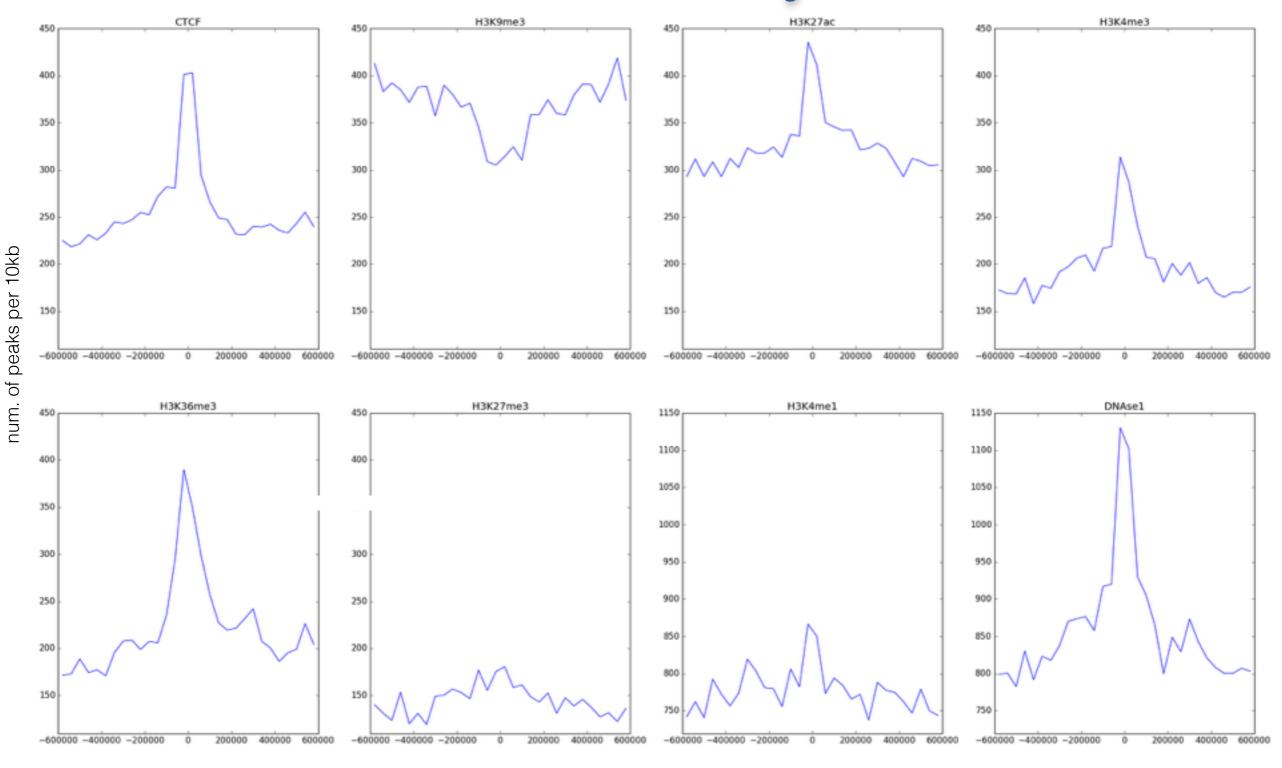


Relationship between TADs and chromatin features



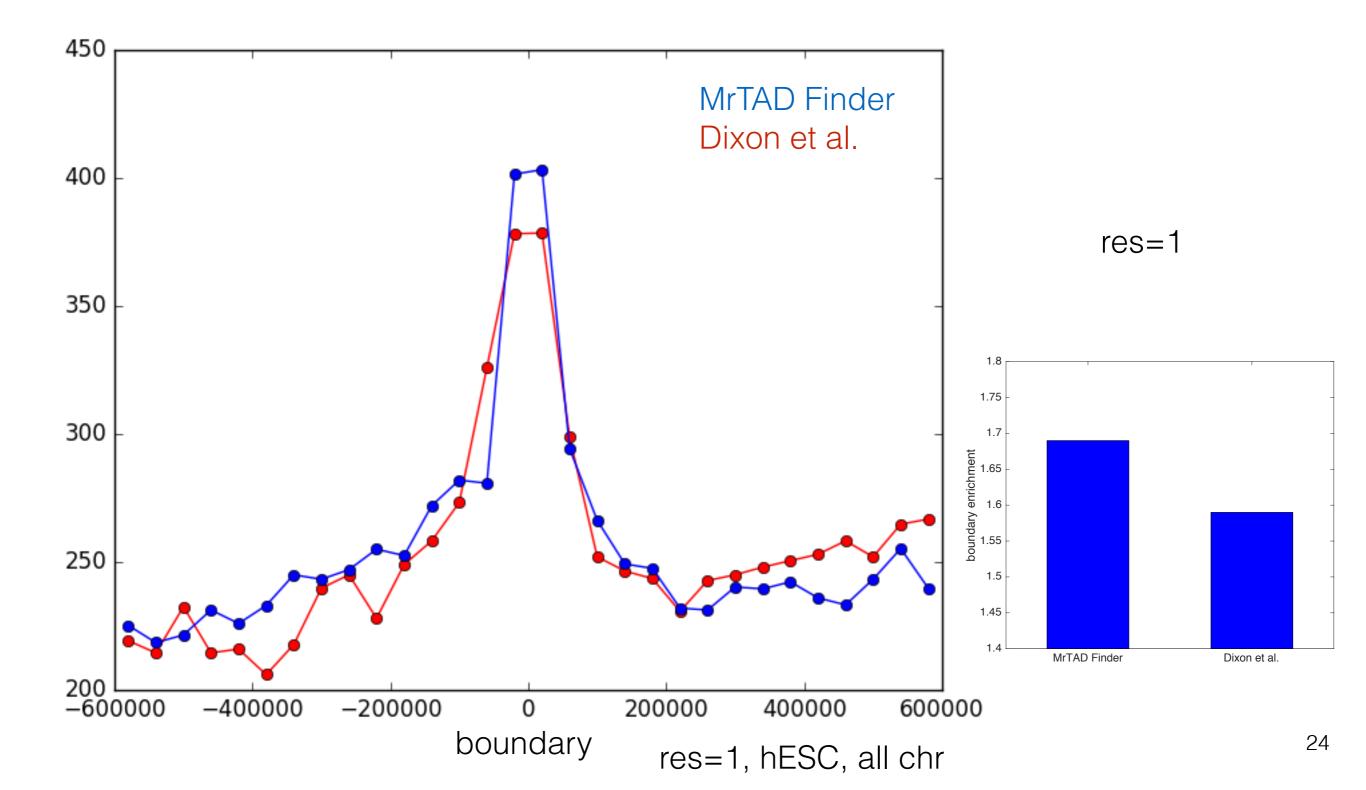
22

chromatin features near domain boundary

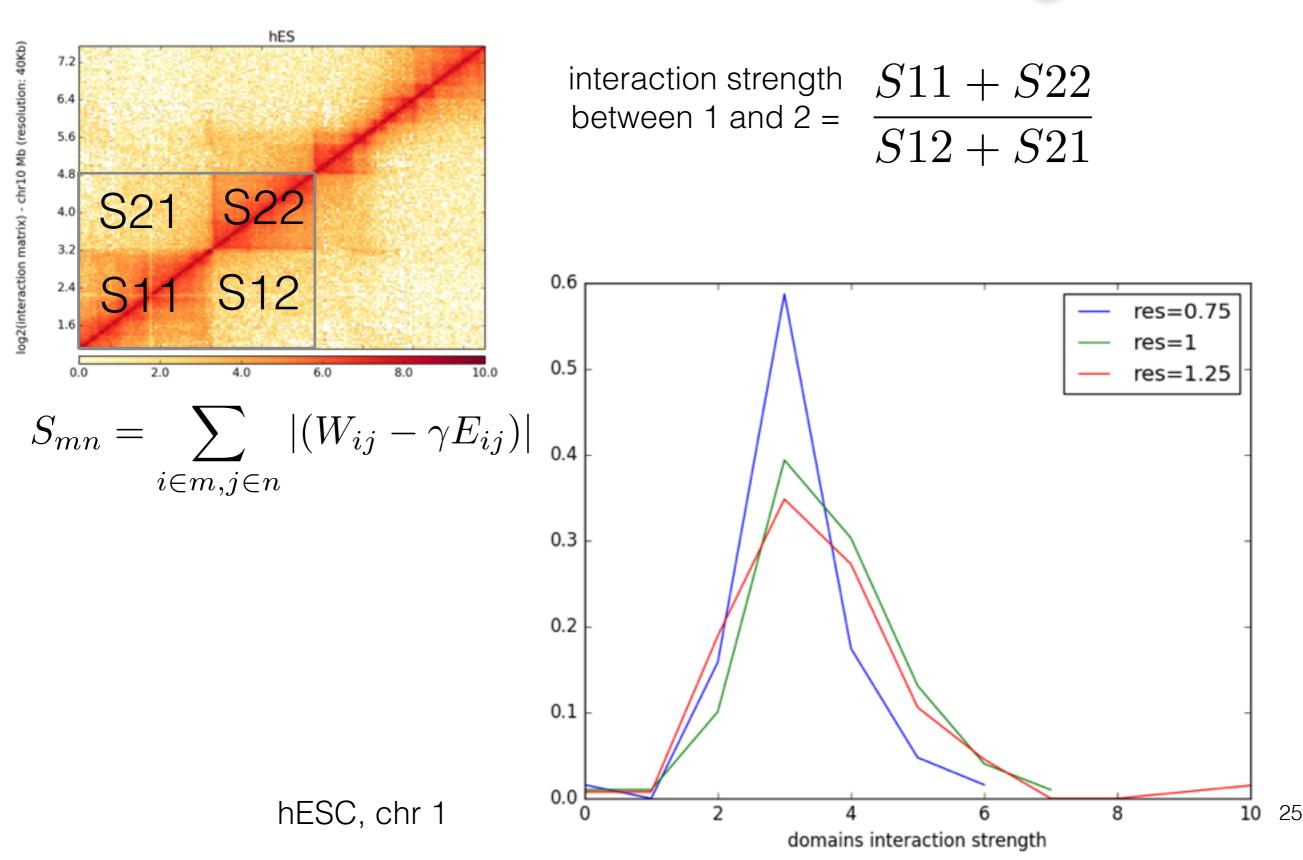


res=1, hESC, all chr

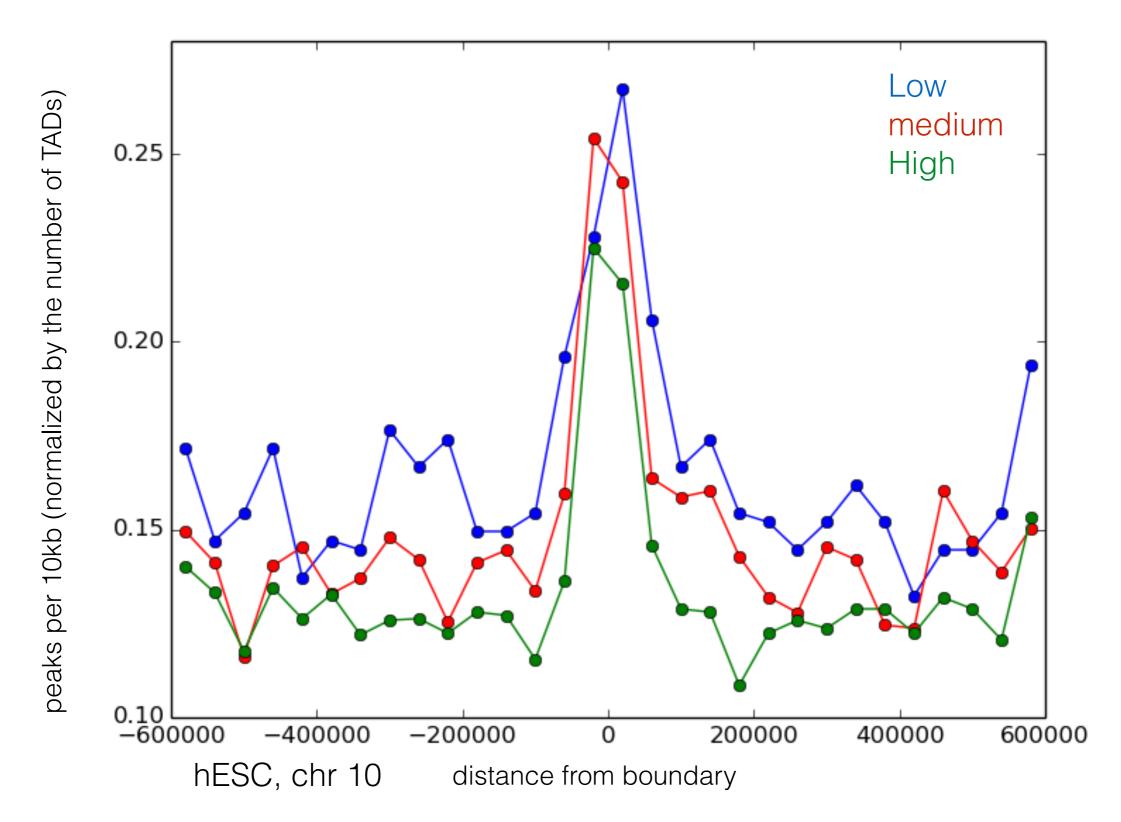
Comparison with existing algorithms



Domains interaction strength

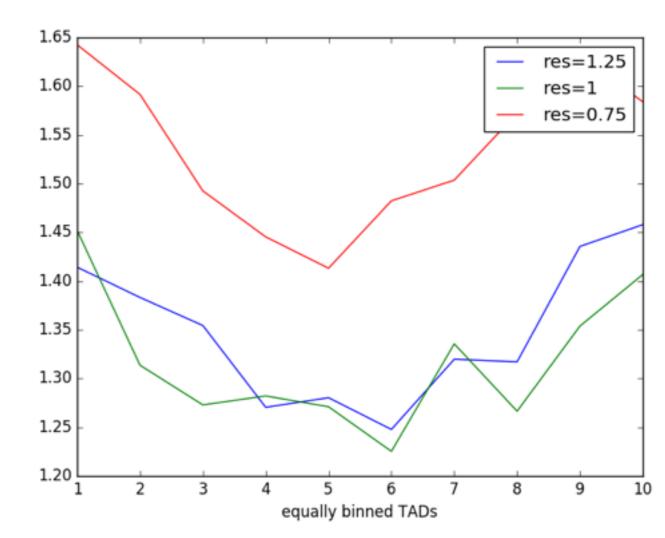


Boundary features with respect to resolution



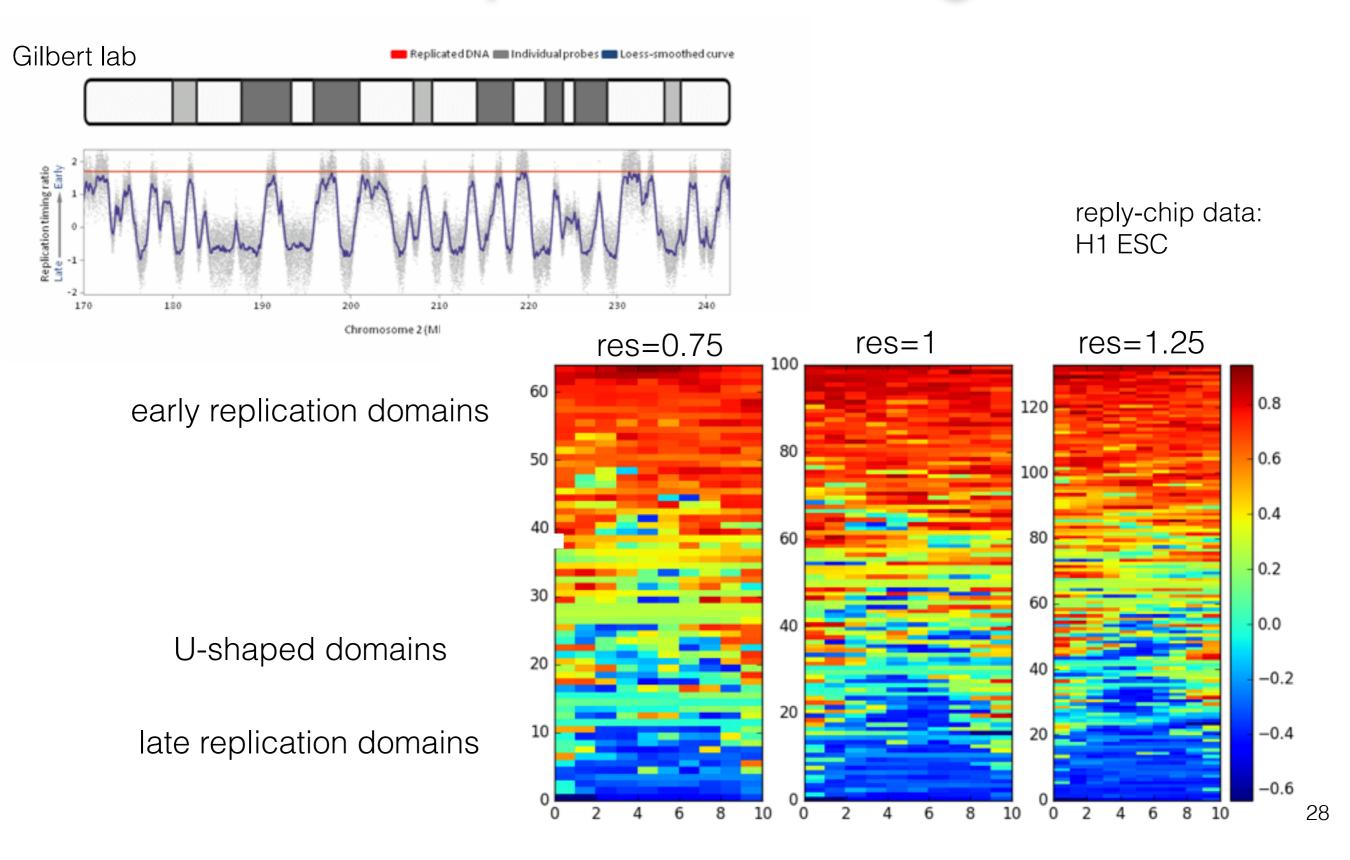
Chromatin signatures in different resolution

- various chromatin features, where are they distributed along the domains?
- effects of the resolutions? types of domains?
- enrichment of peaks/signals?

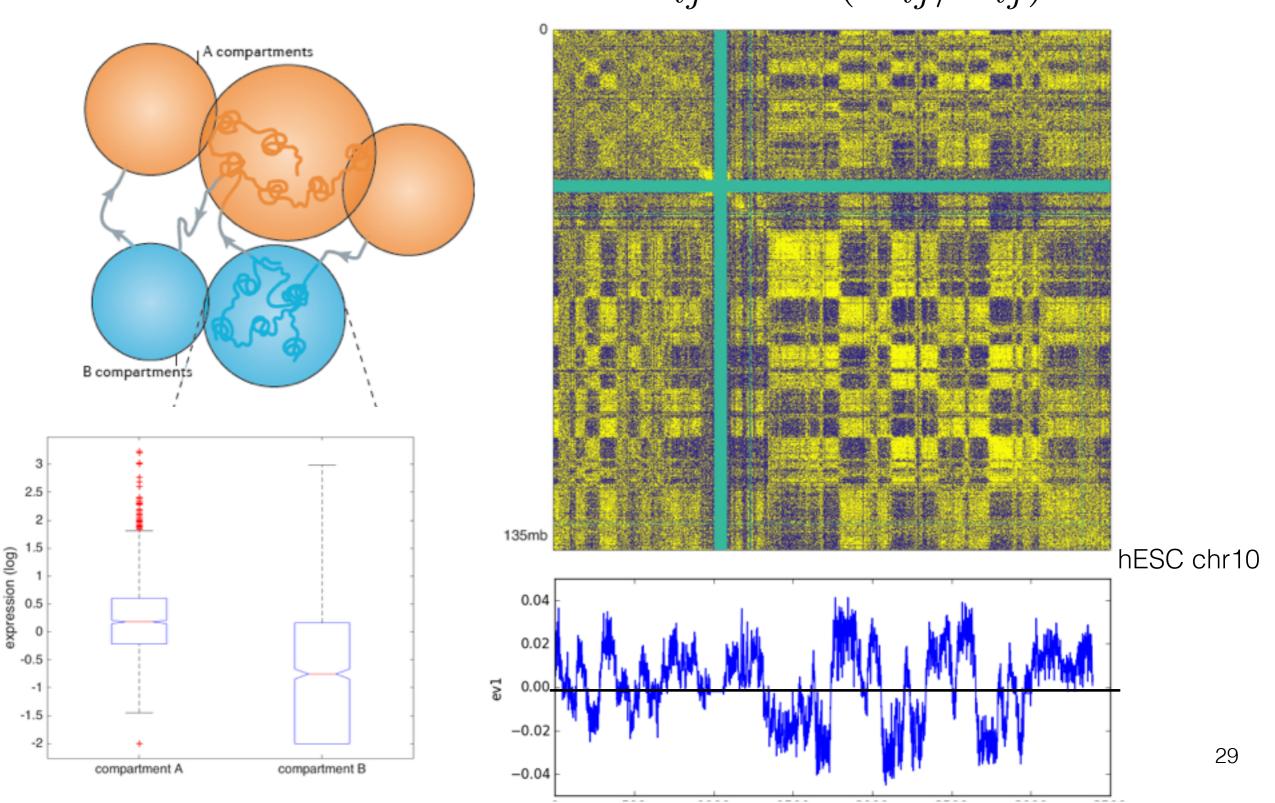


H3K27me3

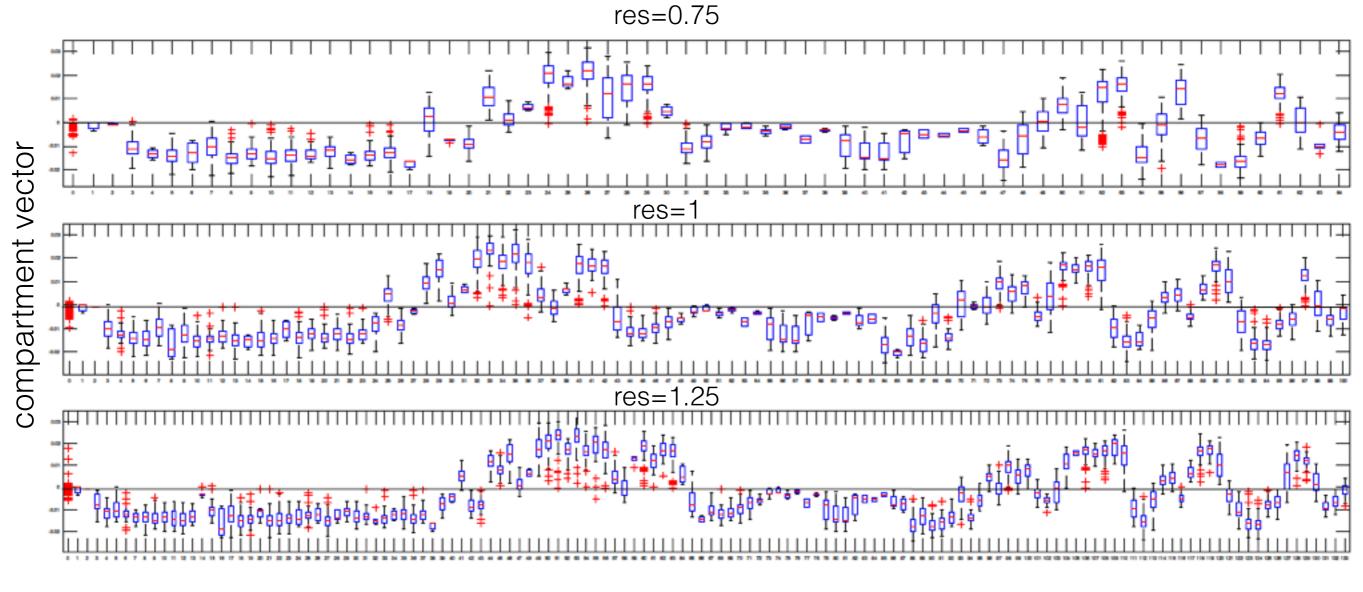
Replication timing



Compartments versus domains $C_{ij} = cor(W_{ij}/E_{ij})$



Compartments versus domains



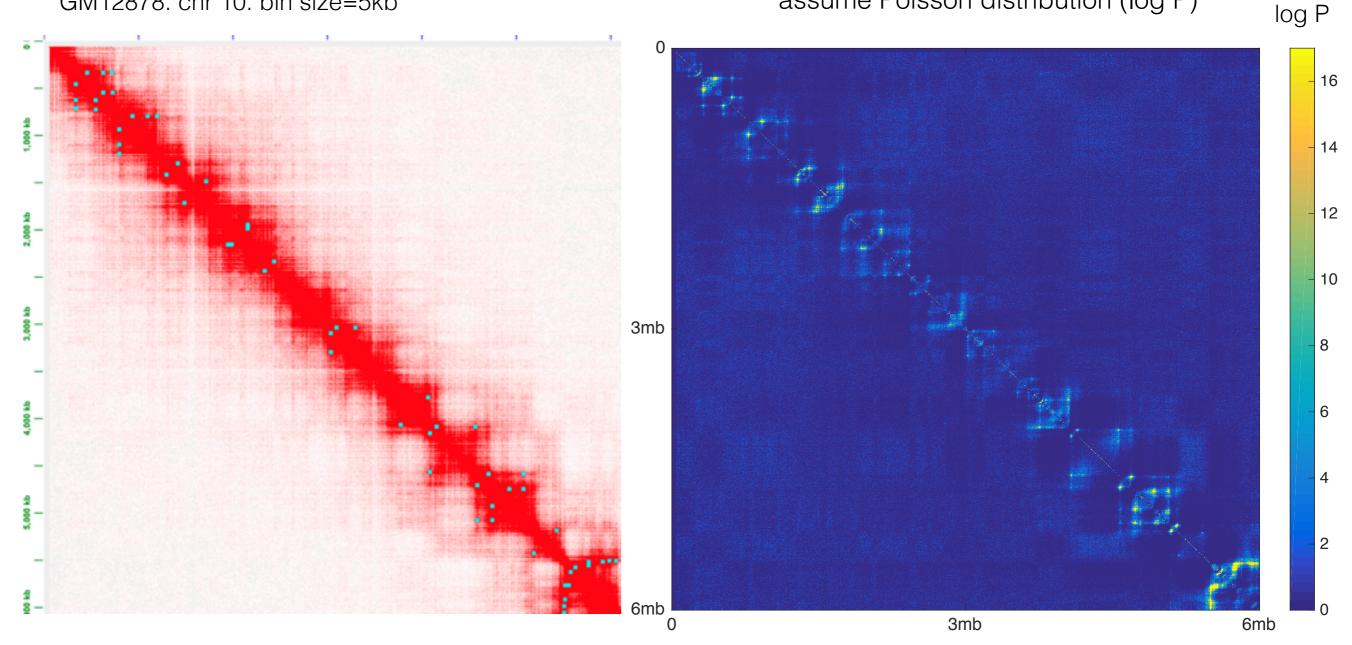
TADs

hESC chr1

Enrichment of contacts

GM12878. chr 10. bin size=5kb

cf. real contacts vs. null assume Poisson distribution (log P)



Likely to have many false positive because of over dispersion.

Summary and the next steps

- MrTAD Finder: a novel tool to identify TADs
 - take into account of a background that captures genomic distance (the idea could be used in other network context)
 - based on global optimization inspired by network modules as oppose to local approaches in existing methods
 - with a concept of continuous resolution, more general than a bottom-up hierarchical structure

Summary and the next steps

- Certain characteristic features for different resolutions
 - histone marks
 - expression (active and inactive domains)
 - replication timing (domain with multiple resolutions too?)
 - k-mer frequency in TADs across multiple resolutions (with ANS, Yunsi)
- Can interaction strength be reflected by chromatin features (combinatorially) near the boundary? CTCF orientation?
- Comparison between different cell types
- To compare with existing methods: Dixon et al. Nature 2012, Rao et al. Cell 2014, Weinreb and Raphael Bioinformatics 2015 (TADtree), Malik and Patro bioRxiv 2015 (Matryoshka)

Acknowledgement

- Tech
 - ANS, TG, JR, RK, AH, MG