

3D genome sequencing

Hi-C data analysis

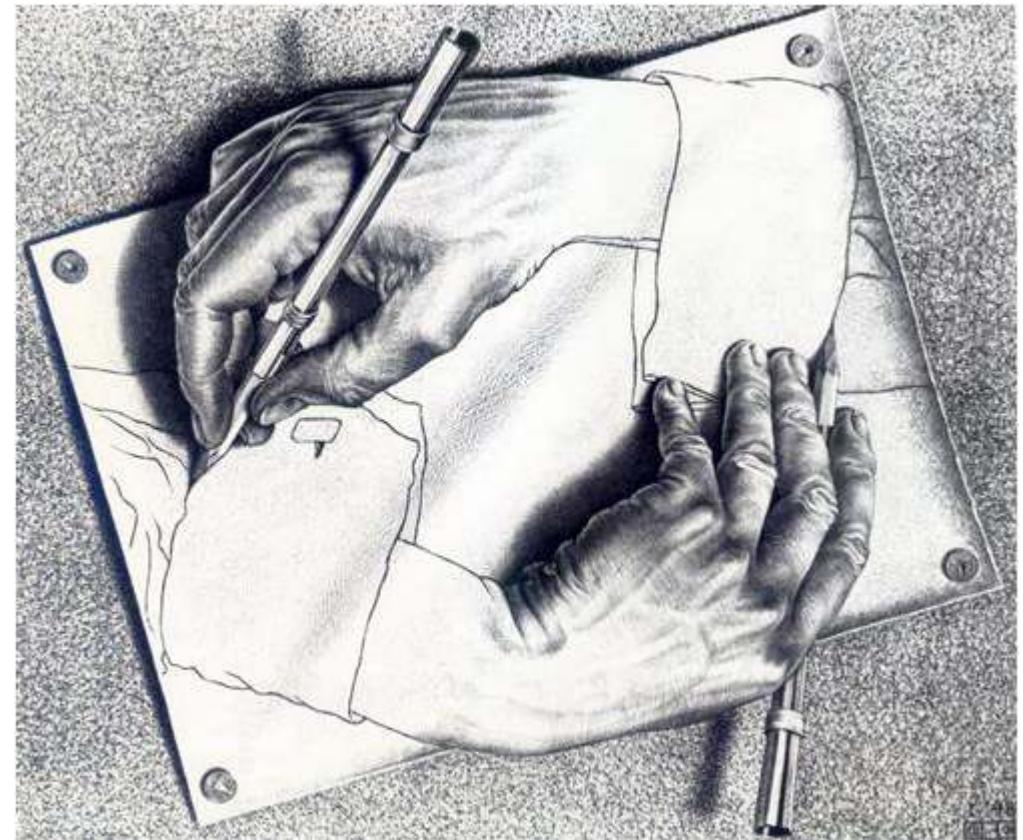


Sylvain Foissac, INRA Toulouse

GenoToul GetPlage - Mai 2016

Outline

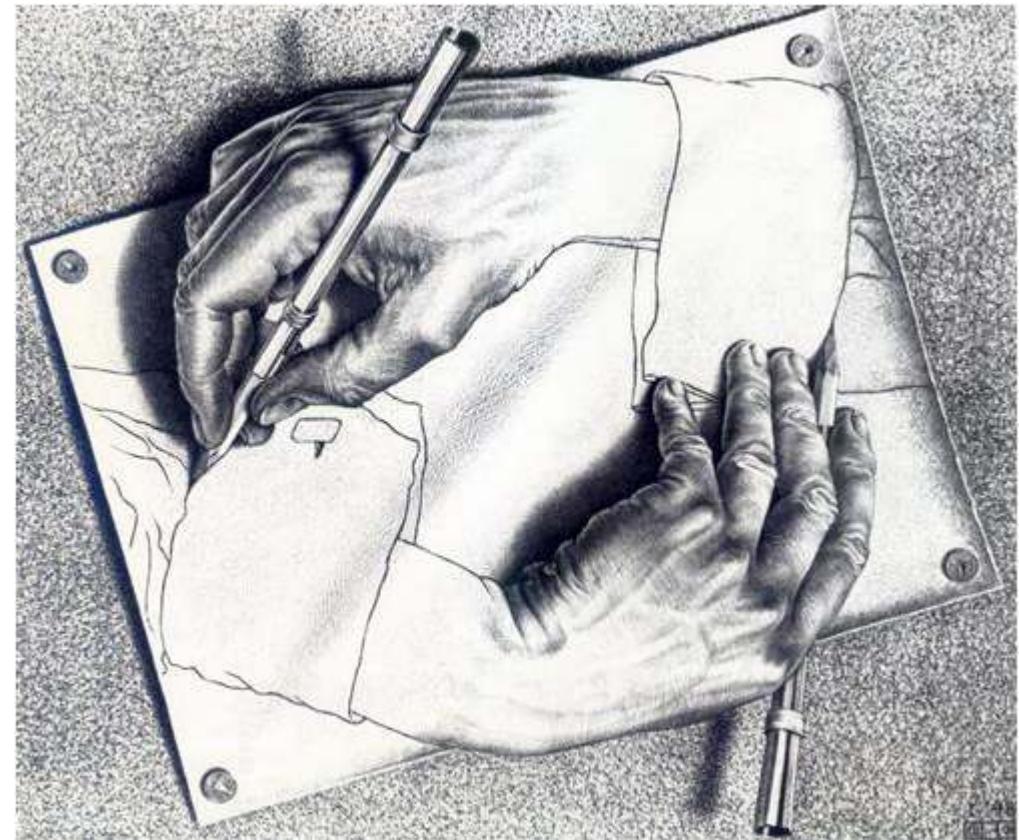
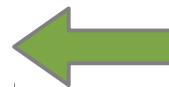
- Why
- What
- How
- Where
- Who



M.C. Escher, 1948

Outline

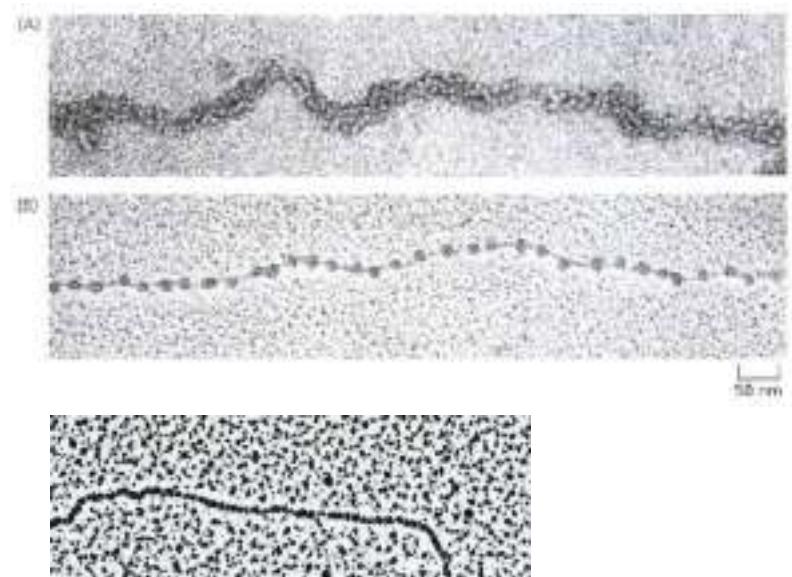
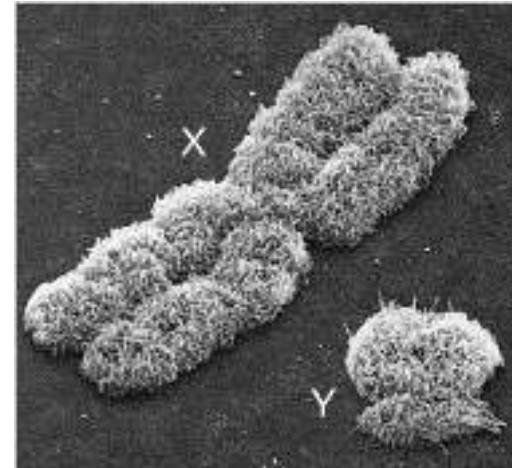
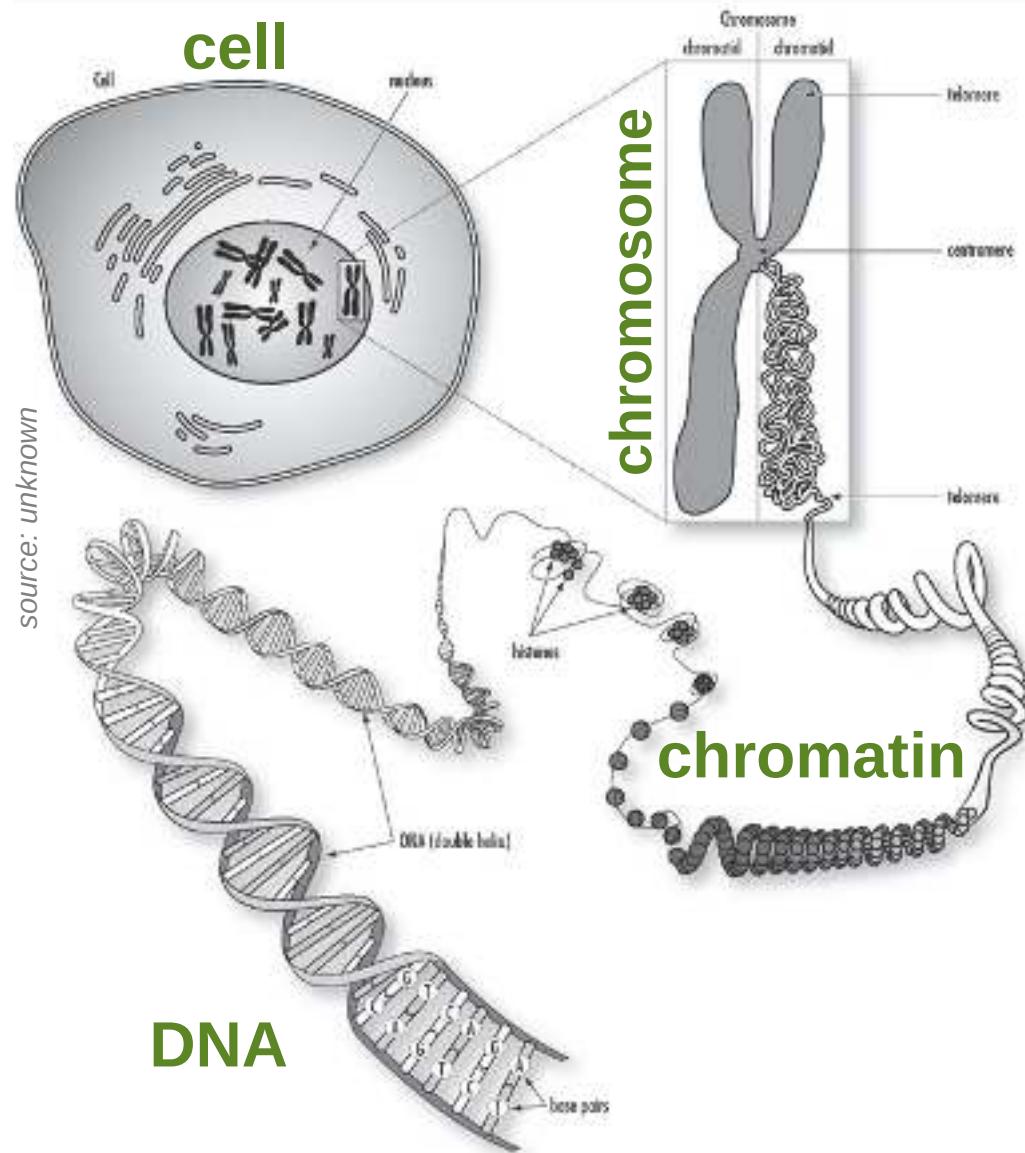
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M.C. Escher, 1948

Life, cell, chromosome & DNA

source: unknown

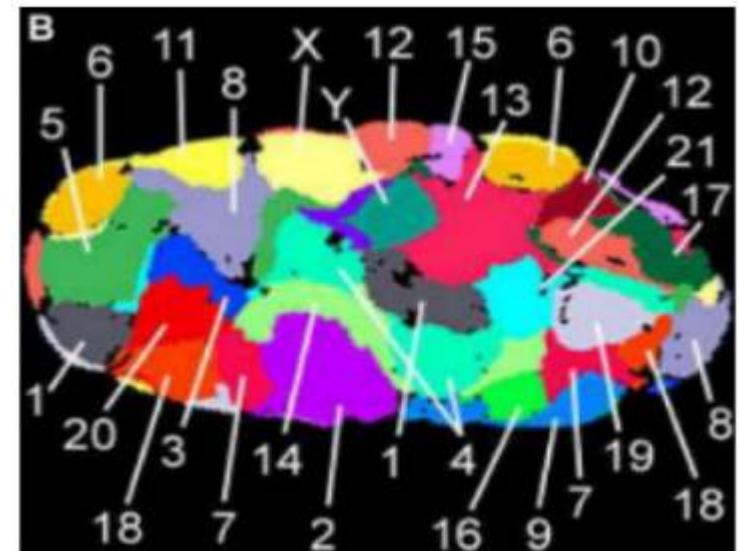


Life, cell, chromosome & DNA

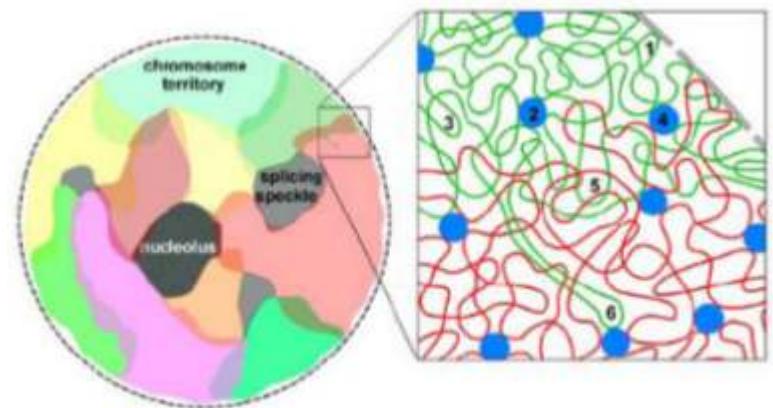


?

Life, cell, chromosome & DNA



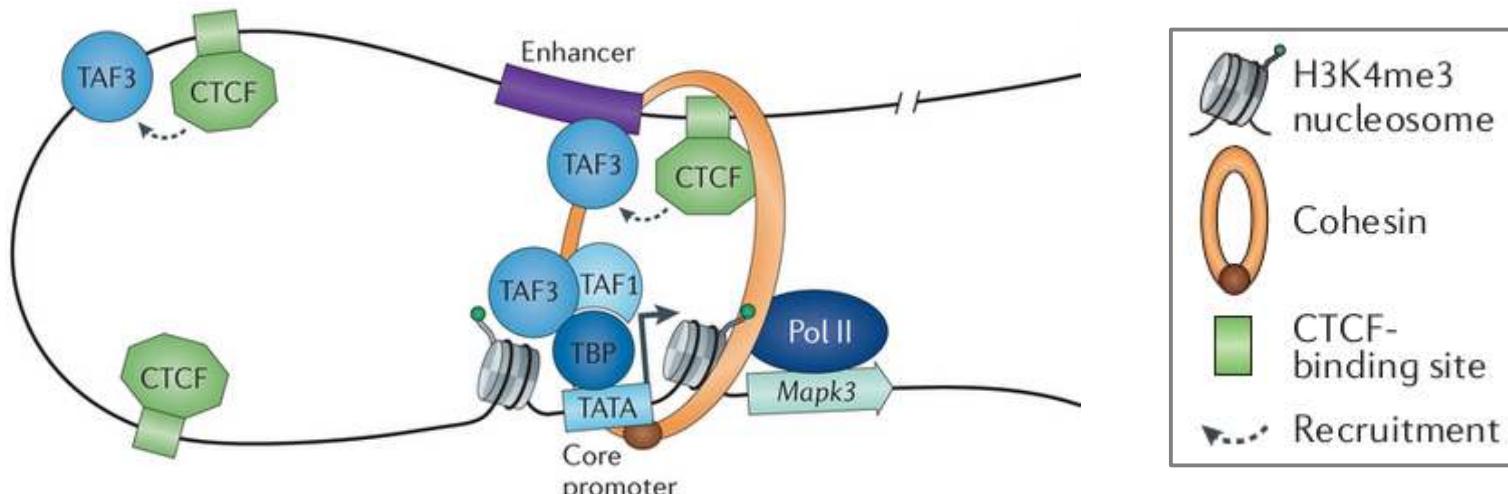
Bolzer A et al. 2005



Branco MR & Pombo A, 2006

Genome 3D structure is organized

From structure to function



Ong & Corces, *Nat. Rev. Genet.*, 2014

3D structure regulates gene expression

From structure to function

Chromosomal Contact Permits Transcription between Coregulated Genes

Stephanie Fanucchi,¹ Youtaro Shibayama,¹ Shaun Burd,¹ Marc S. Weinberg,^{3,4} and Musa M. Mhlanga^{1,2*}

¹Gene Expression and Biophysics Group, Synthetic Biology Emerging Research Area, Biosciences Unit, Council for Scientific and Industrial Research, Pretoria, Gauteng 0001, South Africa

²Unidade de Biofísica e Expressão Genética, Instituto de Medicina Molecular, Faculdade de Medicina, Universidade de Lisboa, Lisboa, 1649-026 Portugal

³Antiviral Gene Therapy Research Unit, Department of Molecular Medicine and Haematology, School of Pathology, Faculty of Health Sciences, University of the Witwatersrand, Johannesburg, Gauteng 2193, South Africa

⁴Department of Molecular and Experimental Medicine, The Scripps Research Institute, La Jolla, CA 92093, USA

*Correspondence: yod@mhlangalab.org

<http://dx.doi.org/10.1016/j.cell.2013.09.051>

SUMMARY

Transcription of coregulated genes occurs in the context of long-range chromosomal contacts that

2012). These highly sensitive assays can detect nascent mRNA and have revealed the FISH foci in a fraction of the population (2010; Fanucchi et al., 2010). This study

Nucleic Acids Research Advance Access published February 4, 2015

Nucleic Acids Research 2015 43: doi: 10.1093/nar/gkv046



Spatial re-organization of myogenic regulatory sequences temporally controls gene expression

Akihito Harada¹, Chandrashekara Mallappa², Seiji Okada¹, John T. Butler², P. Baker^{2,3}, Jeanne B. Lawrence², Yasuyuki Ohkawa^{1,2,*} and Anthony N. Imamura¹

¹Department of Advanced Medical Initiatives, JST-CREST, Faculty of Medicine, Kyushu University, Fukuoka 812-8582, Japan, ²Department of Cell and Developmental Biology, University of Massachusetts

³ S. A. Wiley, M. Z. Werle, G. E. B. 2010, 22, 5–18. DOI 10.1016/j.cell.2009.12.001.

and K. P. Paluszak and the Center for Learning Service (for genomics sequencing services). This work was also supported by grants from the Ministry of Education, Culture, Sports, Science and Technology of Japan.

TRANSCRIPTION

CTCF establishes discrete functional chromatin domains at the *Hox* clusters during differentiation

Vasun Narendra,^{1,2} Pedro P. Rocha,³ Didi An,⁴ Ramya Ravichandran,⁵ Esteban O. Maciaozzi,⁴ Danny Reinberg,^{1,2*}

Polycomb and Trithorax group proteins encode the epigenetic identity by establishing inheritable domains of repressive and active chromatin within the *Hox* clusters. Here we demonstrate that the CCCTC-binding factor (CTCF) functions



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Next Genera
USA Congre

27–28 October 2015

Spatial enhancer clustering and regulation of enhancer-proximal genes by cohesin

Elizabeth Ing-Simmons^{1,2,7}, Vlad C. Seitan^{1,7}, Andre J. Faure^{3,8}, Paul Flicek^{3,4},

Disruptions of Topological Chromatin Domains Cause Pathogenic Rewiring of Gene-Enhancer Interactions

Dario G. Lupiáñez,^{1,2} Katerina Kraft,^{1,2} Verena Heinrich,² Peter Krawitz,^{1,2} Francesco Brancati,³ Eva K. Denize Horn,² Hülya Kayserili,⁵ John M. Opitz,⁶ Renata Laxova,⁶ Fernando Santos-Simarro,^{7,8} Brigitte Gilbert-Dussardier,⁹ Lars Wittler,¹⁰ Marilena Borschluwer,¹ Stefan A. Haas,¹¹ Marco Osterwalder,¹² Bernd Timmermann,¹³ Jochen Hecht,^{1,14} Malte Spielmann,^{1,2,14} Axel Visel,^{12,15,16} and Stefan Mundlos¹

¹Max Planck Institute for Molecular Genetics, RG Development & Disease, 14195 Berlin, Germany

²Institute for Medical and Human Genetics, Charité Universitätsmedizin Berlin, 13353 Berlin, Germany

³Medical Genetics Unit, Policlinico Tor Vergata University Hospital, 00133 Roma, Italy

Nuclear Aggregation of Olfactory Receptor Genes Governs Their Monogenic Expression

E. Josephine Clowney,¹ Mark A. LeGros,^{2,4} Colleen P. Eirene C. Markenskoff-Papadimitriou,³ Markko Myllys,¹ and Stavros Lomvardas^{1,2,3,*}

¹Program in Biomedical Sciences

Leading Edge
Previews

A CRISPR Connection between Chromatin Topology and Genetic Disorders

Wing Man Chan, Jun Li, Paul D. Zhou
Scripps Research Institute, La Jolla, CA 92093, USA
Yannick J. Lutzow, Daniel R. Lieberman
Massachusetts General Hospital, Boston, MA 02114, USA
Eduardo D. Simeone
University of California, San Francisco, CA 94158, USA

Structural variations are common in the human genome, but their contribution to human disease is not well understood. Lomvardas et al. demonstrate that some structural variants can disrupt gene expression by altering the spatial organization of chromatin. In the *OR52B4* gene, located on chromosome 10, a deletion of 10 kb creates a new transcription start site, which increases the expression of the gene. This leads to increased protein levels, which in turn results in altered spatio-temporal gene expression patterns, altered gene-gene interactions, and developmental disorders.

3D structure regulates gene expression

*Physical Biosciences Division, Lawrence Berkeley National Laboratory, Berkeley, CA 94720, USA

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Massachusetts General Hospital, Boston, MA 02114, USA
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Hi-C data analysis

GenoToul GetPlage, INRA Toulouse - Mai 2016

From structure to function

Cell

Chromosomal Contact Permits Transcription between Coregulated Genes

Cell

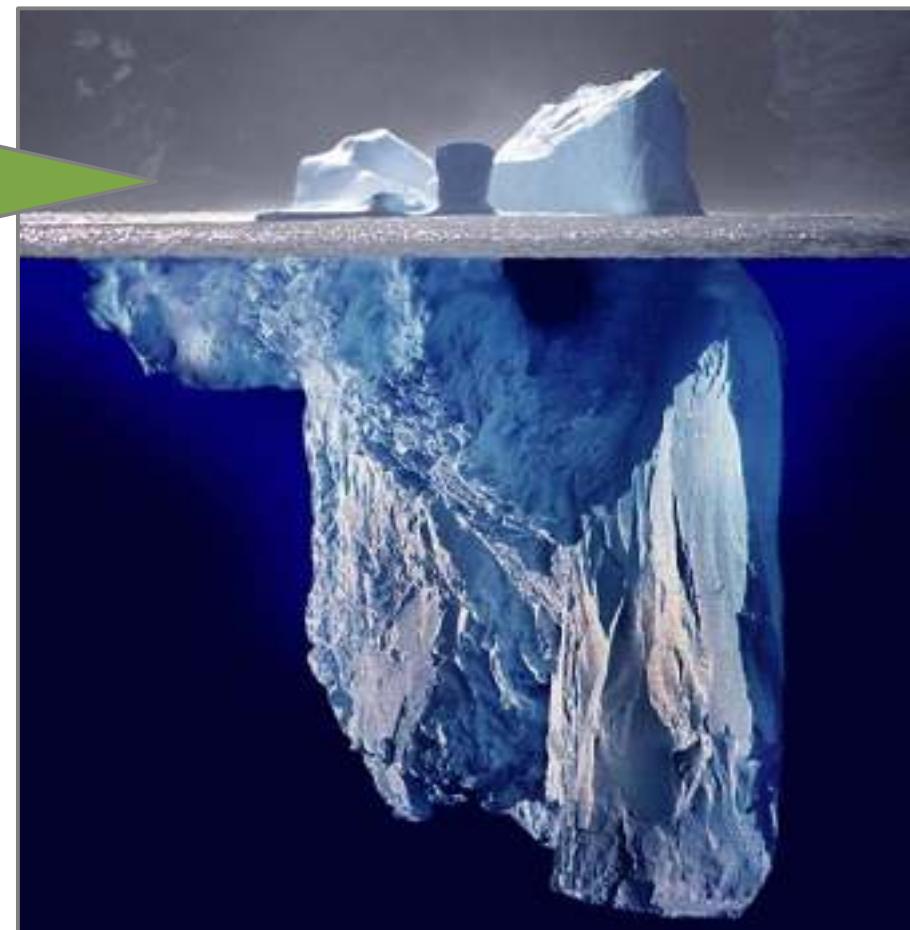
Spatial enhancer clustering and regulation of enhancer-promoter genes by cohesin

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CTCF sets establishes discrete functional chromatin domains at the Zox clusters during differentiation

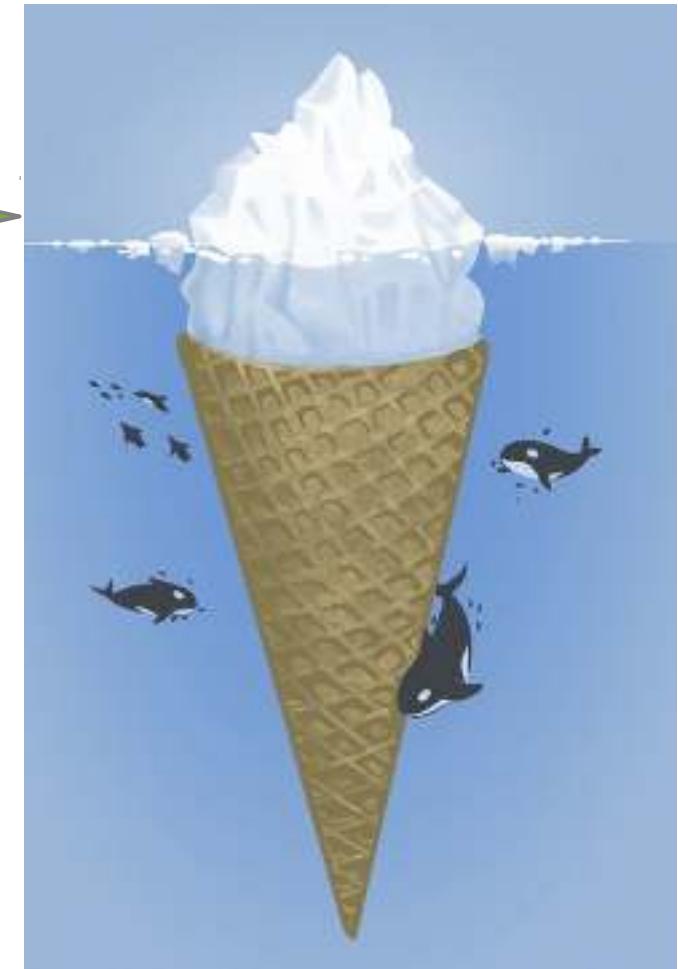
Cell

Nuclear Aggregation of Cleftatory Receptor Genes Go Their Monogenic Way



3D structure regulates gene expression

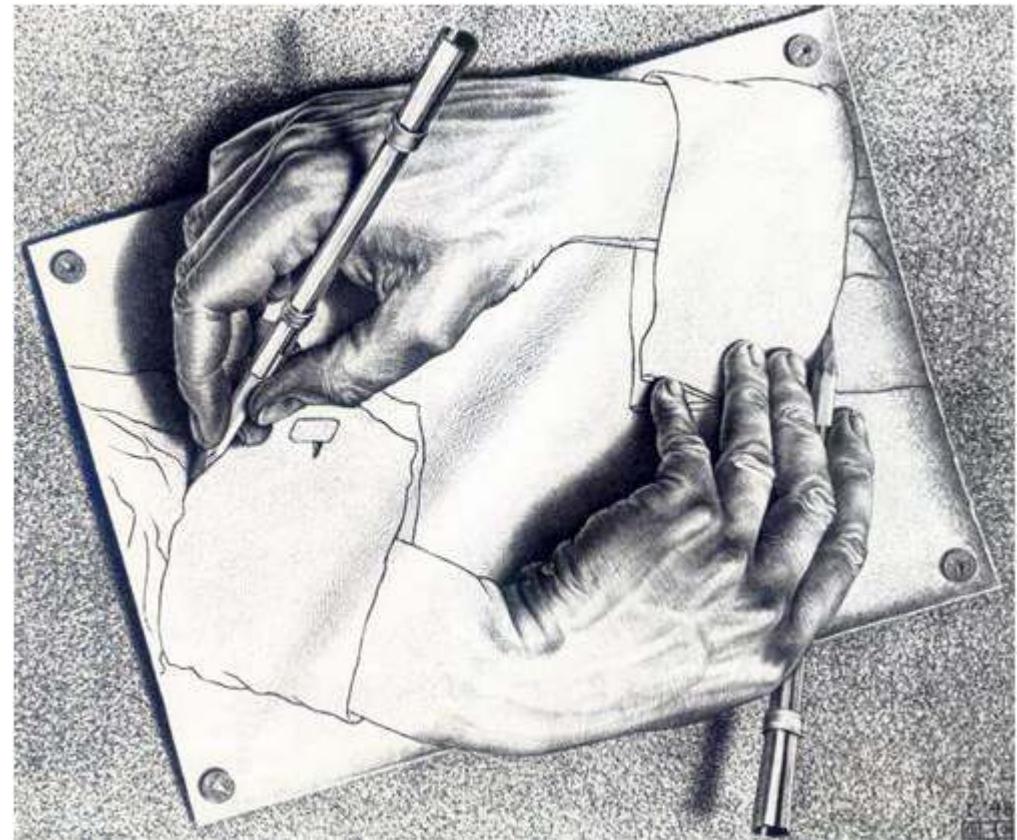
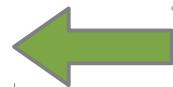
From structure to function



3D structure regulates gene expression

Outline

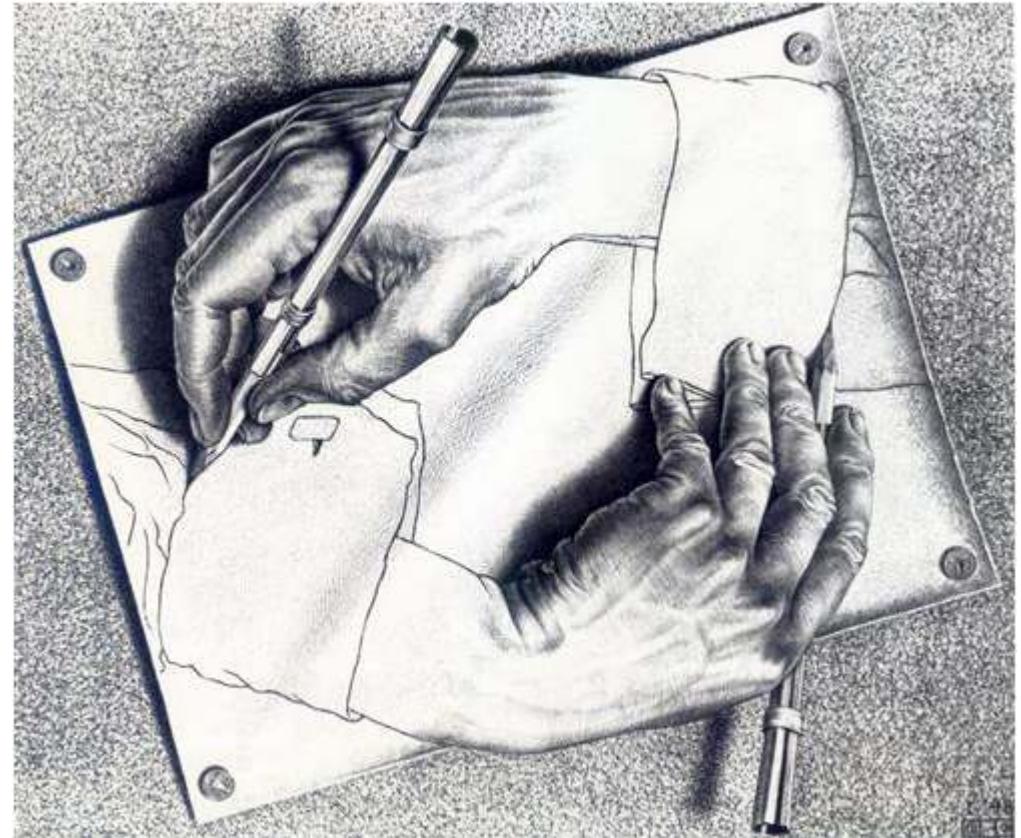
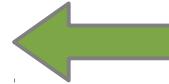
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M.C. Escher, 1948

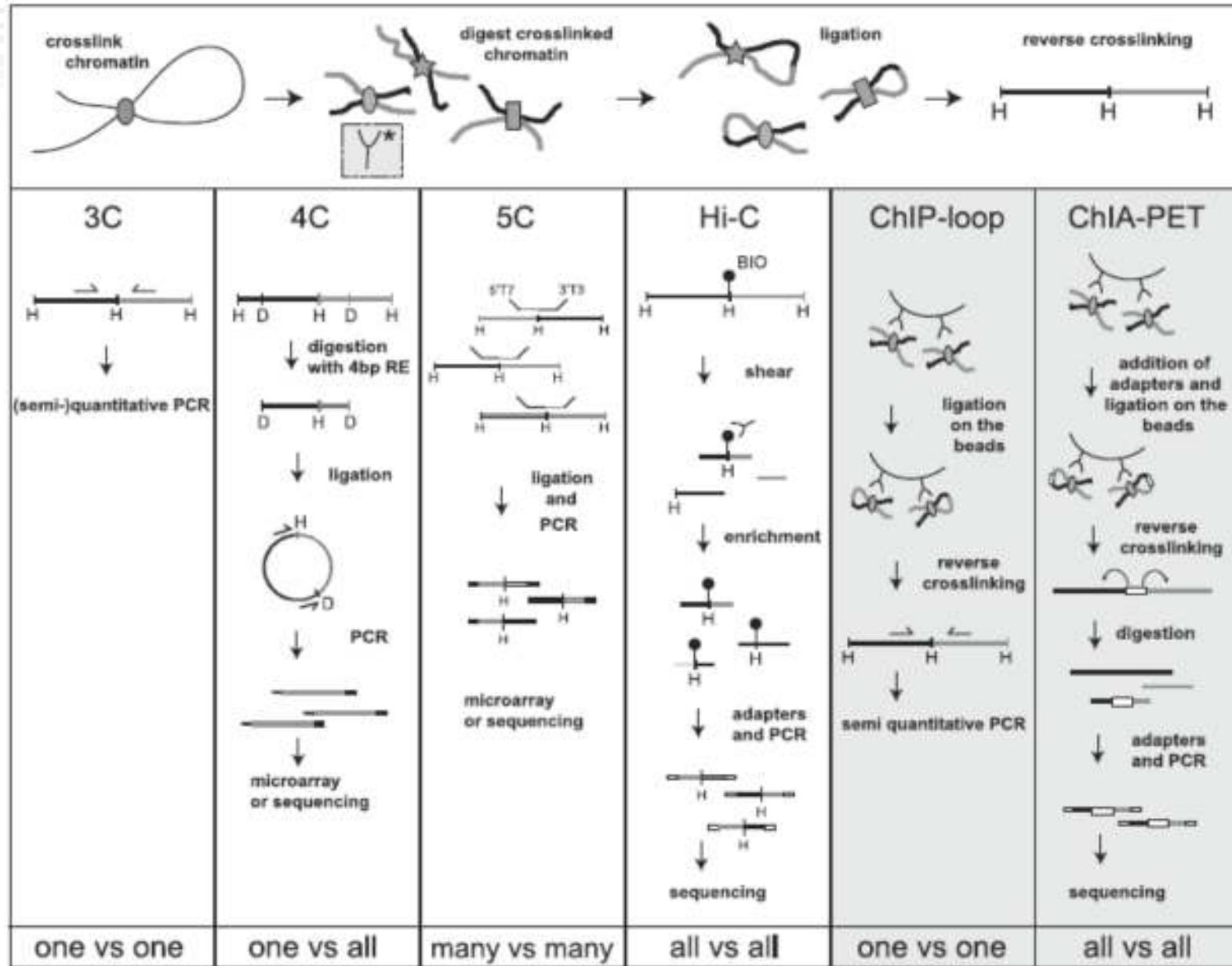
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M.C. Escher, 1948

Chromosome Conformation Capture assays

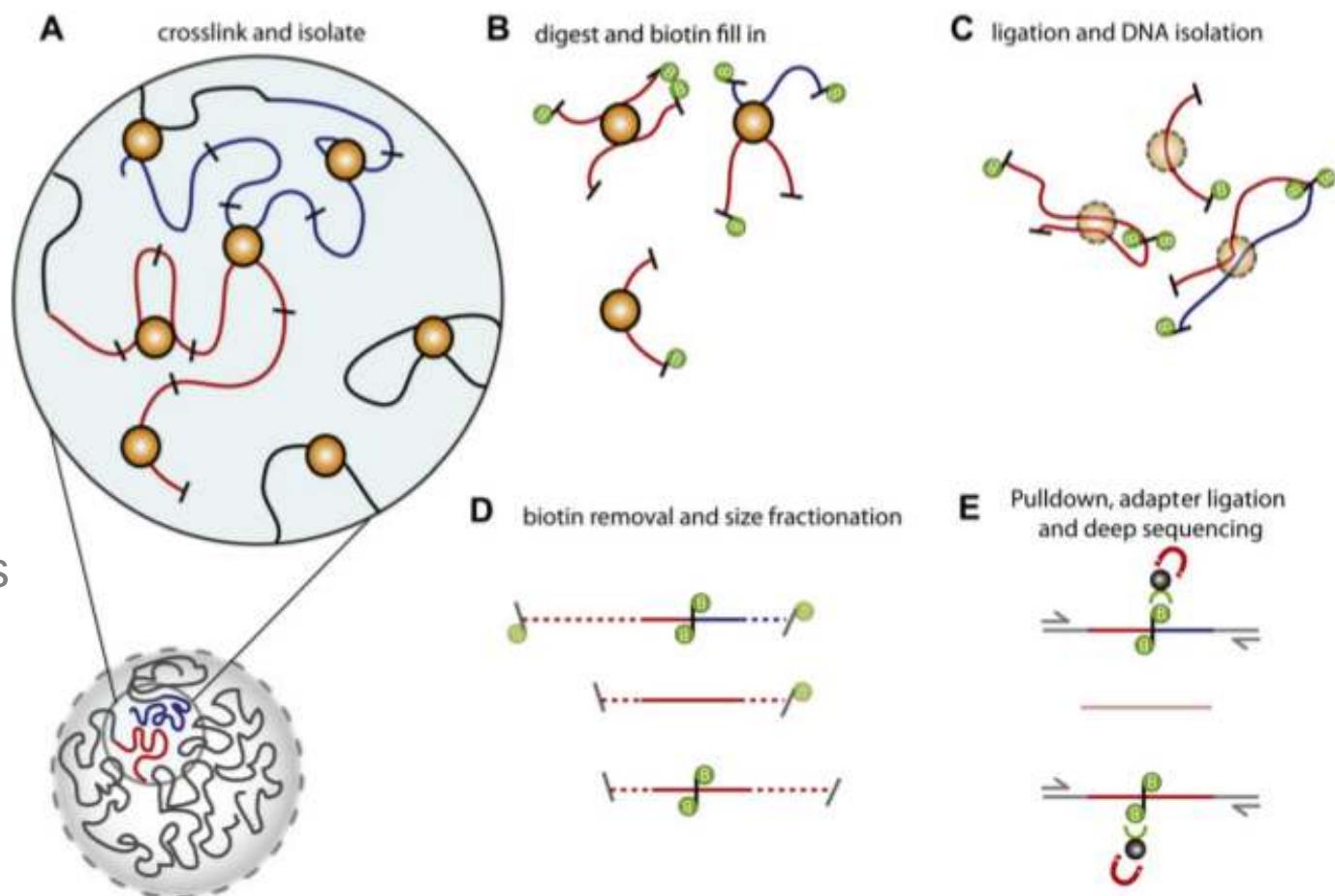


Hi-C: the experiment

Hi-C: high-throughput chromatin conformation capture
(Lieberman-Aiden et al, Science, 2009, Rao et al, Cell, 2014)

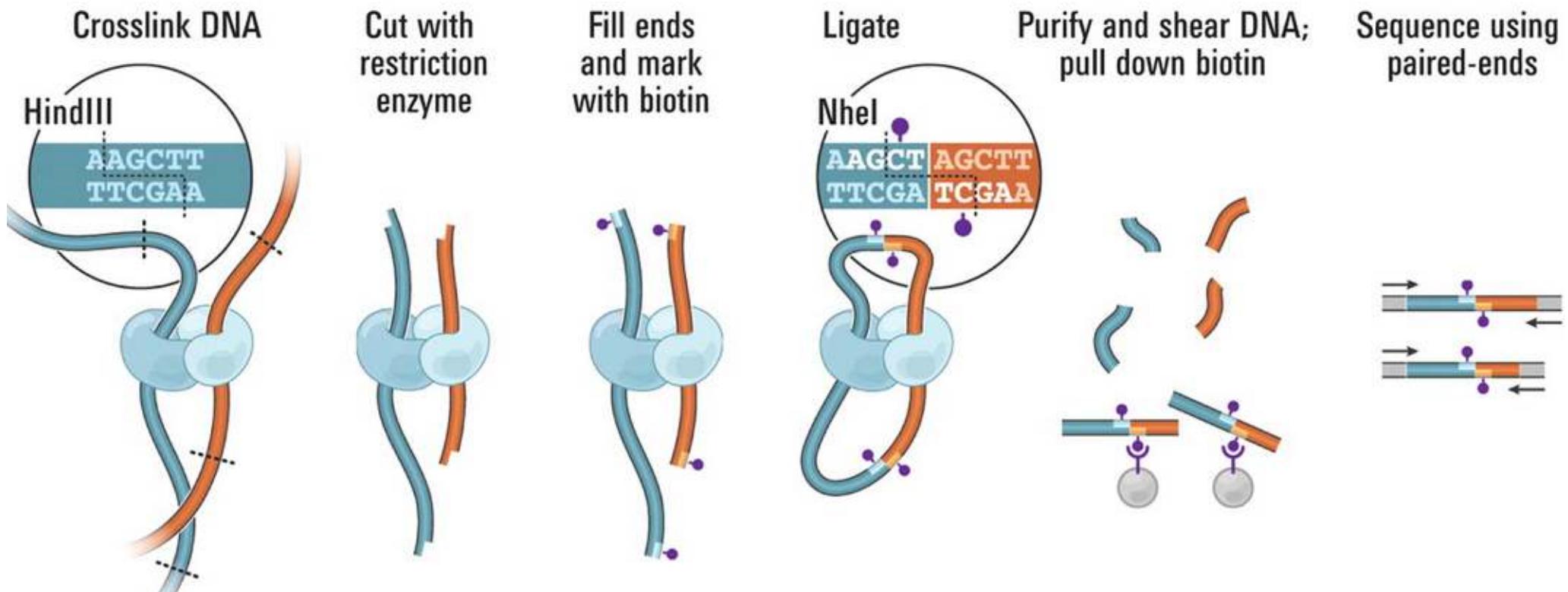
J.-M. Belton et al./Methods 58 (2012) 268–276

- ◆ crosslink DNA (“fixation”)
- ◆ cleave genome with restriction enzyme
- ◆ biotin-mark and ligate extremities
- ◆ fragment, select biotin-marked junctions
- ◆ sequence fragments (paired-ends)



Hi-C: the experiment

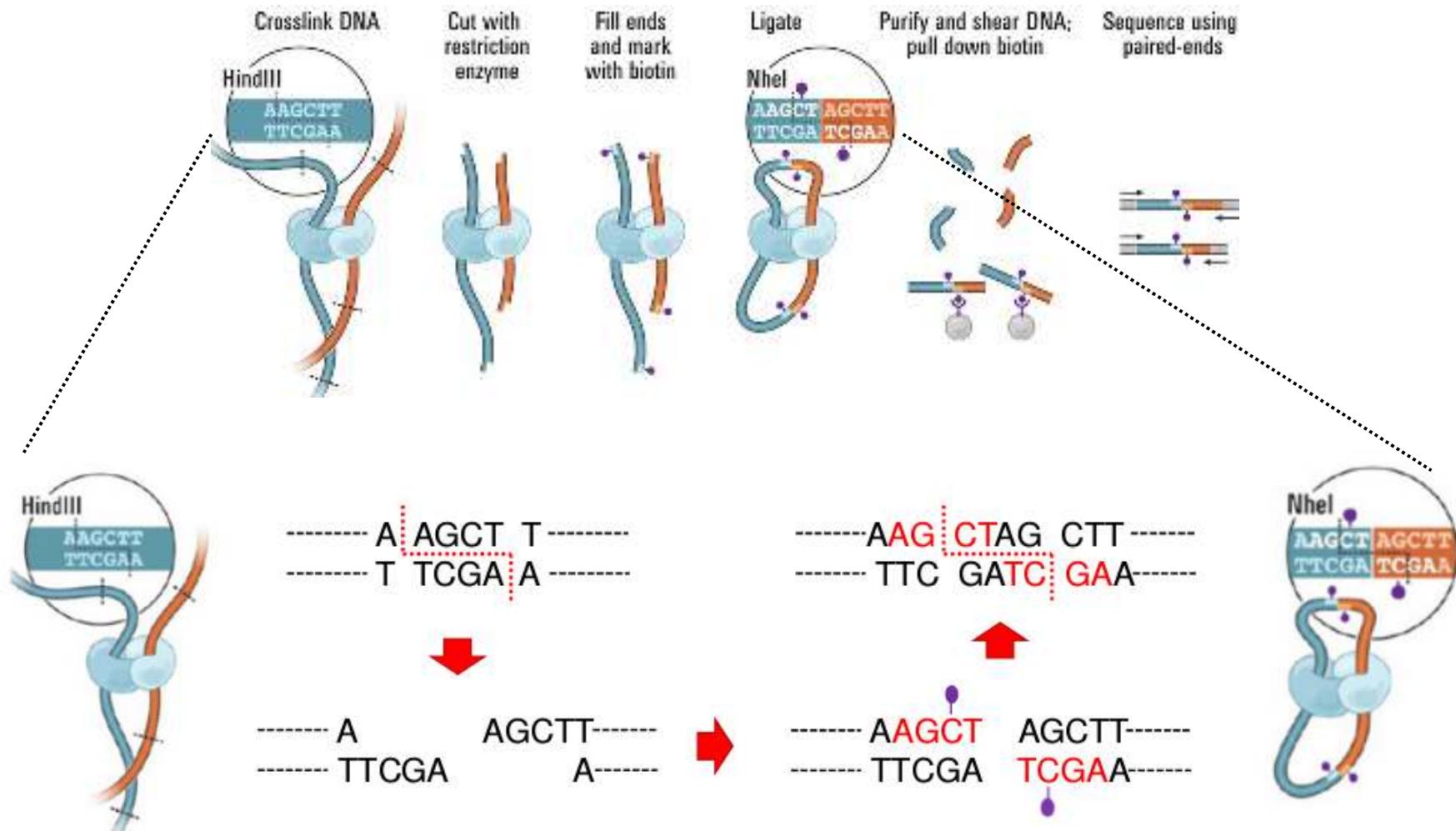
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Rao et al, Cell, 2014

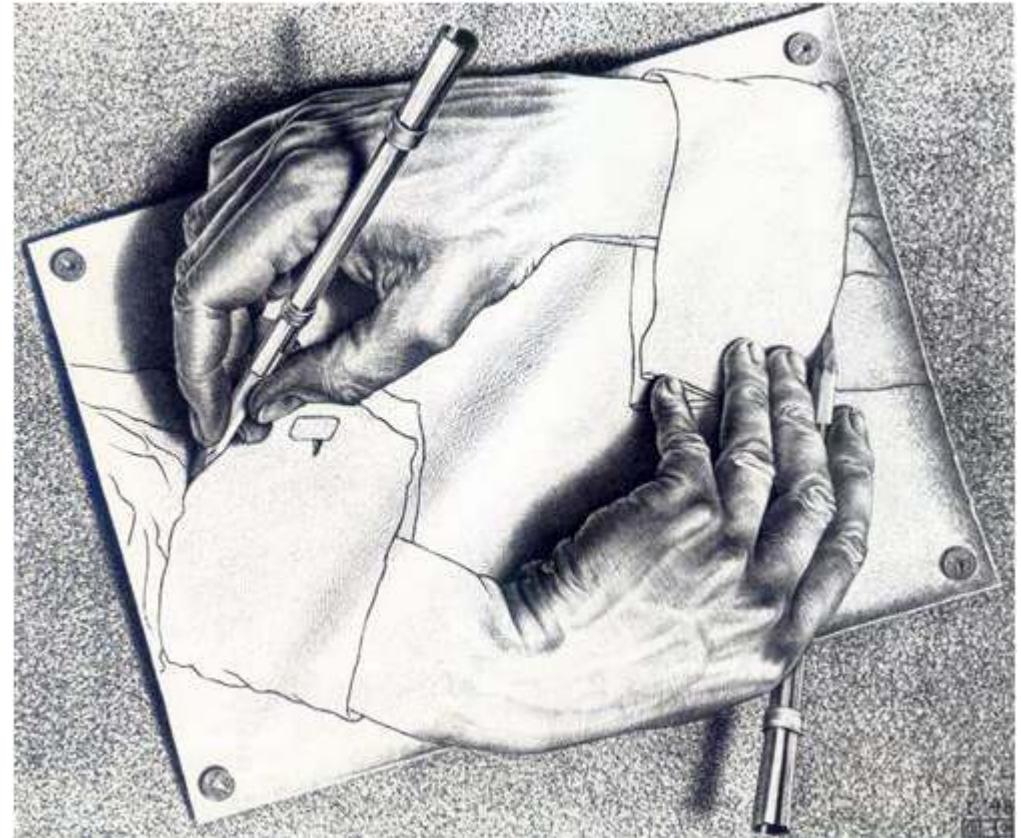
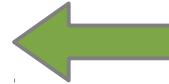
Hi-C: the experiment

Hi-C: high-throughput chromatin conformation capture
(Lieberman-Aiden et al, Science, 2009, Rao et al, Cell, 2014)



Outline

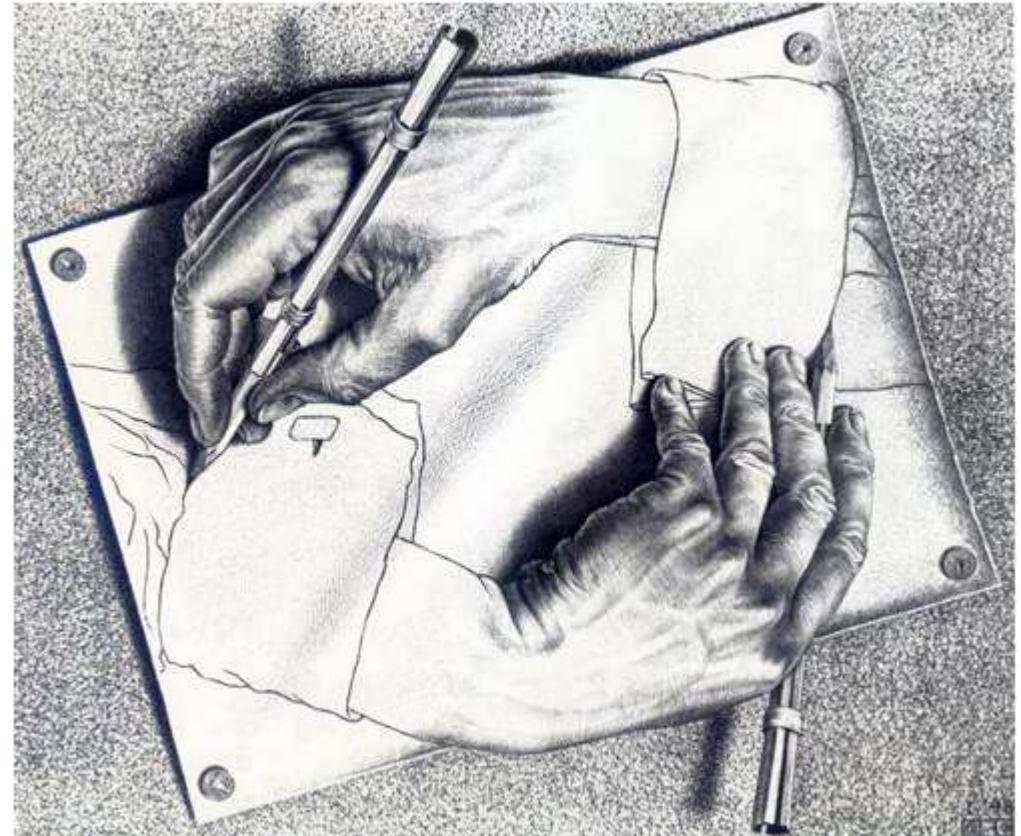
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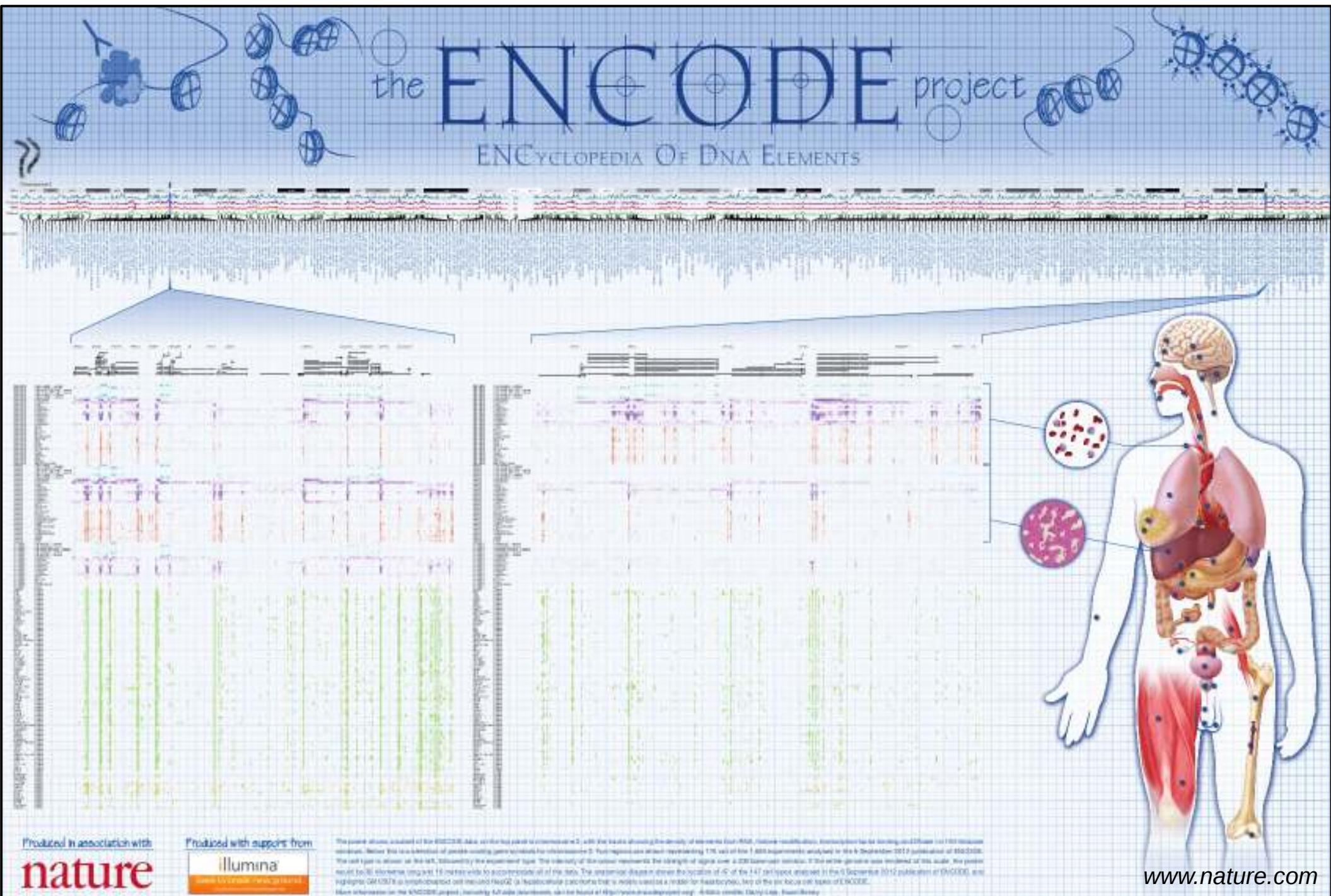
M.C. Escher, 1948

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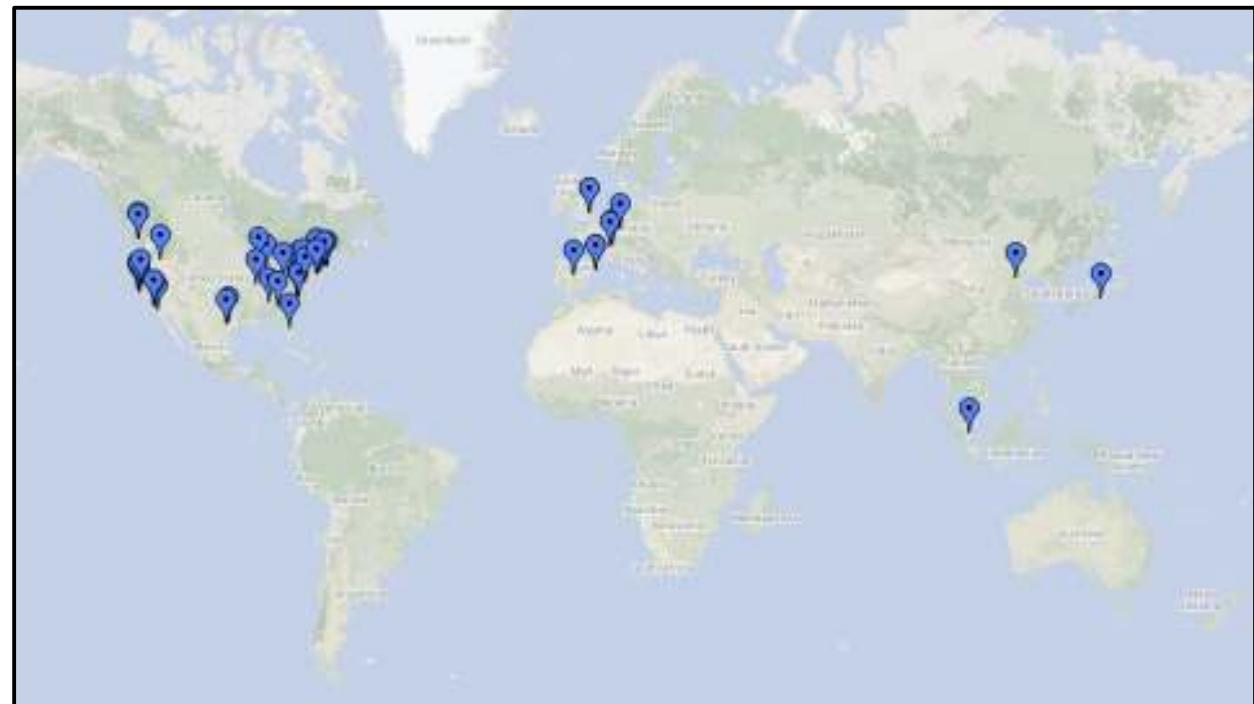


M.C. Escher, 1948

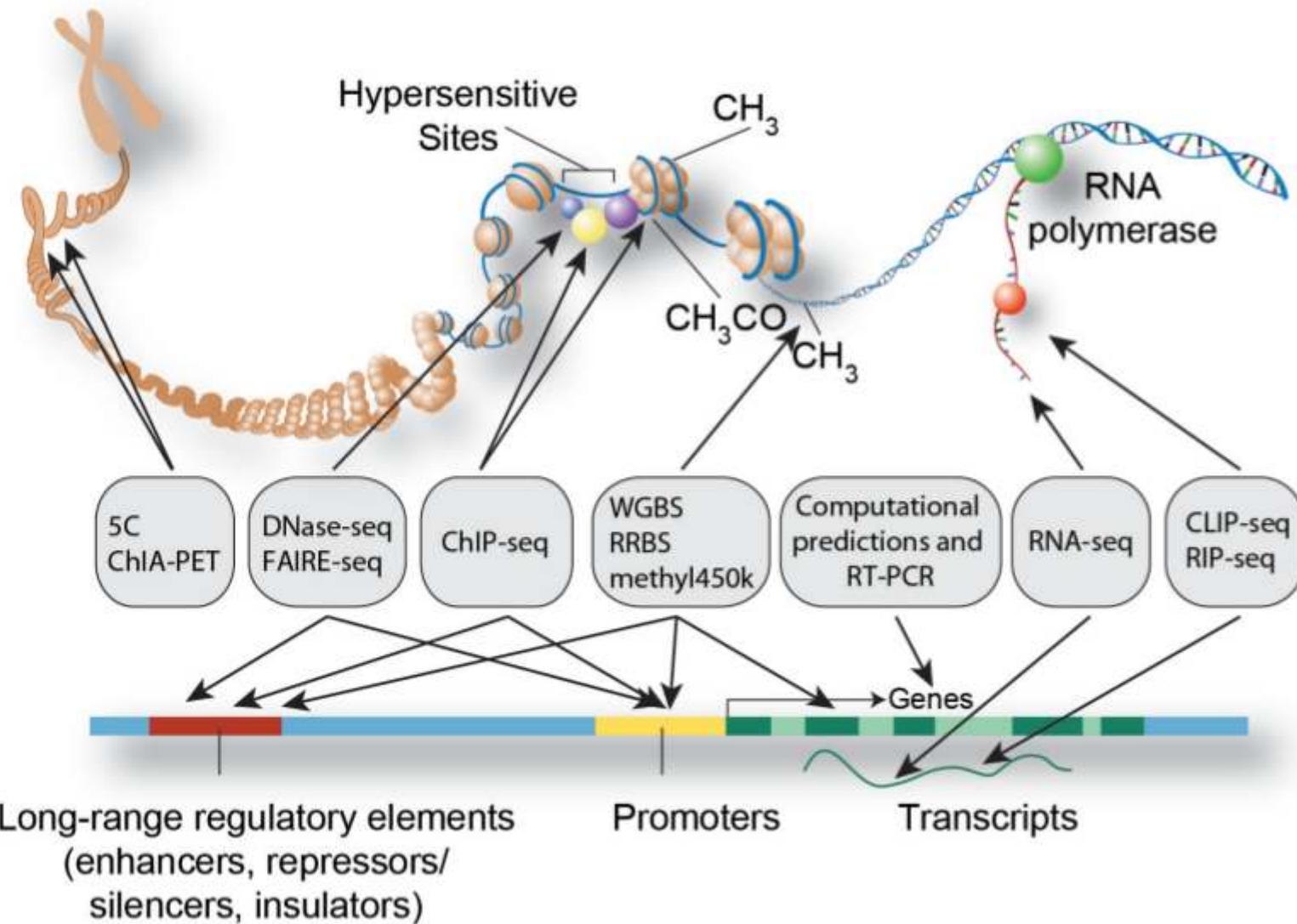


The ENCODE project

- ◆ Encyclopedia of DNA elements
- ◆ goal: characterize (annotate) all functional elements of the human genome
- ◆ mainly USA, UK, Spain, Singapore, Japan
- ◆ 32 institutes
- ◆ 440 scientists
- ◆ \$300M budget from NHGRI (pilot phase + production, 2003-2012)



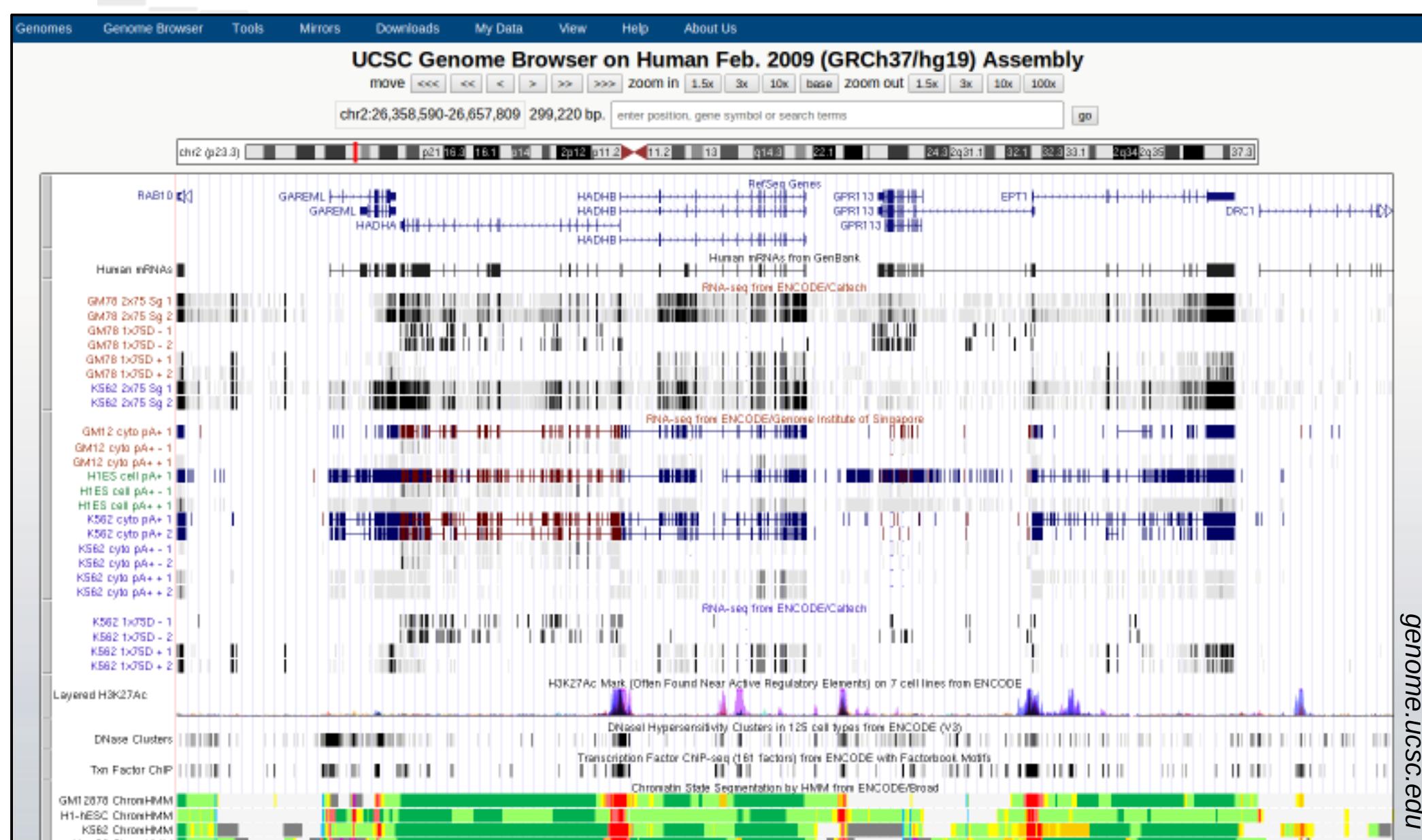
ENCODE experiments



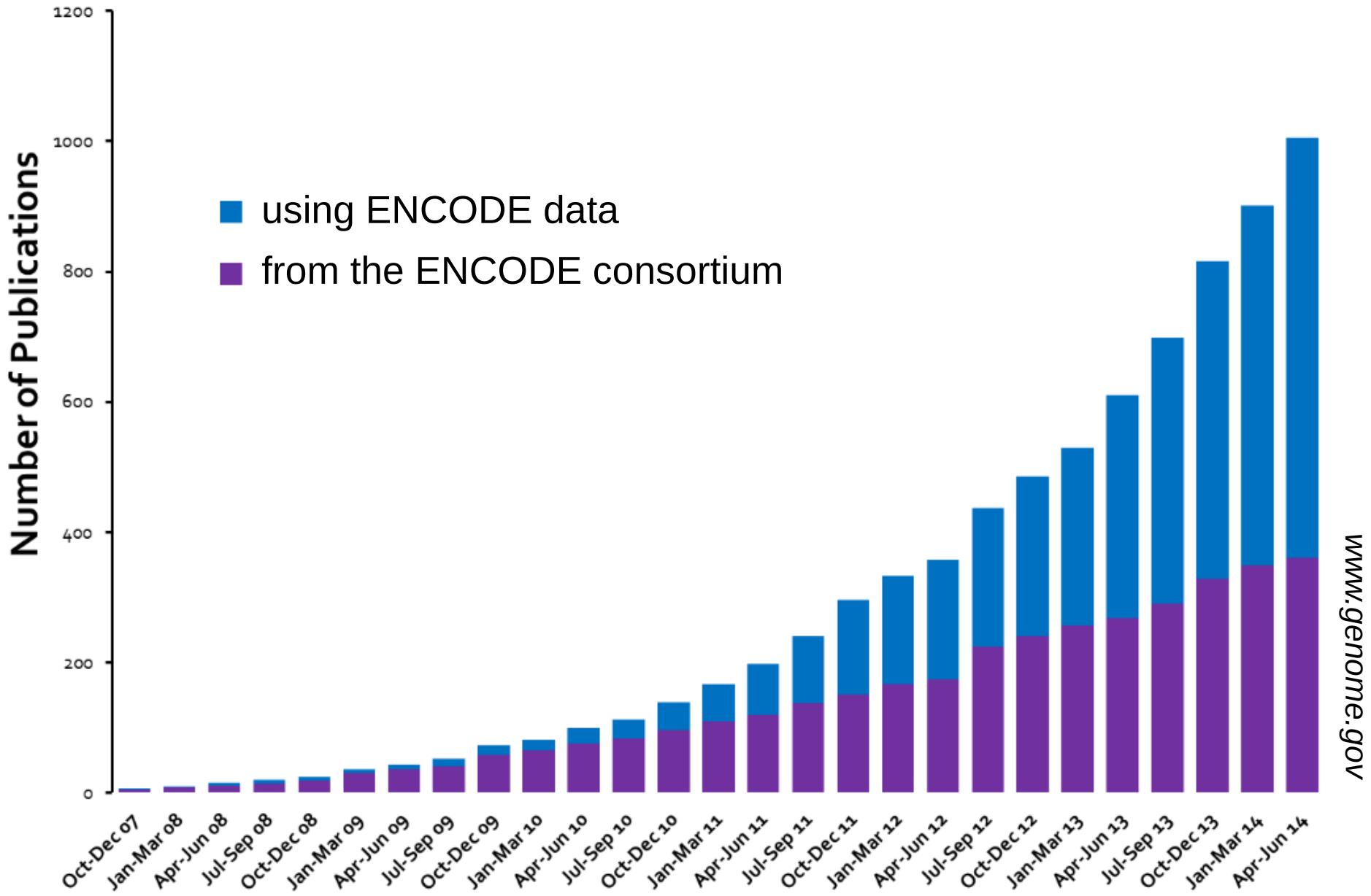
Modified from PLoS Biol 9-e1001046, 2011 & Science 306:636, 2004

Image credits: Darryl Leja (NHGRI), Ian Dunham (EBI), Michael Pazin (NHGRI)

ENCODE data



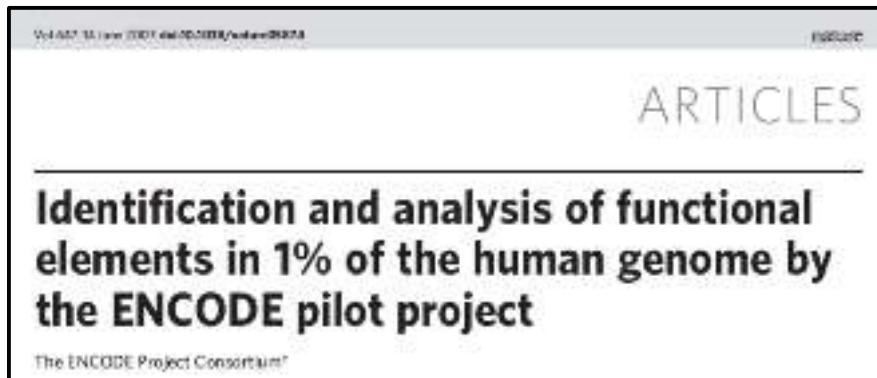
ENCODE publications



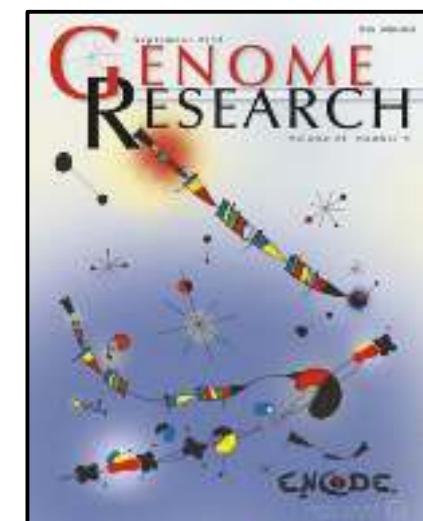
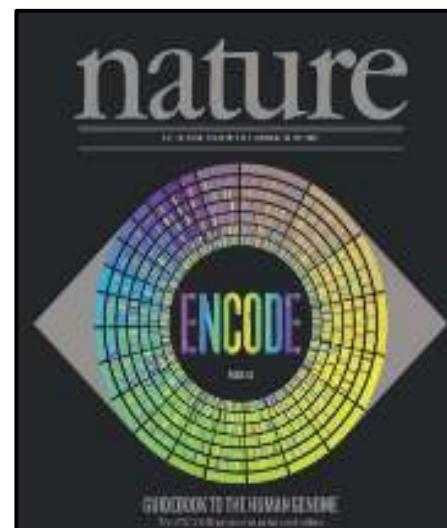
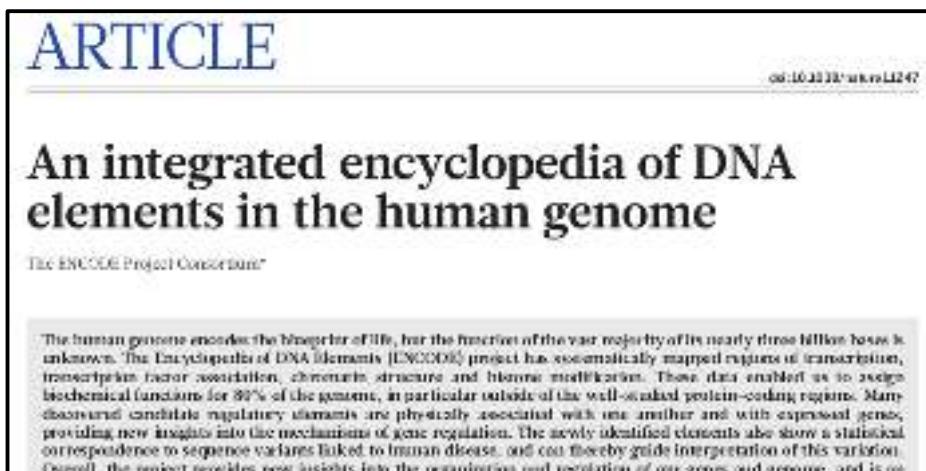
www.genome.gov

ENCODE publications

2007: pilot project



2012: 30 articles



'Junk DNA' Debunked

Studies Find Human Genomic Makeup Is Vastly Messier; New

By GAUTAM NAIK and ROBERT LEE HOTZ

Updated Sept. 5, 2012 2:01 p.m. ET

The Economist

World politics Business & finance Economics Science & technology

Human genomics

The new world of DNA

A long-term effort to catalogue all the bits of the human genome that do something has revealed a complex, interconnected web of genetic activity.

Los Angeles Times | ARTICLE

← Back to Original Article

ENCODE project sheds light on 'junk DNA'

New findings from the ENCODE project provide a blueprint for personalized medicine.

September 05, 2012 | By Rosie Mestel and Eryn Brown

DNA project interprets 'book of life'

By Elizabeth Landau, CNN

September 5, 2012 – Updated 1841 GMT (0241 HKT)

CNN

CNN

(CNN) -- Our genes play a major role in our bodies, but a lot of information about them remains mysterious.

That's why an international team of scientists is working out what the working parts of the genome are and what they mean for the human body as we know it.

The project is called the Encyclopedia of DNA Elements (ENCODE).

NEWS HEALTH

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Human genome 'more active than'

The New York Times

Science

WORLD U.S. N.Y. / REGION BUSINESS TECHNOLOGY SCIENCE HEALTH ENVIRONMENT S

Bits of Mystery DNA, Far From 'Junk,' Play Critical Role

the guardian

Winner of the Pulitzer prize

home > science

UK world sport football comment culture ecor = all sections

Genetics

Breakthrough study overturns theory of 'junk DNA' in genome

The international Encode project has found that about a fifth of the human genome regulates the 2% that makes proteins.

The Tele

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Science News Space

Night Sky Roger Highfield Dinosaurs Evolution Steve Jones Science

HOME > SCIENCE > SCIENCE NEWS

Worldwide army of scientists cracks the 'junk DNA' code

One of the biggest mysteries in genetics has been solved after an international team of hundreds of scientists uncovered the secrets of "junk DNA".



The FAANG project

FAANG: Functional Analysis of ANimal Genomes

Andersson et al, Genome Biology, 2015

*Coordinated international action to accelerate Genome to Phenome -
the Functional Annotation of Animal Genome (FAANG) project*



wikipedia.org

The FAANG project

FAANG: Functional Analysis of ANimal Genomes

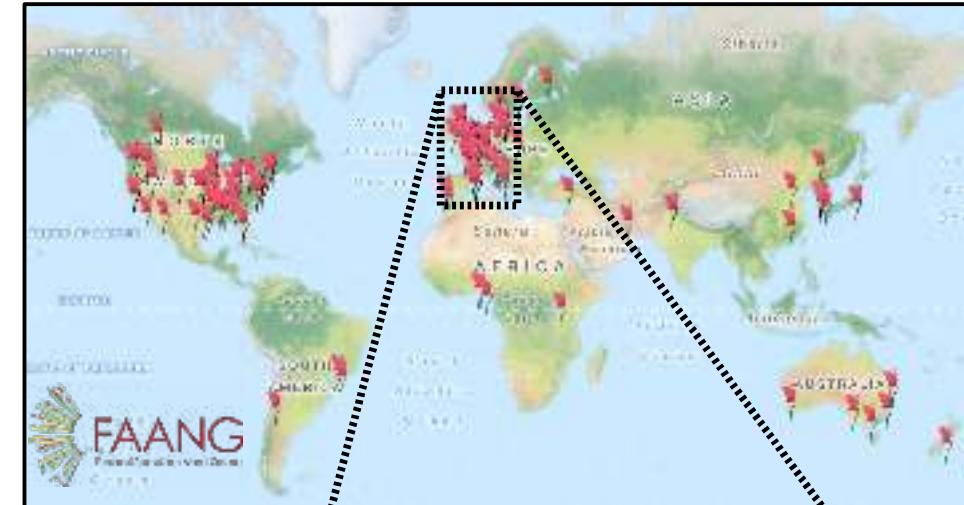
www.faang.org

=> 3 pilot projects

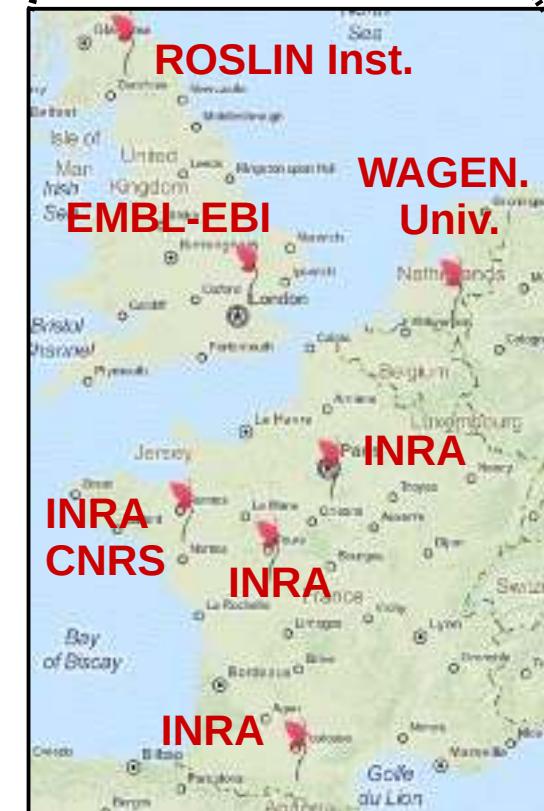
The screenshot shows the FAANG project website. At the top, there is a navigation bar with links for Home, Project description, Publications, GO-FAANG, News, Wiki, and Login. To the right of the navigation bar is a Twitter icon. The main header features the FAANG logo, which consists of a circular arrangement of colored lines radiating from a central point, followed by the text "FAANG" in large red letters and "Functional Annotation of Animal Genomes" in smaller text below it. Below the header, a tagline reads "A coordinated international action to accelerate genome to phenotype". A section titled "FAANG aims to:" lists several goals, including standardizing core assays and experimental protocols, coordinating and facilitating data sharing, establishing an infrastructure for analysis of these data, providing high quality functional annotation of animal genomes, signing up to take part in its activities, and contacting respective working committees to get involved. To the right of this section is a box titled "Working groups" containing a list of six groups: Steering Committee, Animals, Samples and Assays (ASA), Bioinformatics and Data Analysis (B&DIA), Communication (COM), and Metadata and Data Sharing (M&DS). At the bottom of the page are three world maps showing the distribution of White paper authors, FAANG Contributors, and FAANG Signatories.

FR-AgENCODE: overview

- ◆ a French pilot project of FAANG
- ◆ goal: improve functional annotation of livestock genomes
- ◆ founding: INRA, France (300KE)
- ◆ 4 INRA sites, 9 labs, 58 scientists
- ◆ 4 species: pig, chicken, cattle, goat
- ◆ primary targets:
liver & blood cells (CD4+ & CD8+)
- ◆ molecular assays:
RNA-seq, Hi-C & ATAC-seq
- ◆ duration: 2015-2017



*E. Giuffra,
INRA GABI*



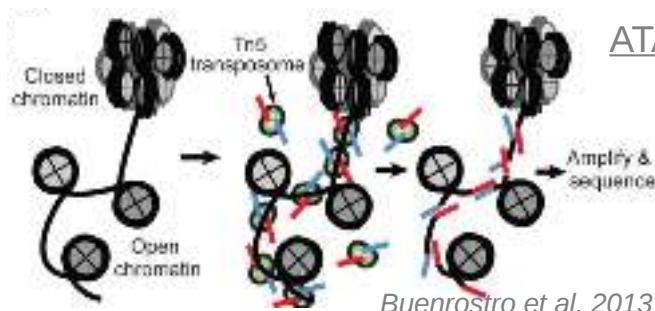
FR-AgENCODE: overview

Sampling: 40+ tissues

(liver, CD4+, CD8+, sperm, plasma, heart, lung, skin, fat, duodenum, ileum, jejunum, cerebellum, frontal lobe, olfactory bulb, trigeminal ganglia, hypothalamus, pancreas, adrenals, kidney, muscle, bone, joints, spleen, lymphatic nodes, peyer's patches, ovary, oocytes, oviduct, uterus, mammary gland, acini, testis, seminal vesicle, etc)



=> INRA CRB-Anim biorepository



ATAC-seq: chromatin accessibility

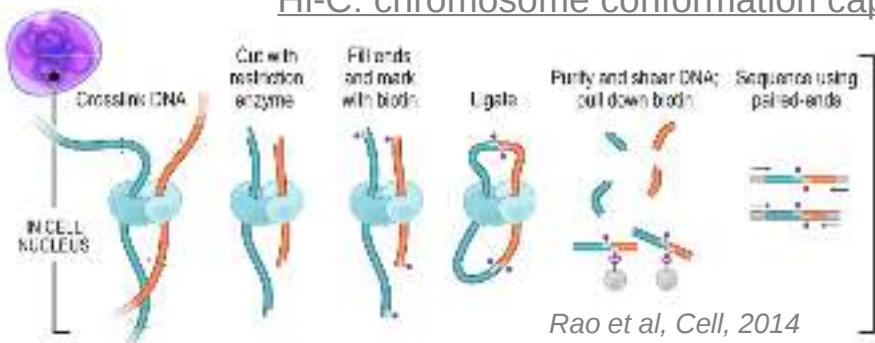
RNA-seq: long & short RNAs

Molecular assays: 3 target tissues

transcriptome & chromatin structure profiling
polyA+ RNA-seq (mRNAs & lncRNAs, 130M RP/lib)
small RNA-seq (miRNAs & <200nt RNAs, 40MR/lib)
Hi-C (130M RP/lib) & ATAC-seq (40M RP/lib)



Hi-C: chromosome conformation capture

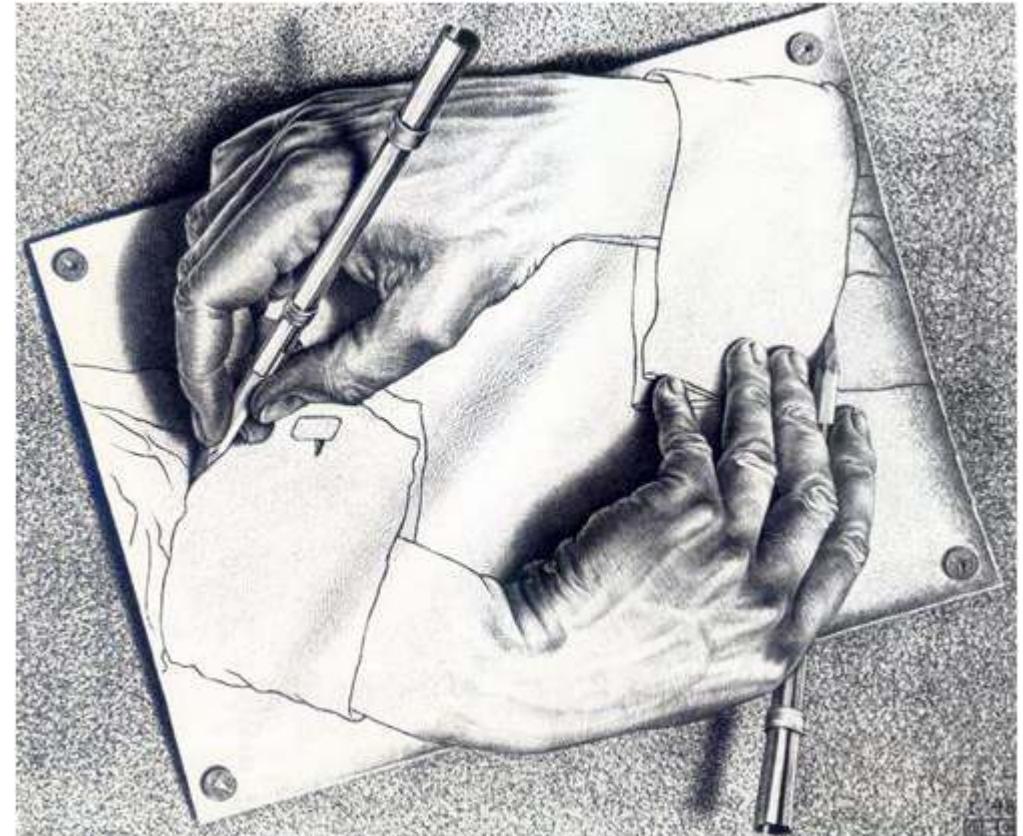


Bioinformatics data analysis

genome annotation, gene expression, lncRNAs & sRNAs annotation/prediction, chromosome interaction matrices & contact heatmaps, allele-specific expression, chimeric transcripts detection, comparative genomics, etc

Outline

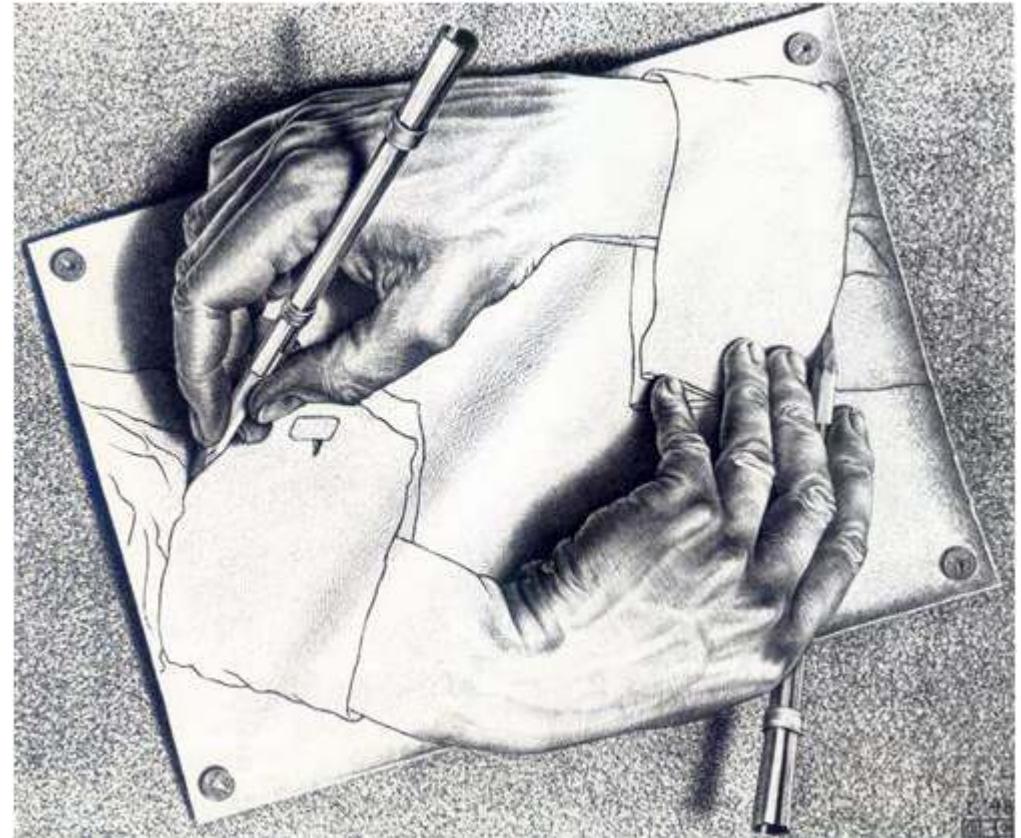
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M.C. Escher, 1948

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M.C. Escher, 1948

FR-AgENCODE: sampling

2x ♂



2x ♀

Sus scrofa
(Large White)



Gallus gallus
(White Leghorn)



Bos Taurus
(Holstein)



Capra hircus
(Alpine)

- ◆ 34 somatic tissues + 13 reproductive tissues (8 female + 5 male):
liver, sperm, CD4+, CD8+, plasma, heart, lung, skin, fat, duodenum, ileum, jejunum, cerebellum, frontal lobe, olfactory bulb, trigeminal ganglia, hypothalamus, pancreas, adrenals, kidney, muscle, bone, joints, spleen, lymphatic nodes, peyer's patches, ovary, oocytes, oviduct, uterus, mammary gland, acini, testis, seminal vesicle, etc...
- ◆ total: 2,000 to 6,000 samples

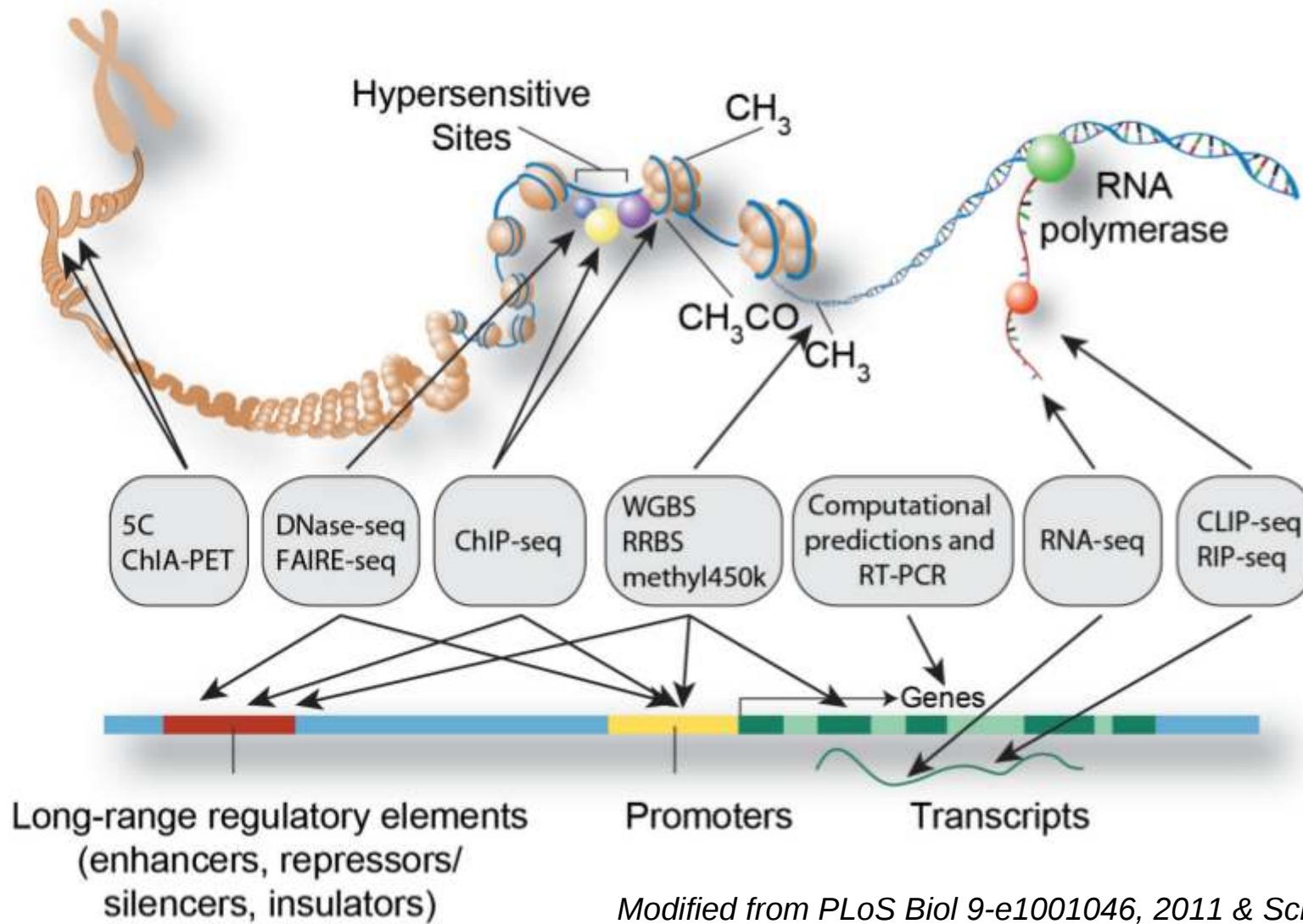


M. Tixier-
Boichard,
INRA GABI S. Fabre
INRA
GenPhySE



INRA CRB-Anim biorepository

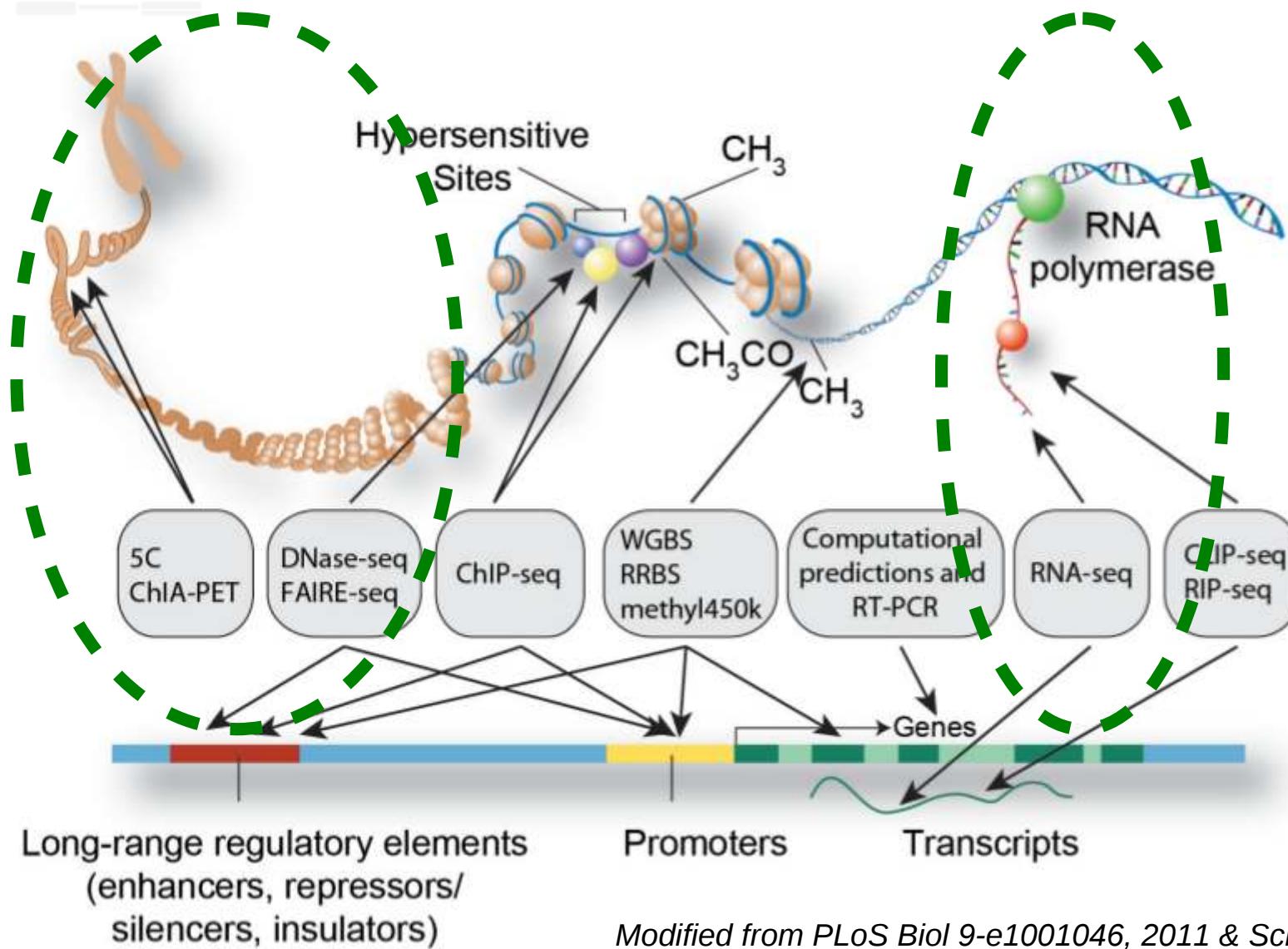
FR-AgENCODE: molecular assays



Modified from PLoS Biol 9-e1001046, 2011 & Science 306:636, 2004

Image credits: Darryl Leja (NHGRI), Ian Dunham (EBI), Michael Pazin (NHGRI)

FR-AgENCODE: molecular assays



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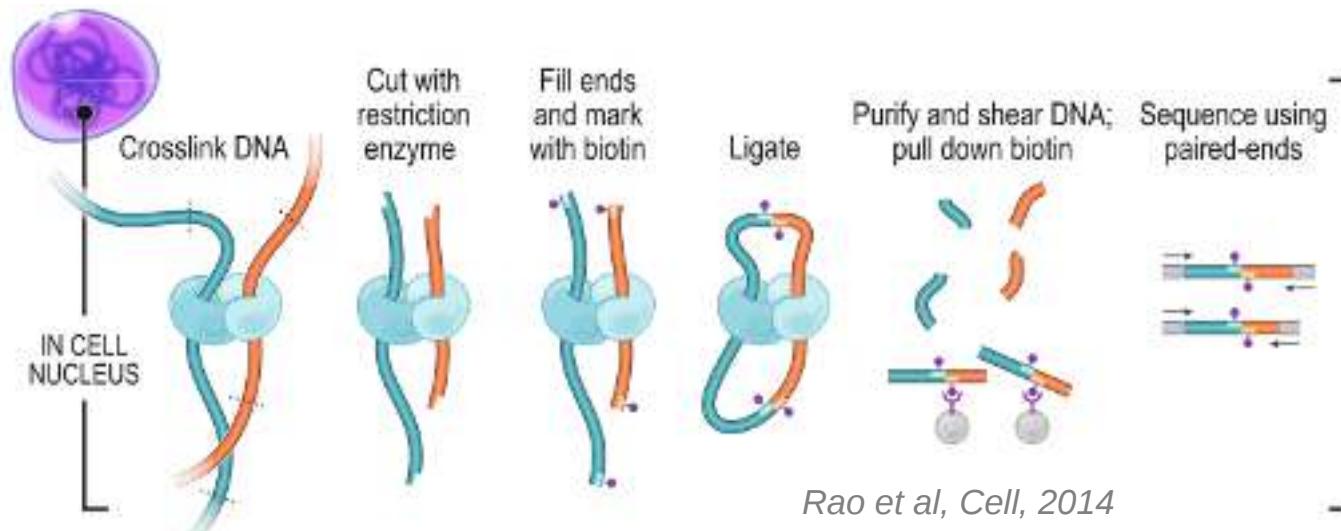
Image credits: Darryl Leja (NHGRI), Ian Dunham (EBI), Michael Pazin (NHGRI)

FR-AgENCODE: molecular assays

- ◆ RNA-seq
 - directional protocol, Illumina Hi-Seq3000
 - ◆ polyA+ RNAs (mRNAs + lncRNAs)
100M+ pairs (2x150bp) per sample (2/lane)
 - ◆ sRNAs (miRNAs and others)
35M reads (1x50bp) per sample (3/lane)

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- ◆ Hi-C: chromosome conformation capture
(Lieberman-Aiden et al, Science, 2009, Rao et al, Cell, 2014)

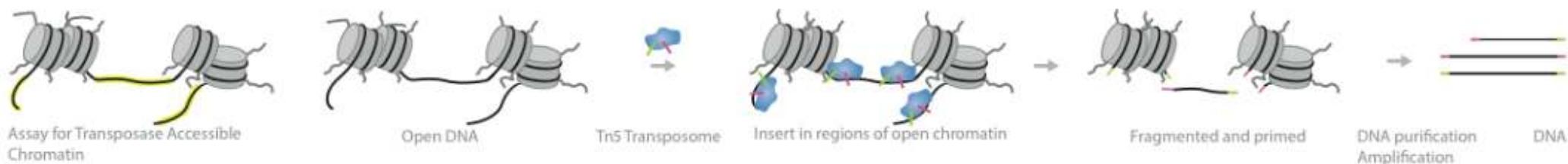


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100M+ pairs (2x150bp) per sample (2/lane)
 - ◆ sRNAs (miRNAs and others)
35M reads (1x50bp) per sample (3/lane)
 - ◆ Hi-C: chromosome conformation capture
 - ◆ 16 samples (4 replicates, 4 species)
 - ◆ 130M read pairs per sample (2 lib/lane)

FR-AgENCODE: molecular assays

- ◆ RNA-seq
 - directional protocol, Illumina Hi-Seq3000
 - ◆ polyA+ RNAs (mRNAs + lncRNAs)
100M+ pairs (2x150bp) per sample (2/lane)
 - ◆ sRNAs (miRNAs and others)
35M reads (1x50bp) per sample (3/lane)
 - ◆ Hi-C: chromosome conformation capture
 - ◆ 16 samples (4 replicates, 4 species)
 - ◆ 130M read pairs per sample (2 lib/lane)
 - ◆ ATAC-seq: chromatin accessibility



Adapted from: www.illumina.com/techniques

FR-AgENCODE: molecular assays

- ◆ RNA-seq
 - directional protocol, Illumina Hi-Seq3000
 - ◆ polyA+ RNAs (mRNAs + lncRNAs)
100M+ pairs (2x150bp) per sample (2/lane)
 - ◆ sRNAs (miRNAs and others)
35M reads (1x50bp) per sample (3/lane)
 - ◆ Hi-C: chromosome conformation capture
 - ◆ 16 samples (4 replicates, 4 species)
 - ◆ 130M read pairs per sample (2 lib/lane)
 - ◆ ATAC-seq: chromatin accessibility



D. Esquerre
INRA GetPlage

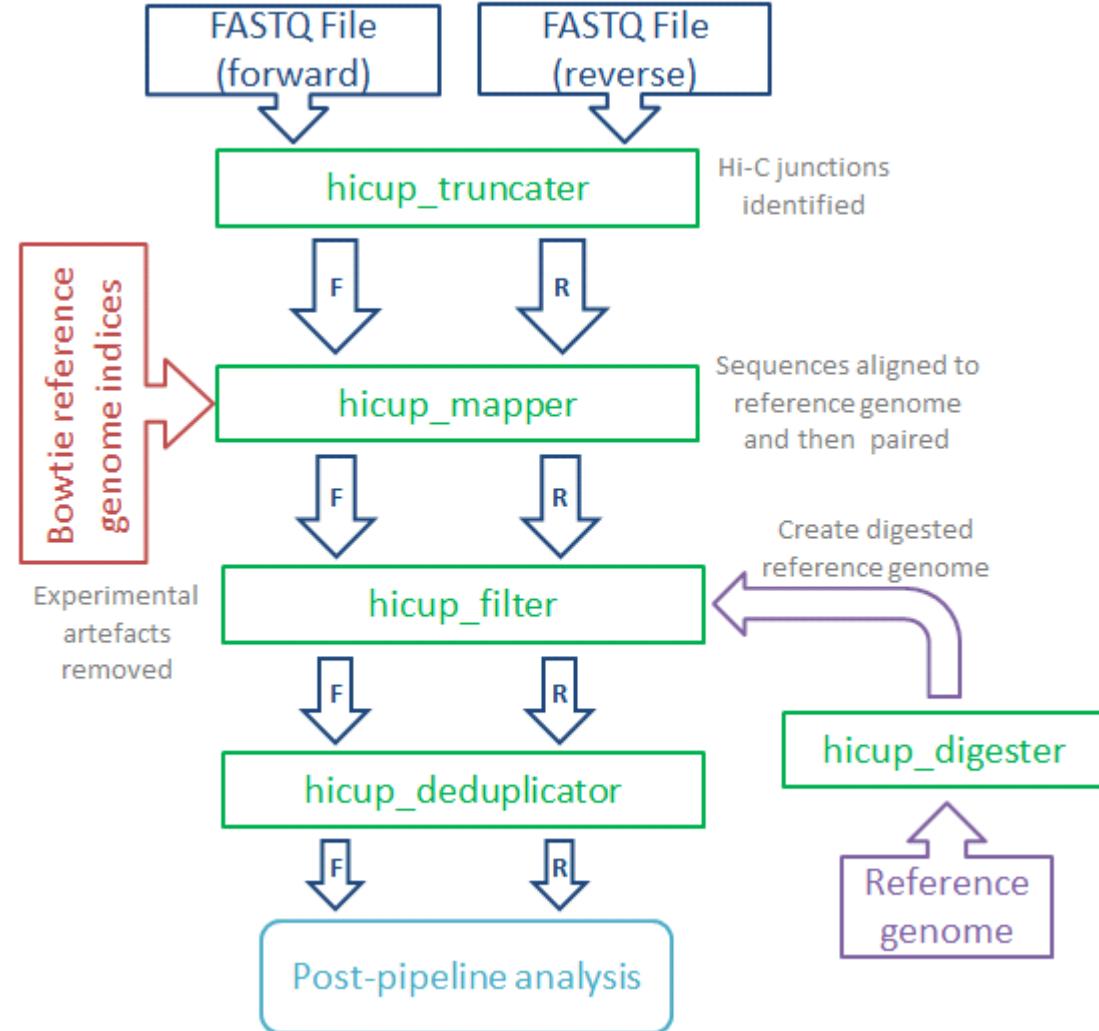
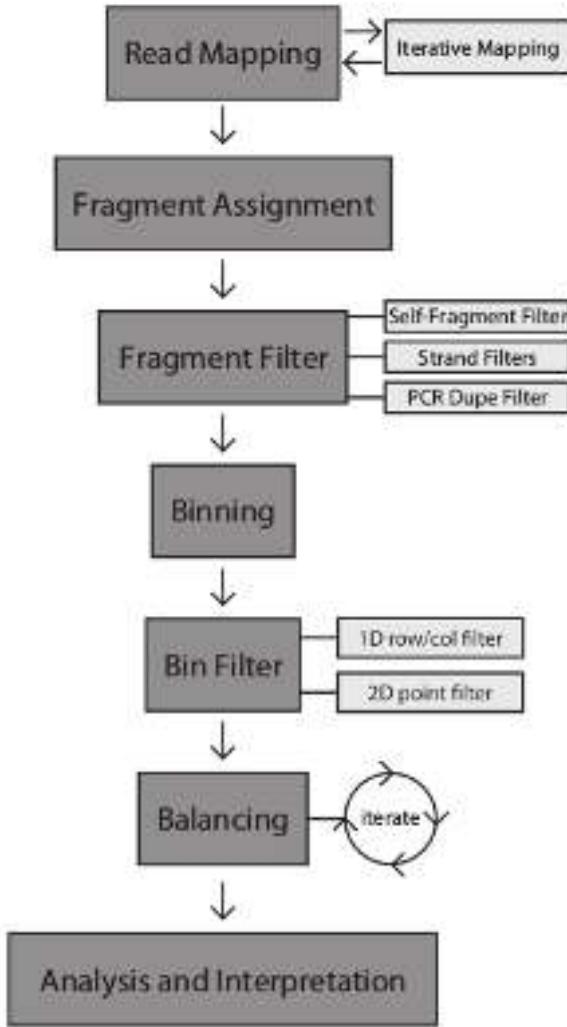


H. Acloque,
INRA GenPhySE

→ Comparative analysis of genome topology and expression

Hi-C data analysis: overview

Hi-C Processing Flow Chart



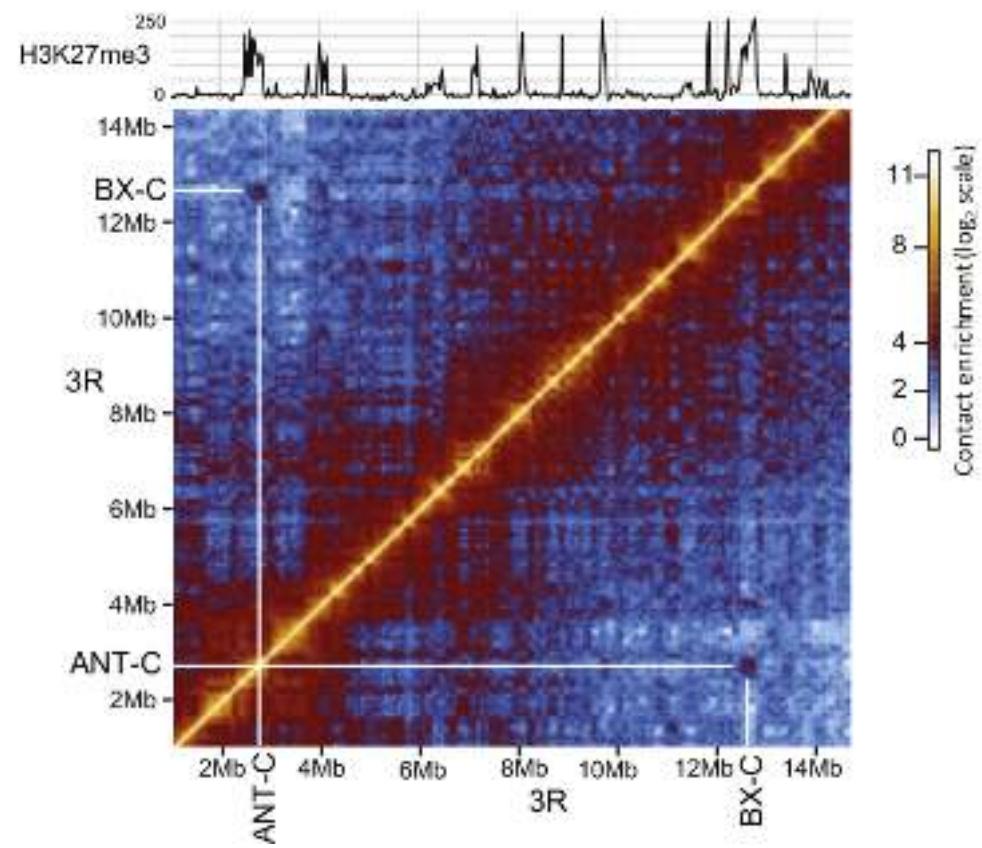
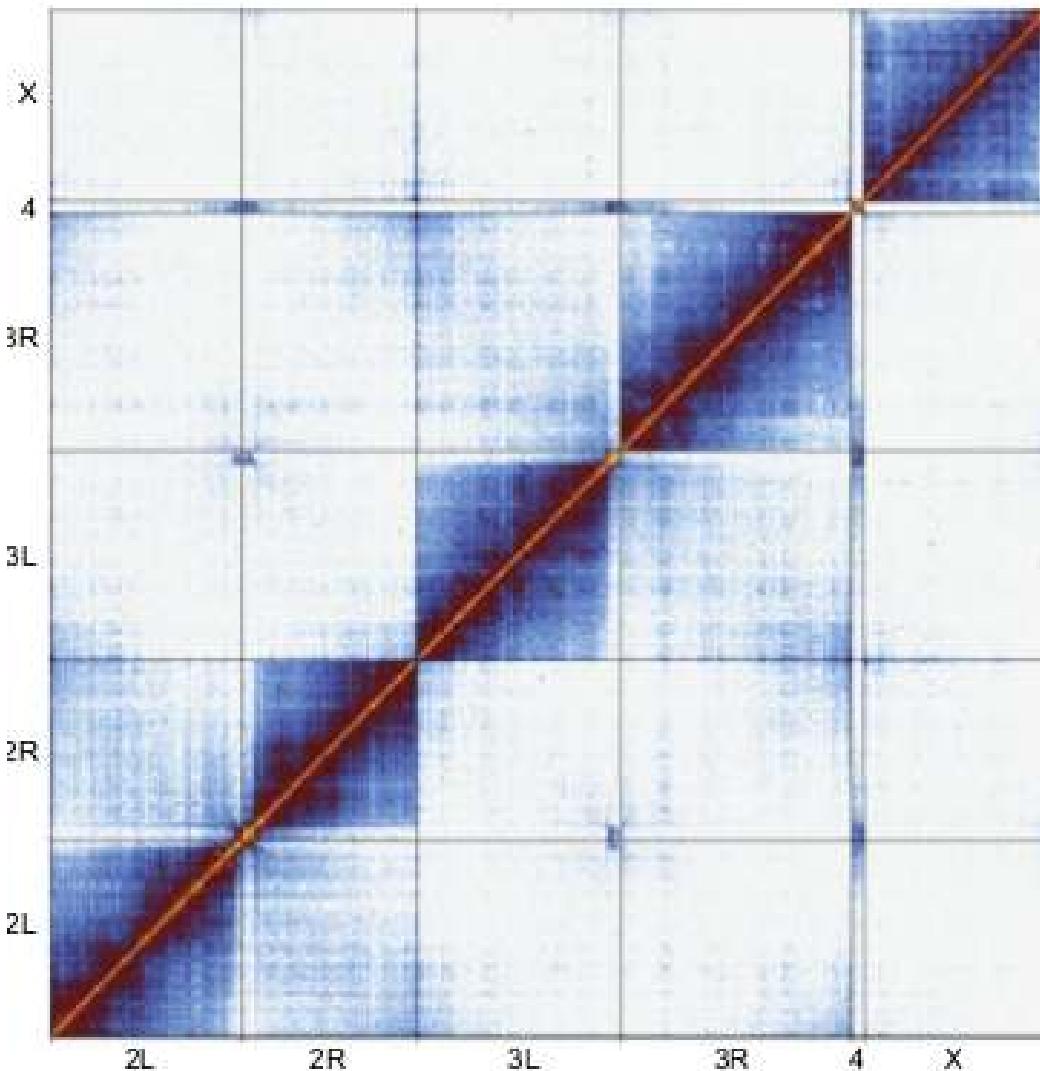
Lajoie et al, 2015

HiCUP, www.bioinformatics.babraham.ac.uk/

Hi-C data analysis: overview

- ◆ clean and trim the reads
- ◆ map the reads on the genomic reference
- ◆ filter bogus configurations
- ◆ count the reads per genomic bin => contact matrix
- ◆ normalize the matrix
- ◆ identify topological domains, cis- and trans- interactions
- ◆ comparative/integrative analysis

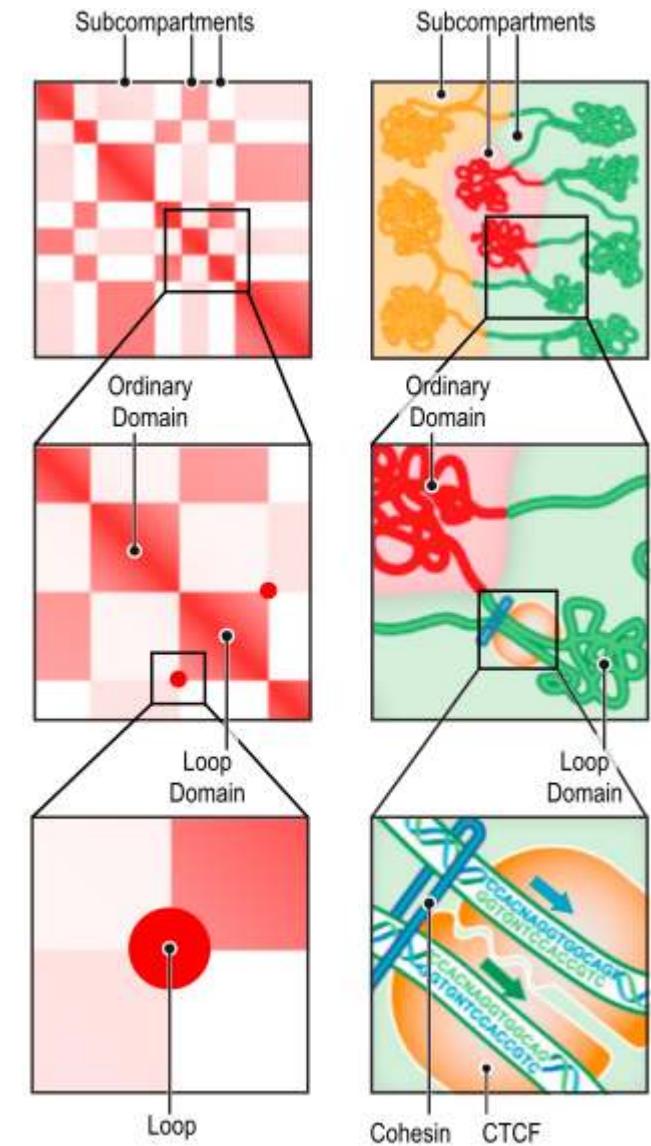
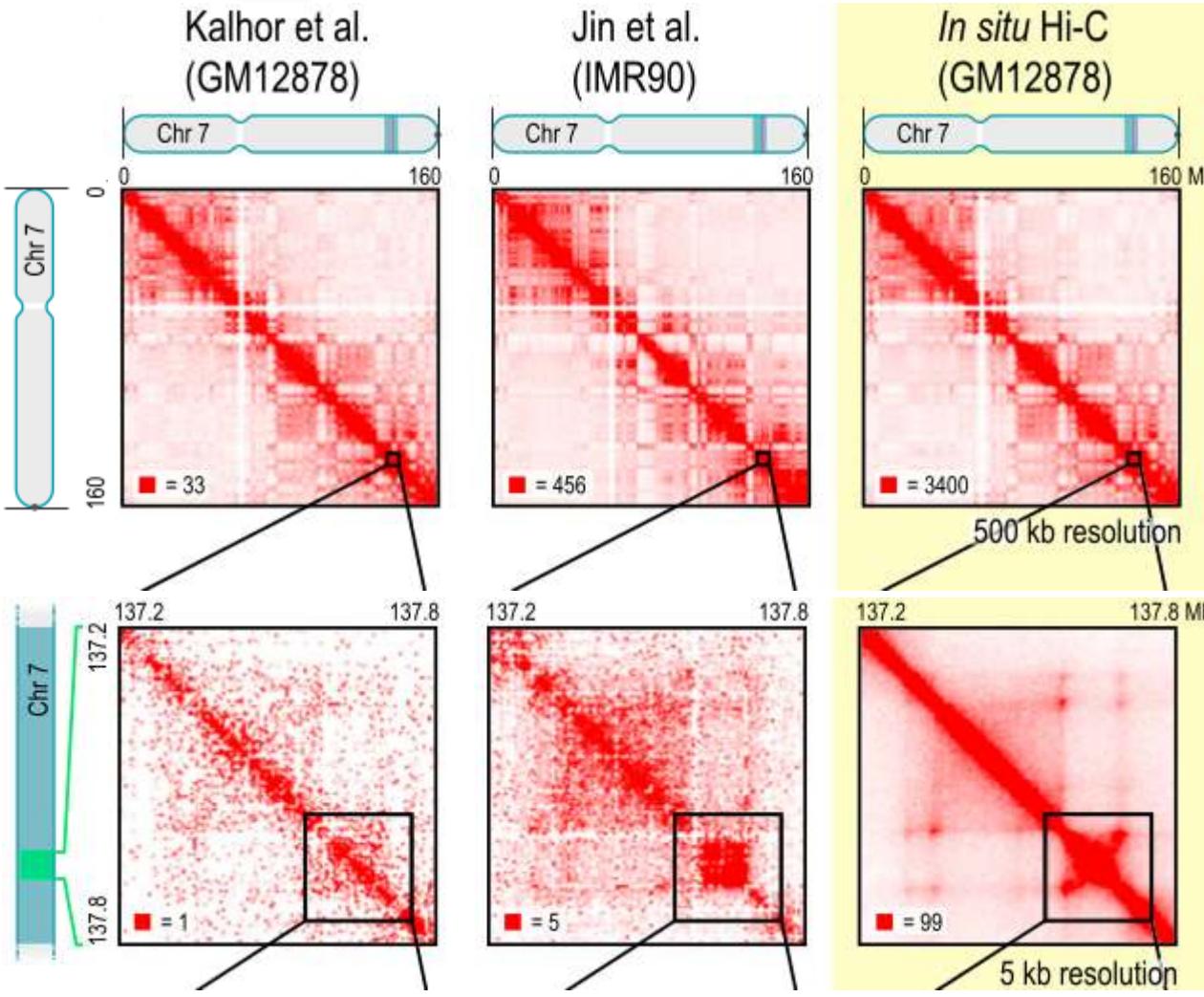
Hi-C data analysis: the contact matrix



Sexton et al 2012

Hi-C data analysis: the contact matrix

Rao et al, Cell, 2014



Hi-C data analysis: matrix normalization

Number of reads per bin (coverage) depends on:

- ◆ GC%
- ◆ density of restriction sites
- ◆ repeats and “mappability”
- ◆ overall depth of coverage
- ◆ Others?

=> “Parametric” vs. “non-parametric” normalization

Hi-C data analysis: matrix normalization

A FAST ALGORITHM FOR MATRIX BALANCING

PHILIP A. KNIGHT* AND DANIEL RUIZ†

Abstract. As long as a square nonnegative matrix A contains sufficient nonzero elements, then the matrix can be balanced, that is we can find a diagonal scaling of A that is doubly stochastic. A number of algorithms have been proposed to achieve the balancing, the most well known of these being Sinkhorn-Knopp. In this paper we derive new algorithms based on inner-outer iteration schemes. We show that Sinkhorn-Knopp belongs to this family, but other members can converge much more quickly. In particular, we show that while stationary iterative methods offer little or no improvement in many cases, a scheme using a preconditioned conjugate gradient method as the inner iteration can give quadratic convergence at low cost.

Key words. Matrix balancing, Sinkhorn-Knopp algorithm, doubly stochastic matrix, conjugate gradient iteration.

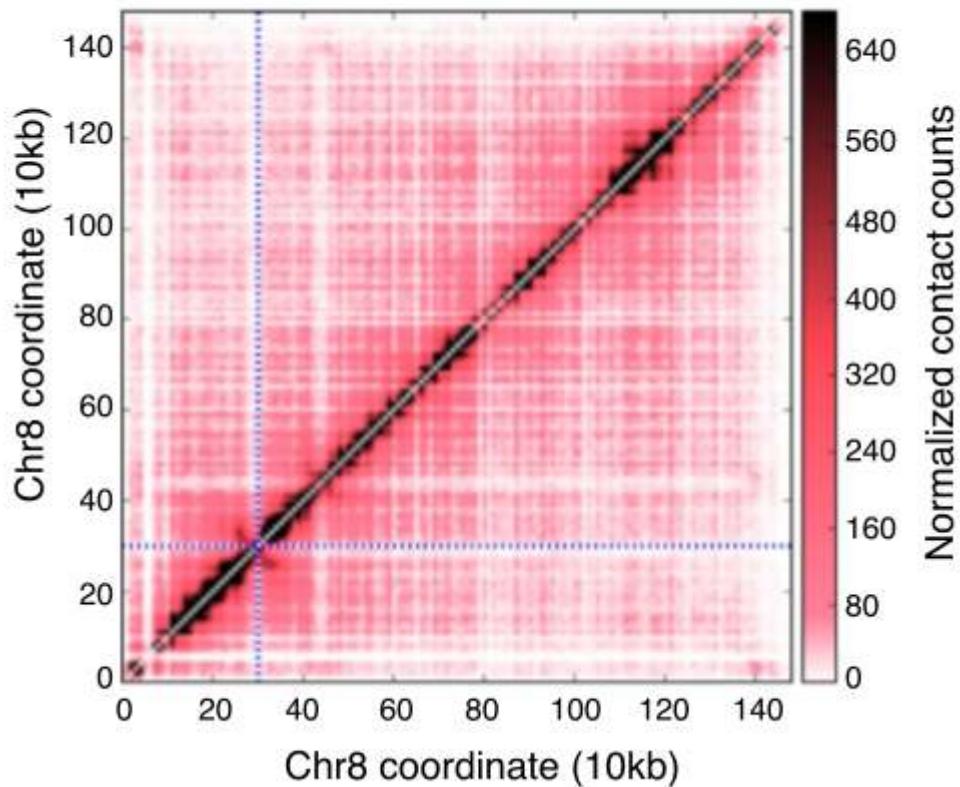
AMS subject classifications. 15A48, 15A51, 65F10, 65H10.

1. Introduction. For at least 70 years, scientists in a wide variety of disciplines have attempted to transform square nonnegative matrices into doubly stochastic form by applying diagonal scalings. That is, given $A \in \mathbb{R}^{n \times n}$, $A \geq 0$, find diagonal matrices D_1 and D_2 so that $P = D_1AD_2$ is doubly stochastic. Motivations for achieving this balance include interpreting economic data [1], preconditioning sparse matrices [16], understanding traffic circulation [14], assigning seats fairly after elections [3], matching protein samples [4] and ordering nodes in a graph [12]. In all of these applications, one of the main methods considered is SK¹. This is an iteration process that attempts to find D_1 and D_2 by alternating in normalizing columns

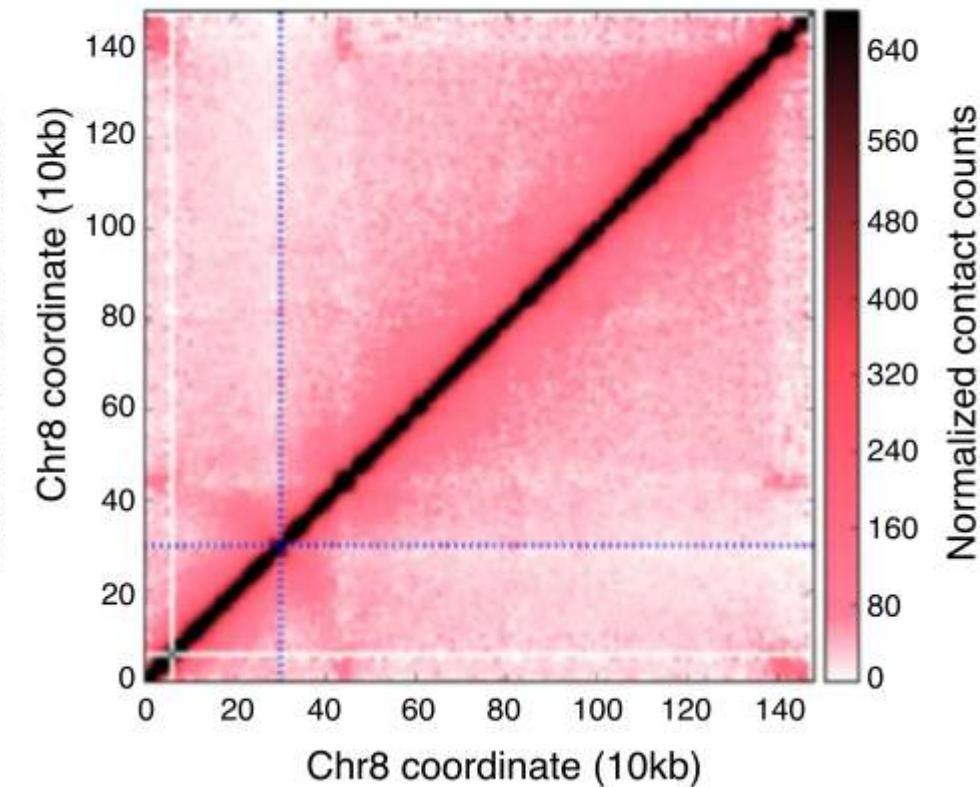
Knight & Ruiz, IMA J. Numer. Anal., 2013

Hi-C data analysis: matrix normalization

(a) Raw contact map

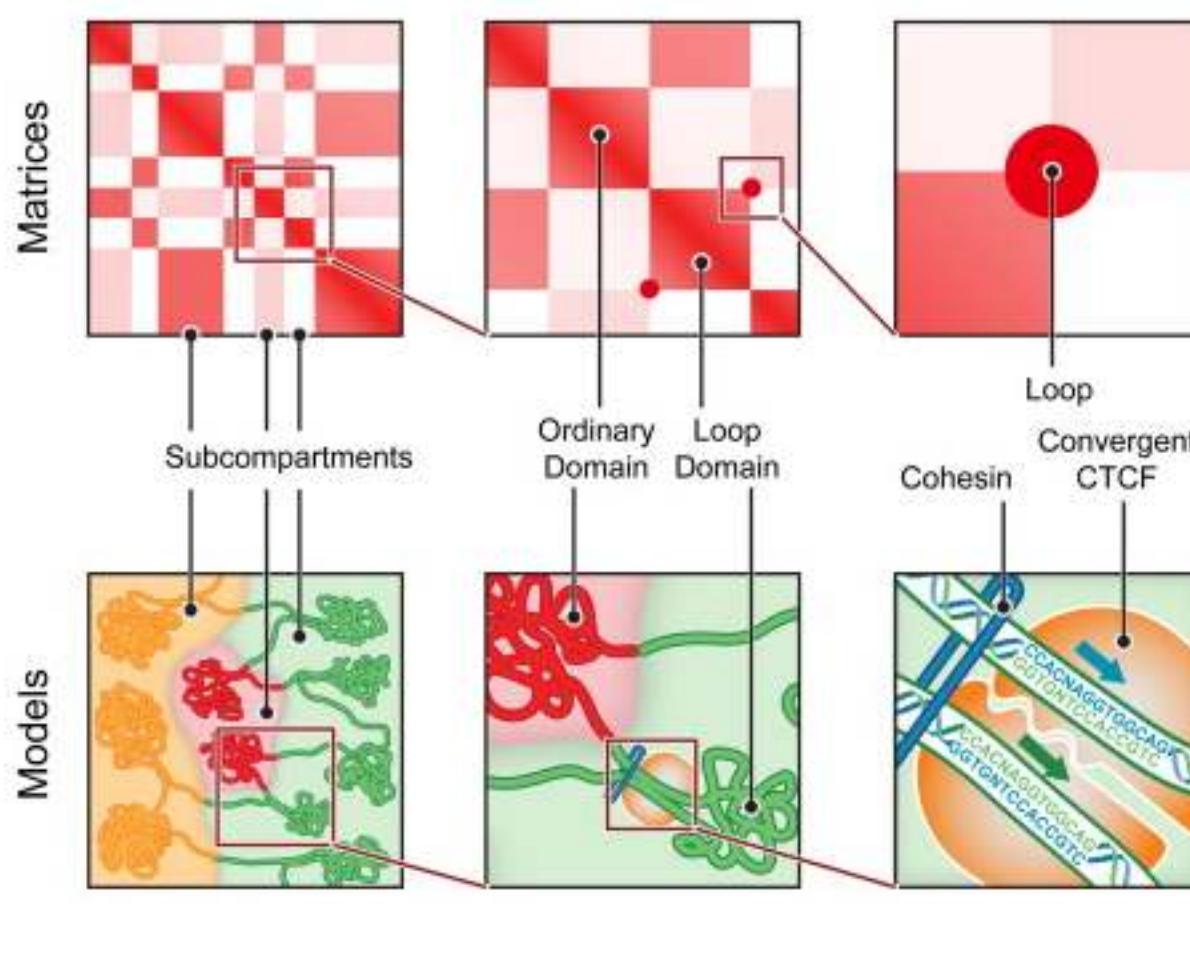


(b) Normalized contact map



Ay & Noble, *Genome Biology*, 2015

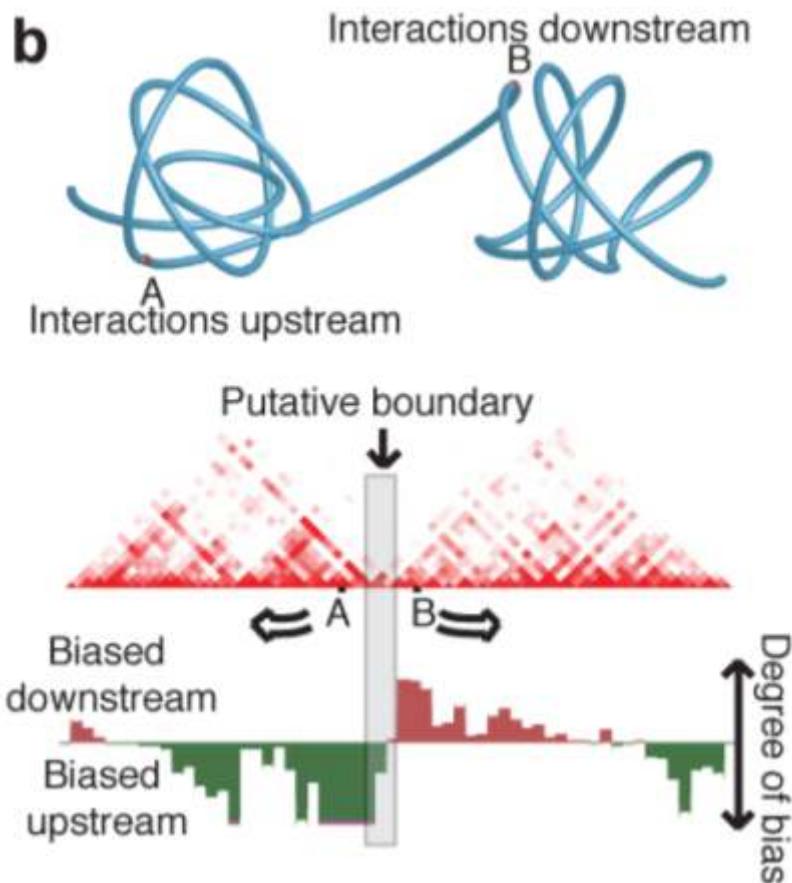
Hi-C data analysis: finding topologically associated domains (TADs)



Rao et al, Cell, 2014

- ◆ methods: clustering, 2D-segmentation, etc

Hi-C data analysis: finding topologically associated domains (TADs)



Dixon et al., Nature, 2012

NIH Public Access
Author Manuscript

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Topological Domains in Mammalian Genomes Identified by Analysis of Chromatin Interactions

Jesse R. Dixon^{1,2,4}, Siddarth Beliveau^{1,5}, Feng Yue¹, Audrey Kim¹, Yan Li¹, Yin Shen¹, Ming Hu⁶, Jun S. Liu⁶, and Bing Ren^{1,2,7}

¹Ludwig Institute for Cancer Research

²University of California, San Diego School of Medicine, Department of Cellular and Molecular Medicine, Institute of Genomic Medicine, 9500 Gilman Drive, La Jolla, CA 92093

³Medical Scientist Training Program, University of California, San Diego, La Jolla CA 92093

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⁶Department of Statistics, Harvard University, 1 Oxford Street, Cambridge, MA 02138

Abstract

The spatial organization of the genome is intimately linked to its biological function, yet our understanding of higher order genomic structure is coarse, fragmented and incomplete. In the nucleus of eukaryotic cells, interphase chromosomes occupy distinct chromosome territories (CT),

Directionality Index

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

DI HMM => TADs

Hi-C data analysis: finding topologically associated domains (TADs)

Identification of hierarchical chromatin domains

Caleb Weinreb¹, and Benjamin J. Raphael^{1,2*}

¹Center for Computational Molecular Biology, Brown University, Providence, RI

²Department of Computer Science, Brown University, Providence, RI

Associate Editor: Prof. Gunnar Rätsch

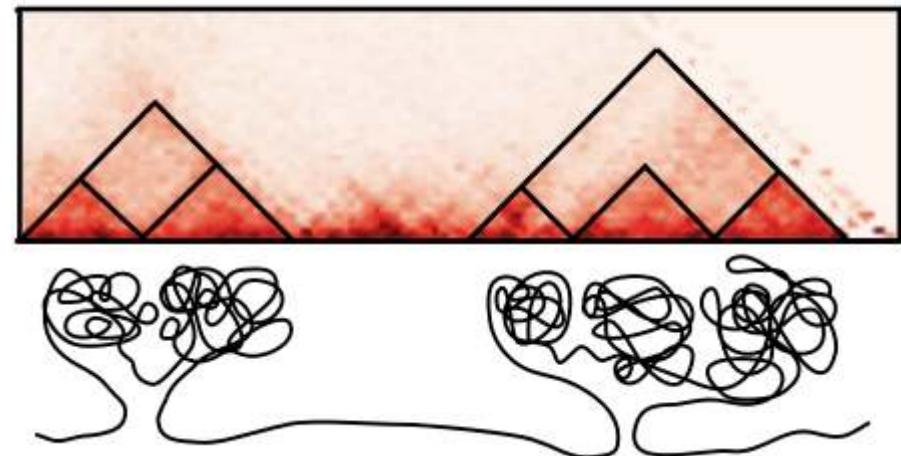
ABSTRACT

Motivation: The 3D structure of the genome is an important regulator of many cellular processes including differentiation and gene regulation. Recently, technologies such as Hi-C that combine proximity ligation with high-throughput sequencing have revealed domains of self-interacting chromatin, called topologically associating domains (TADs), in many organisms. Current methods for identifying TADs using Hi-C data assume that TADs are non-overlapping, despite evidence for a nested structure in which TADs and sub-TADs form a complex hierarchy.

Results: We introduce a model for hierarchical decomposition of contact frequency to infer a hierarchy of nested TADs. This model is based on empirical distributions of contact frequencies within TADs, where positions that are less active have a greater enrichment of contacts than

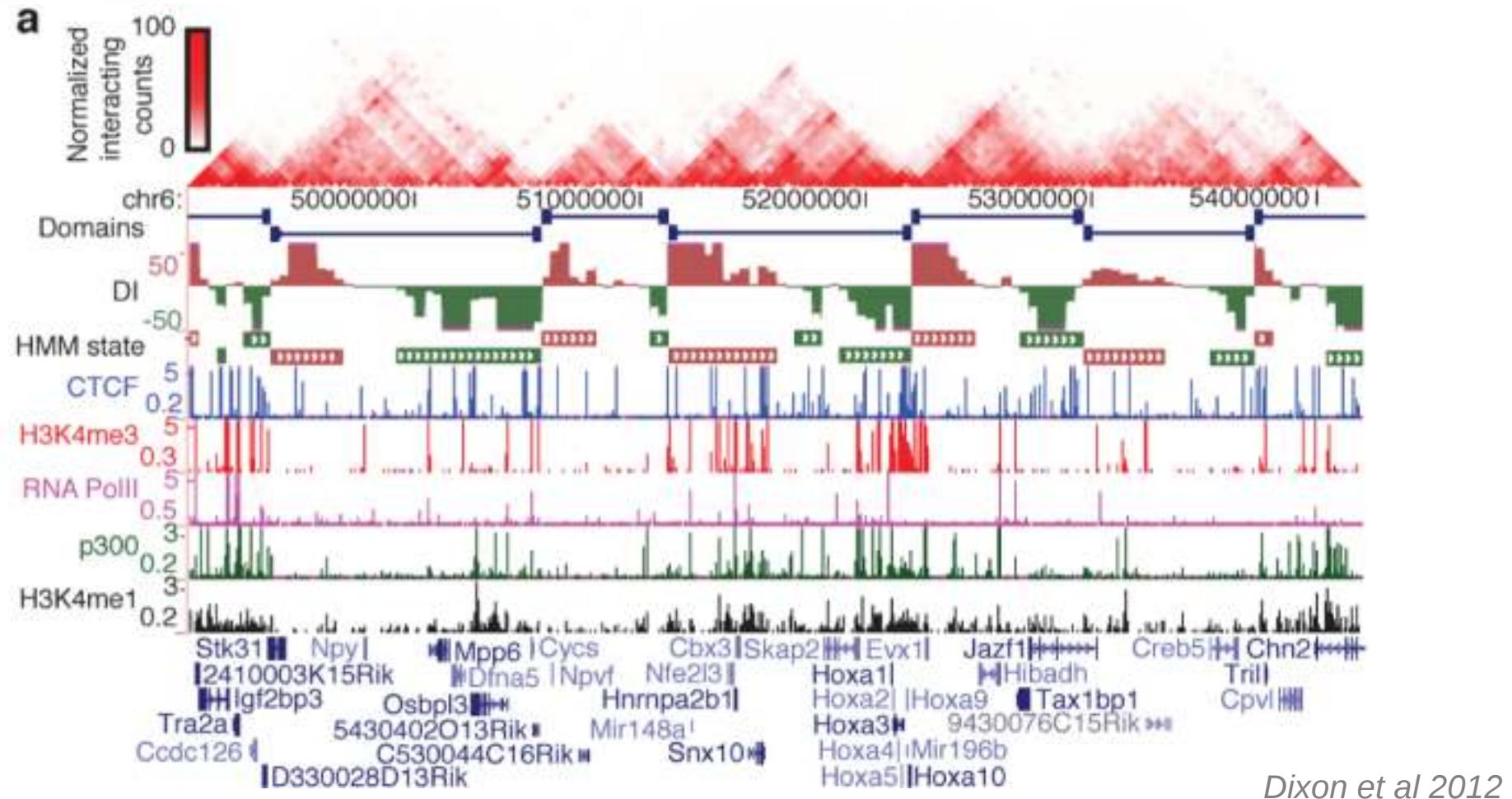
resulting in a contact matrix A , where A_{ij} is the number of contacts between bins i and j , normalized for experimental bias. Several methods have been developed for the identification of TADs from Hi-C data. These methods may be roughly classified into two categories: (1) methods that define a 1D statistic from the contact matrix A_{ij} ; (2) methods that exploit the 2D structure of the contact matrix.

Duan et al. (2012) compute a 1D “transitivity index” (DI) from the contact matrix. This index defines whether contacts have an upstream bias, downstream bias or no bias. Next, they use a hidden Markov model (HMM) to partition the genome into regions defined by changes in the transitivity index. Each transition into downstream bias marks the start of a domain and the next transition out of upstream bias marks its end. Saito et al. (2014) introduce a 1D



Weireb & Raphael, *Bioinformatics*, 2015

Hi-C data analysis: integrative analysis



Hi-C data analysis: FR-AgENCODE pipeline

Data analysis

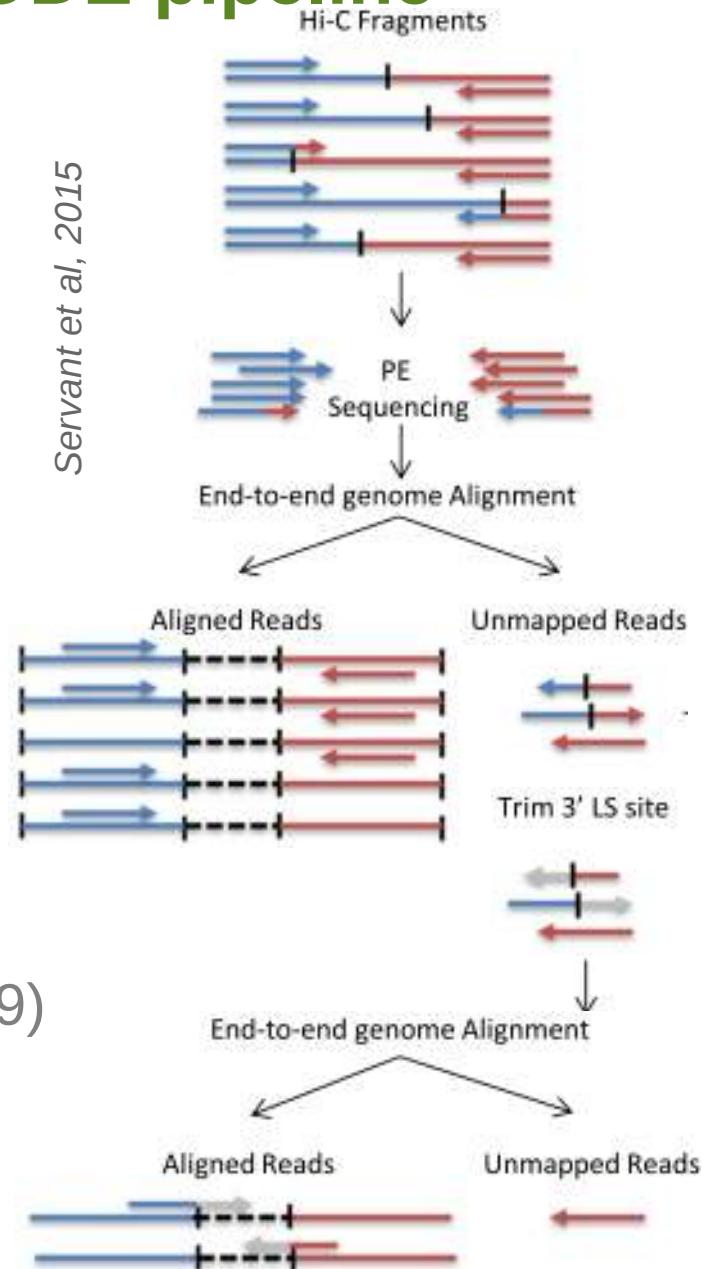
Pipeline

- ▶ Trim reads (ligation site)
- ▶ Map on reference genome
- ▶ Discard inconsistent pairs
- ▶ Count reads in pairs of genomic bins & generate contact matrix
- ▶ Normalize contact matrix (non parametric, matrix balancing)
- ▶ Identify Topologically Associated Domains, *cis* and *trans* interactions

Software

- ▶ HiC-Pro pipeline (Servant et al 2015)
- ▶ Bowtie2 mapping (Langmead et al, 2009)
- ▶ ICE normalization (Imakaev et al, 2012)
- ▶ HiTC display (Servant et al, 2012)
- ▶ HiFive pipeline (Sauria et al, 2015)

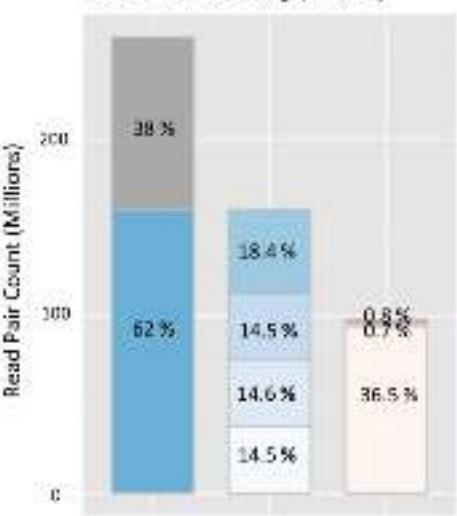
Servant et al, 2015



FR-AgENCODE Hi-C preliminary results

Read pairs status after mapping

Read Pair Filtering (IMR90)

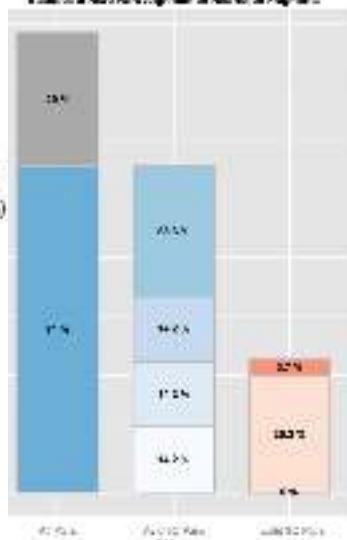


Dixon et al data

(human, from Servant et al 2015)

**62%
valid pairs**

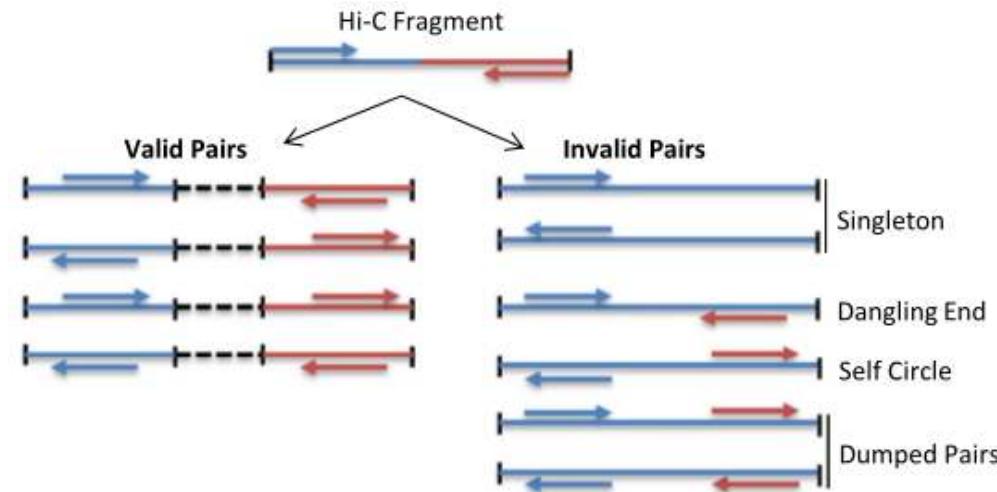
Read Pair Filtering (IMR90)



Rao et al data

(mouse, CH12 cells)

**71%
valid pairs**

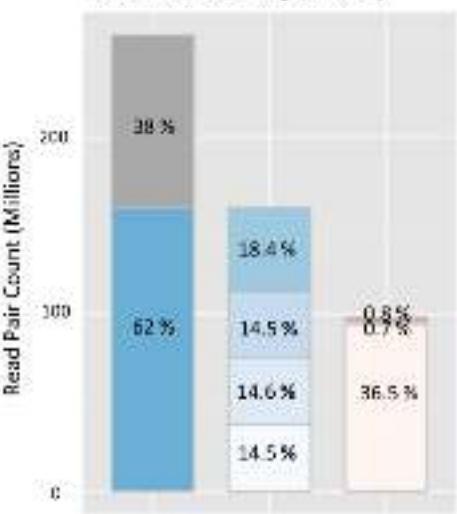


Servant et al 2015

FR-AgENCODE Hi-C preliminary results

Read pairs status after mapping

Read Pair Filtering (IMR90)

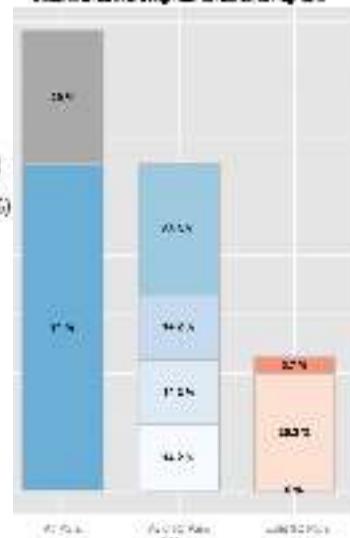


Dixon et al data

(human, from Servant et al 2015)

62%
valid pairs

Statistics of Read Pairs Alignment Post Filtered (IMR90)

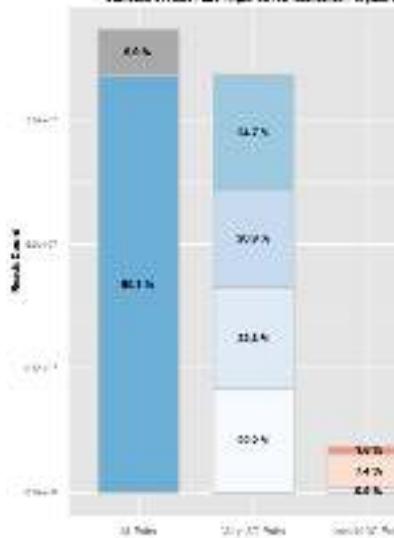


Rao et al data

(mouse, CH12 cells)

71%
valid pairs

Statistics of Read Pairs Alignment Post Filtered (IMR90)



FR-AgENCODE data

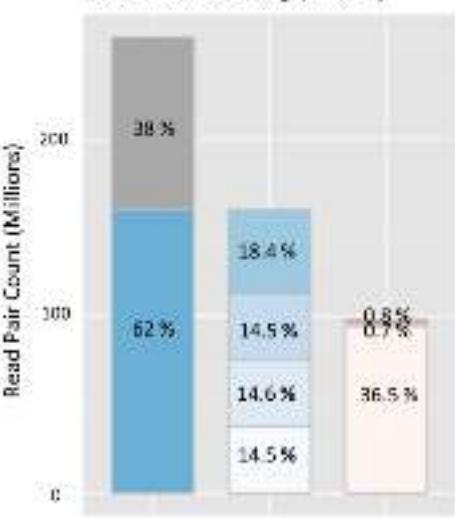
(mouse, STO cells)

90%
valid pairs

FR-AgENCODE Hi-C preliminary results

Read pairs status after mapping

Read Pair Filtering (IMR90)

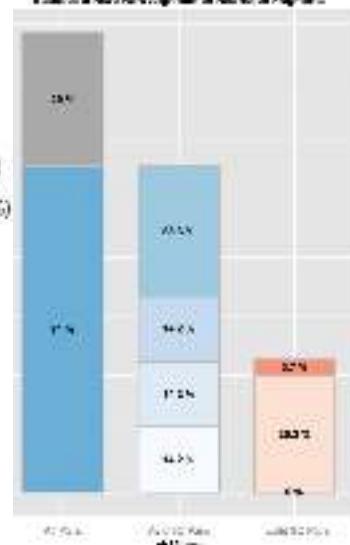


Dixon et al data

(human, from Servant et al 2015)

**62%
valid pairs**

Statistics of Read Pairs Alignment Rate after Filtering

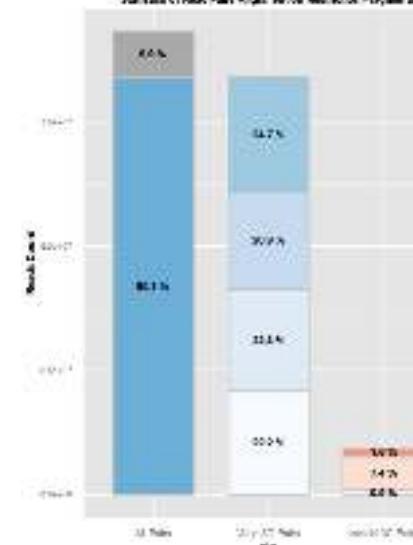


Rao et al data

(mouse, CH12 cells)

**71%
valid pairs**

Statistics of Read Pairs Alignment Rate after Filtering

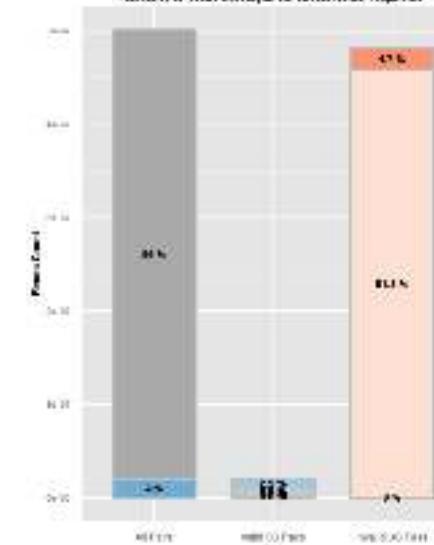


FR-AgENCODE data

(mouse, STO cells)

**90%
valid pairs**

Statistics of Read Pairs Alignment Rate after Filtering



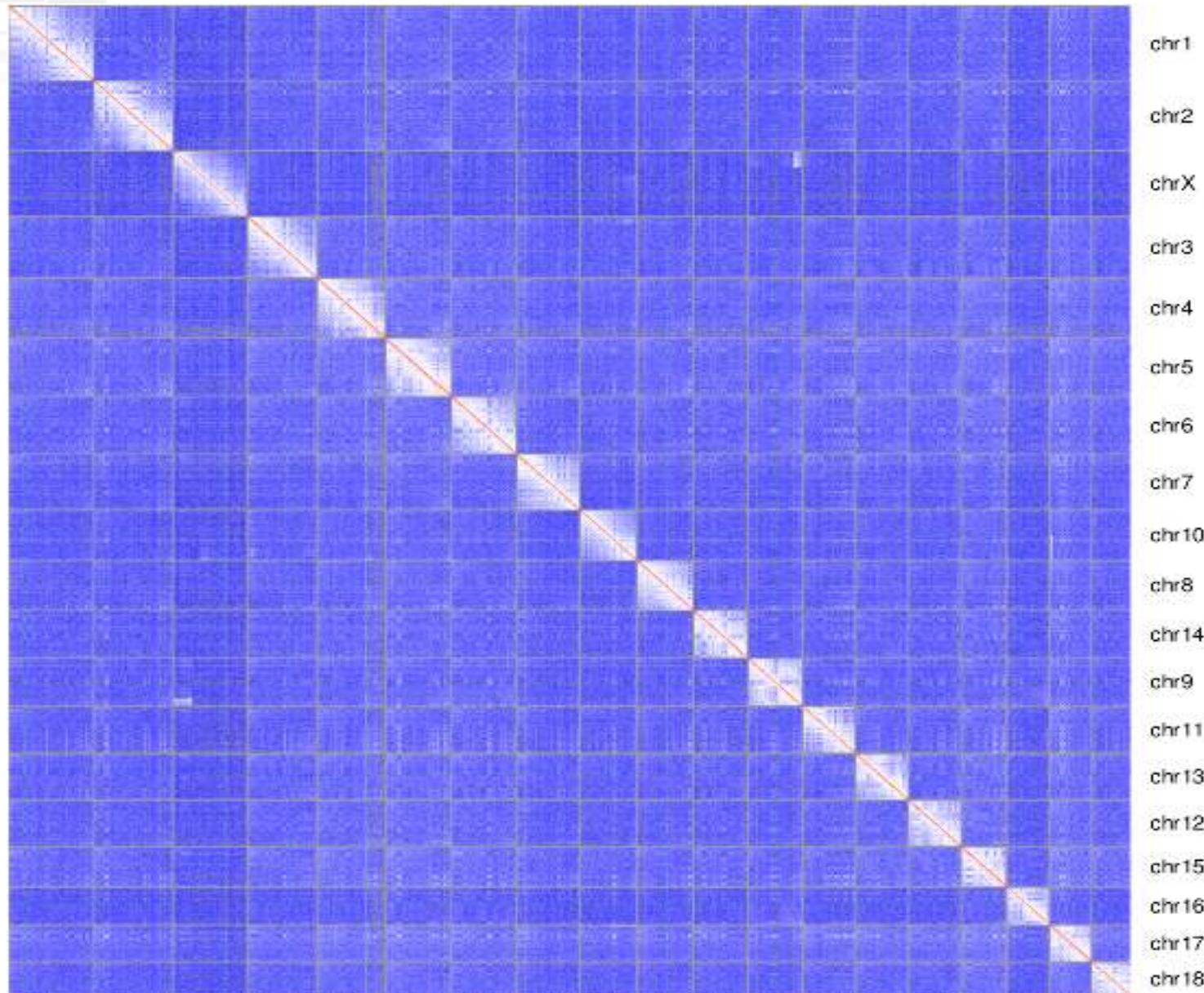
FR-AgENCODE data

(pig, hepatocytes)

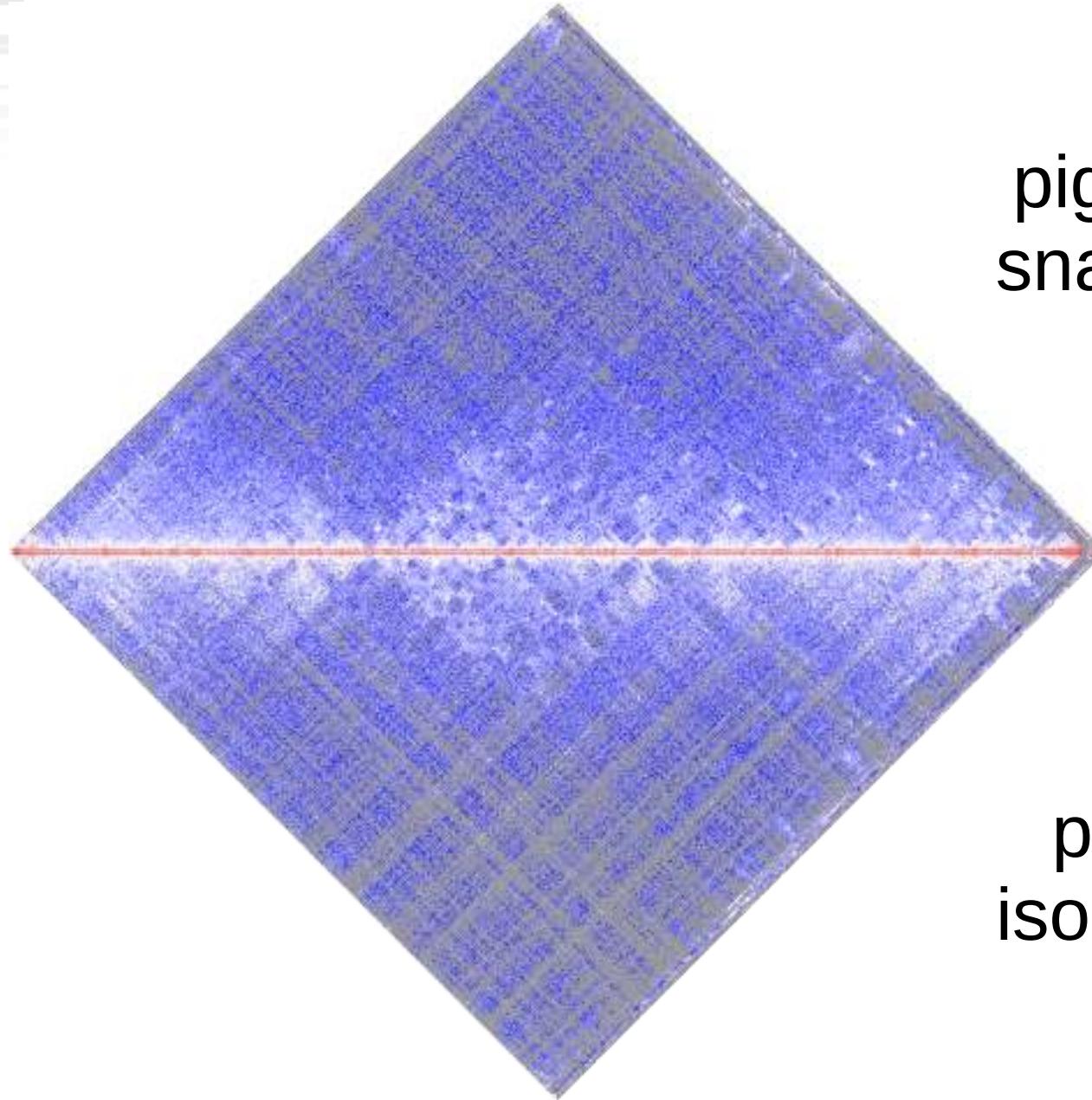
**4%
valid pairs**

FR-AgENCODE: Hi-C preliminary set-up

Mouse STO cells,
whole genome



FR-AgENCODE: Hi-C preliminary set-up

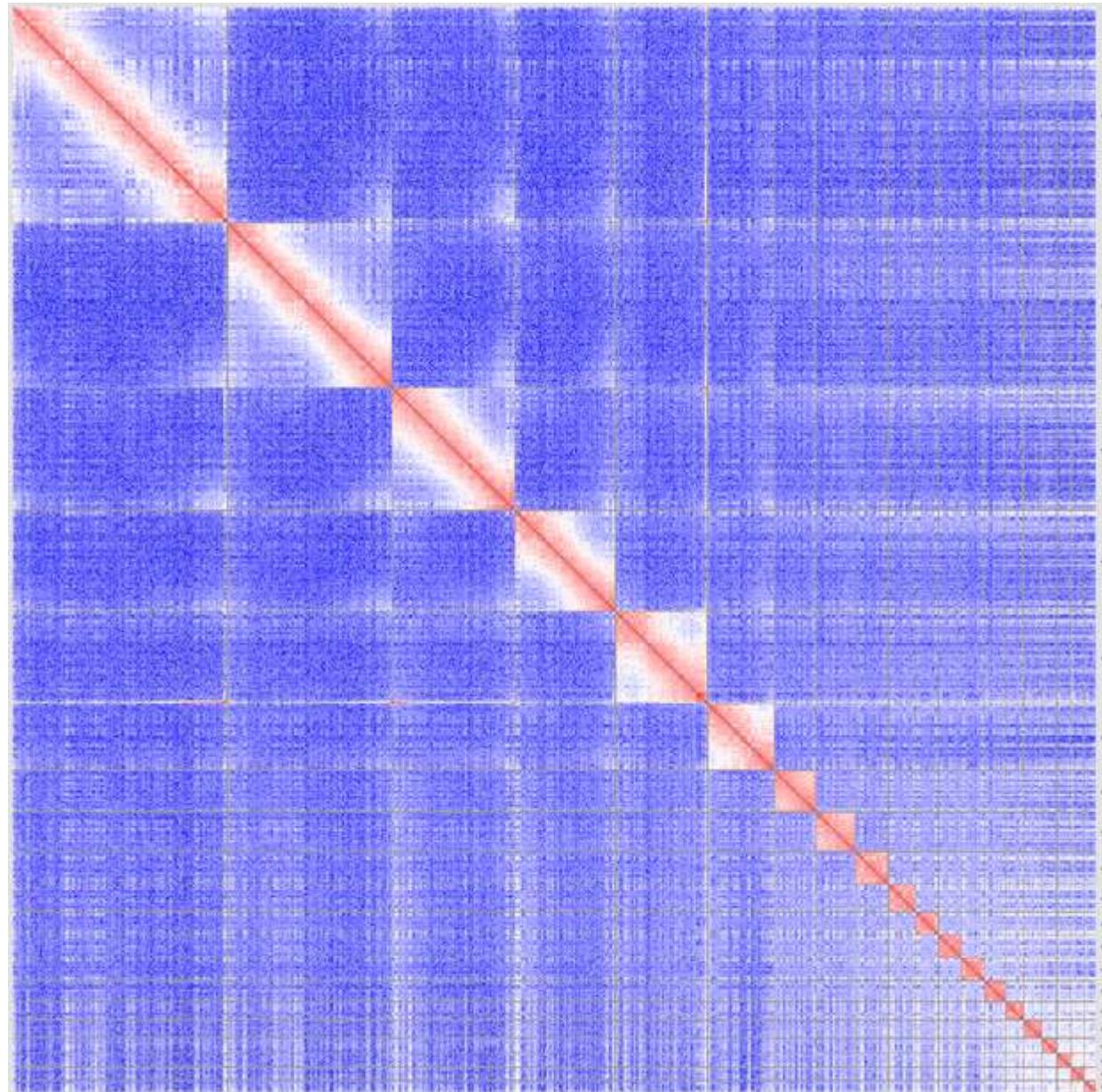


pig muscle,
snap frozen,
chr1

pig liver,
isopentane,
chr1

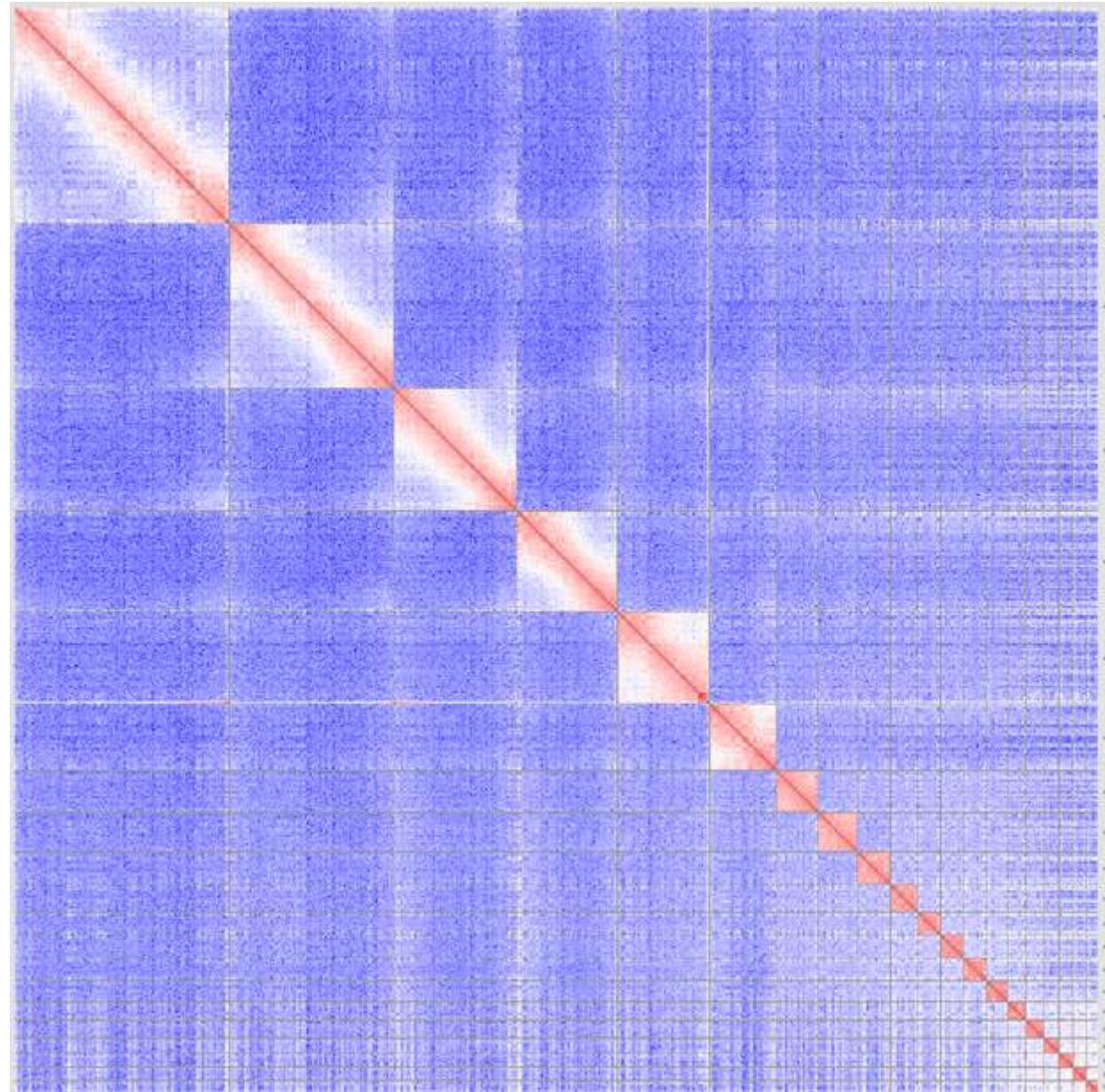
FR-AgENCODE: Hi-C preliminary results

chicken
liver,
animal #2



FR-AgENCODE: Hi-C preliminary results

chicken
liver,
animal #4



FR-AgENCODE: data management and analysis

- ◆ Data production
 - ◆ RNA-seq: 2.2T
 - ◆ Hi-C: 0.8T
 - ◆ ATAC-seq: 1.1T
- => Expected total: > 3T of raw sequence data
- ◆ Data storage
 - ◆ GenoToul ng6
 - ◆ EMBL-EBI (FAANG rapid data release policy)
- ◆ Data analysis
 - ◆ INRA units
 - ◆ EMBL-EBI (FAANG analysis pipelines)

FR-AgENCODE in FAANG

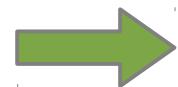
Bioinformatics and Data Analysis Committee

- ◆ aim: define standard pipelines
- ◆ Working Groups
 - ◆ transcriptome: RNA-seq, lncRNA-seq, sRNA-seq
 - ◆ regulation: ChIP-seq
 - ◆ methylation: WGBS, RRBS
 - ◆ chromatin structure: Hi-C, DNase-seq, ATAC-seq
- ◆ activities: teleconferences, seminars, hackatons
 - ◆ identification of reference datasets
 - ◆ list and benchmark tools
 - ◆ publicly report



Summary / Conclusion

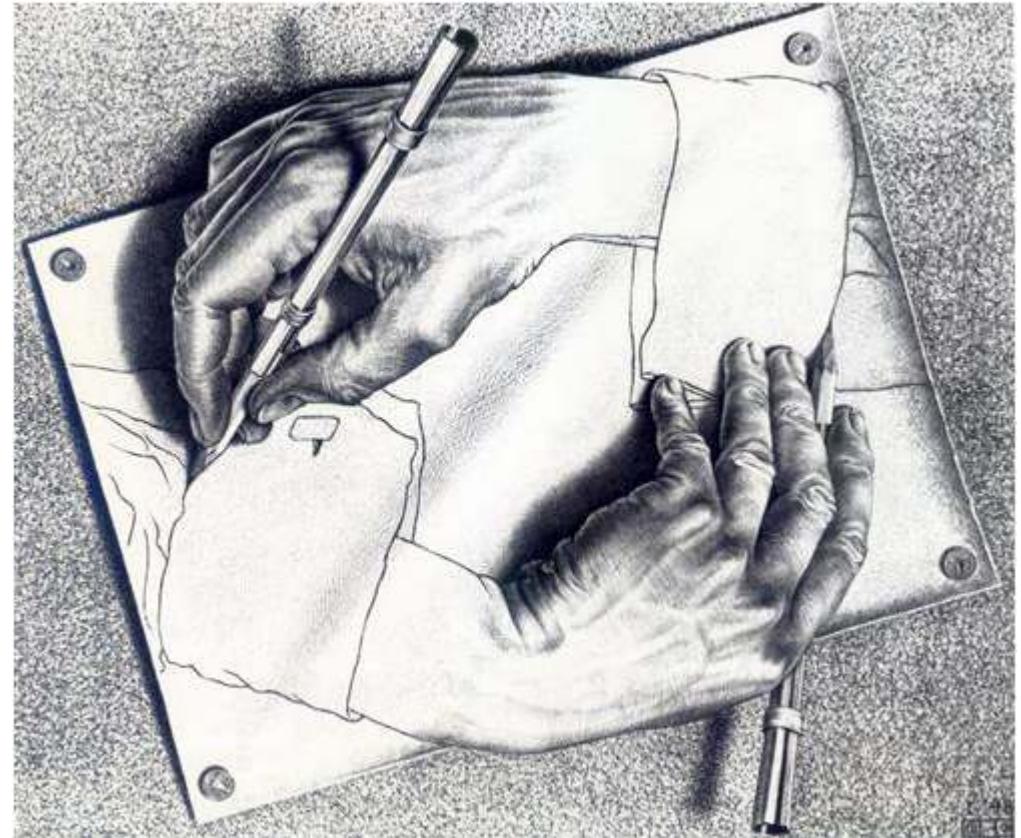
- ◆ Hi-C allows to capture 3D conformation of the chromatin
- ◆ INRA protocol for tissue samples from livestock species
- ◆ INRA contribution to the FAANG action and the Genome to Phenome challenge
 - ◆ FR-AgENCODE pilot project
 - ◆ Samples and Assays committee
 - ◆ Bioinformatics and Data Analysis committee
- ◆ Success story incoming?



To Be Continued...

Outline

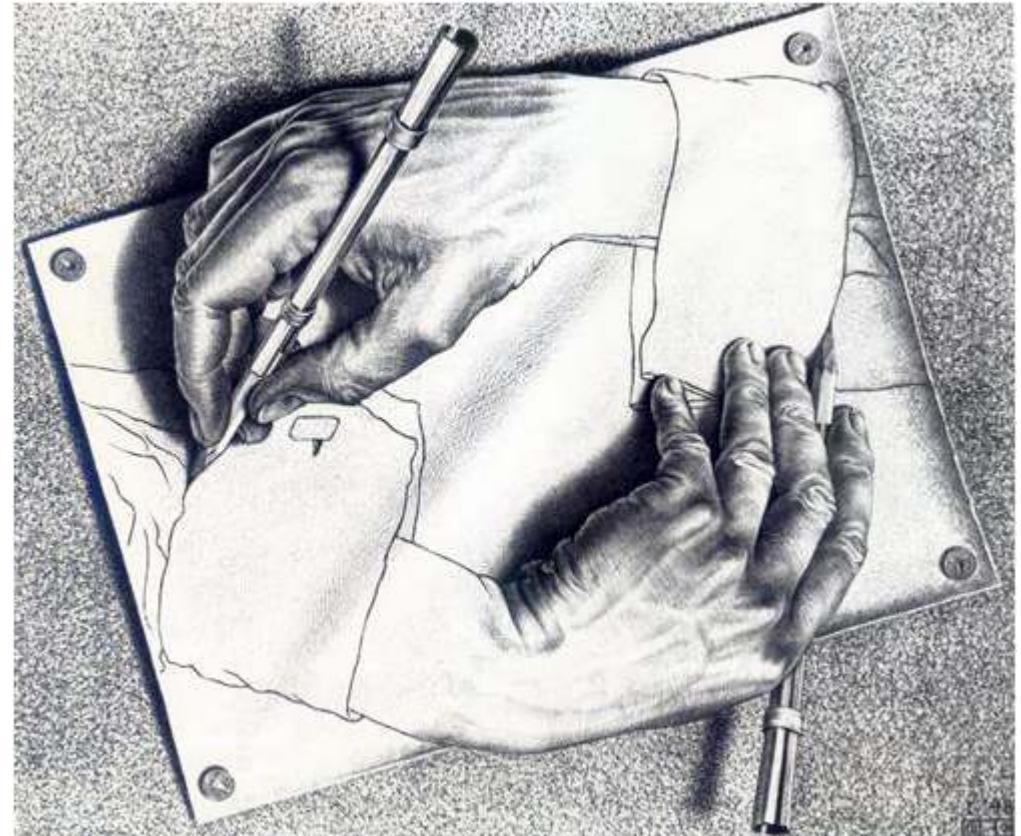
- Why
- What
- How
- Where 
- Who



M.C. Escher, 1948

Outline

- Why
- What
- How
- Where
- Who



M.C. Escher, 1948

Acknowledgments

FR-AgENCODE members

- ◆ Management: Elisabetta Giuffra (coordination), Sandrine Lagarrigue, Marie Hélène Pinard
- ◆ Sampling: Michèle Tixier-Boichard, Stéphane Fabre et al.
- ◆ Assays: **Diane Esquerré, Hervé Acloque** et al.
- ◆ Analysis: Christophe Klopp, Christine Gaspin, **David Robelin, Matthias Zytnicki, Sarah Djebali, Magali San Cristobal, Ignacio Gonzalez**, Kylie Munyard, Céline Noirot, Nathalie Villa Vialaneix, Gaelle Lefort, Marjorie Mersch, Frédérique Pitel et al.

Hi-C team @ GenPhySE, INRA

- ◆ H. Acloque, M. Yerle, Y. Lahbib, F. Mompart, M. Marti et al.

FAANG B&DA committee

- ◆ L. Clarke, D. Zerbino, J. Reecy, P. Ross, L. Eory, M. Watson et al.