

# Studying the 3D structure of the *P. falciparum*'s genome by modeling contact counts as random Negative Binomial variables.

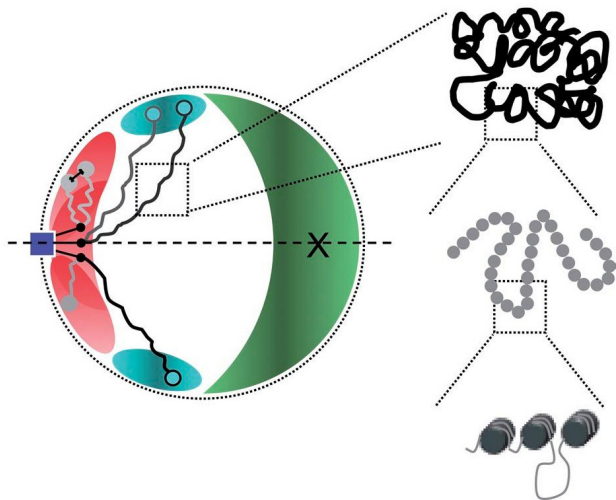
Nelle Varoquaux

*with Kate Cook, Evelien Bunnik, Ferhat Ay,  
Karine LeRoch, William Stafford Noble,  
and Jean-Philippe Vert*



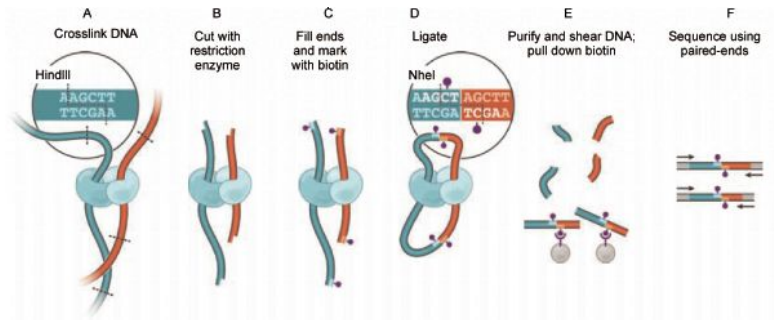
# The 3D structure of the genome is thought to play an important role in many biological processes

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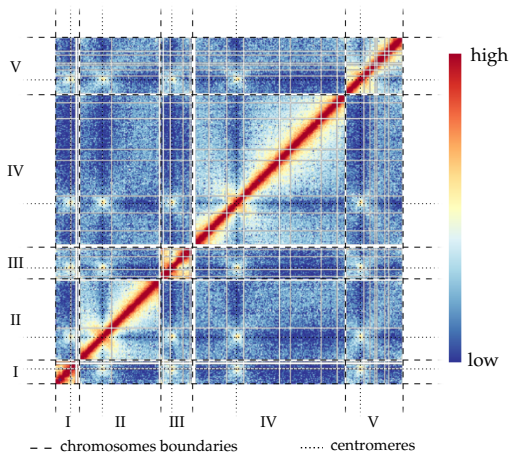
The genome of *S. cerevisiae* is highly organized [Zimmer and Fabre, 2011]

# The Hi-C protocol identifies physical contacts between pairs of loci genome-wide



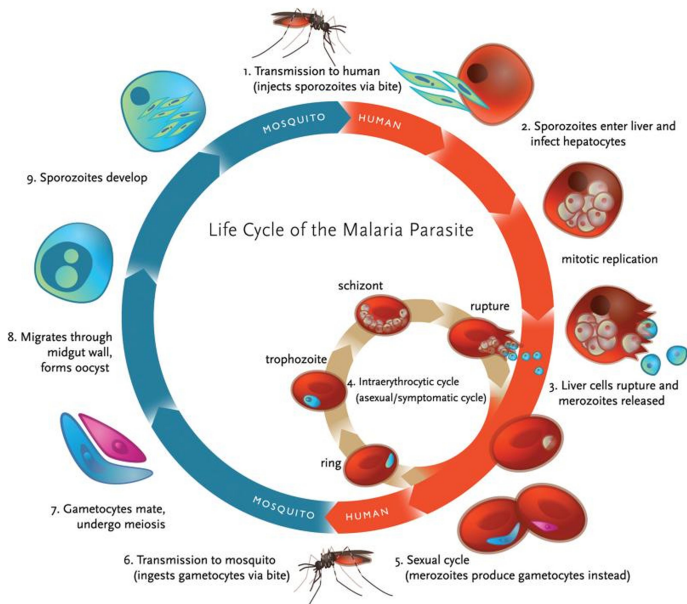
Hi-C paves the way for a systematic and genome-wide analysis of genome architecture [Rao et al., 2014]

# The contact count matrix recapitulates the hallmarks of genome architecture



Contact counts for the first 5 chromosomes of *S. cerevisiae*

# The human malaria parasite *P. falciparum*



# Motivation: the 3D structure of *P. falciparum*

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## Motivation

- One of the main limiting factors for the development of therapies is the **poor understanding of complex gene regulation** of the parasite.
- Relative **paucity of specific transcription factors** points towards **complementary regulatory mechanisms** to control gene expression.
- **Chromatin remodeling enzymes** are **abundant** in Plasmodium genomes.

## Hypothesis

- Both local and global genome architecture play an important role in *P. falciparum*'s gene regulation.

# Assessing the 3D structure changes across timepoints

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## Experiments



## Idea

- Inferring 3D models by modeling overdispersion of Hi-C data.
- Finding relationships between 3D models and gene expression.

# Inferring 3D structures of genome by modeling overdispersion of Hi-C data

*joint work with William S. Noble  
and Jean-Philippe Vert.*



# Inferring 3D models of genome architecture

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## Notations

- Let  $\mathbf{X} \in R^{n \times 3}$  be the coordinates of each bead.
- Let  $\mathbf{C}_{ij}^A \in R^{n \times n}$  be the contact count between loci  $i$  and  $j$ .
- Let  $d_{ij} = \|x_i - x_j\|_2$

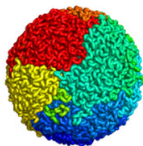
## Optimization problem

$$\underset{\mathbf{x}_1, \dots, \mathbf{x}_n}{\text{minimize}} \quad \sigma(\mathbf{X}, \mathbf{C})$$

# Relationships between contact counts $c$ , genomic distances $s$ and Euclidean distances $d$

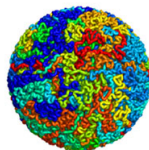
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## Fractal globule



- $c \sim s^{-1}$
- $d \sim s^{1/3}$

## Equilibrium

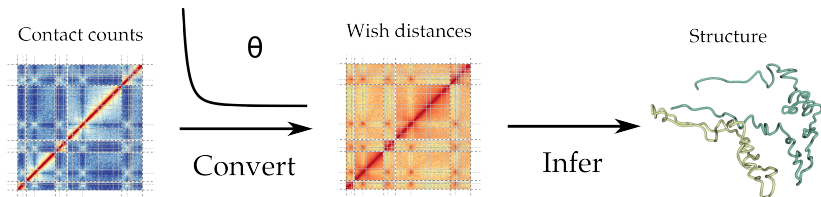


- $c \sim s^{-3/2}$
- $d \sim s^{1/2}$  for  $s < s_{\max}^{2/3}$

## Relationship between contact counts and Euclidean distances

$$d_{ij} = \gamma c_{ij}^{-1/3},$$

# Metric MDS-based methods



## Formulation

$$\underset{\mathbf{x}_1, \dots, \mathbf{x}_n}{\text{minimize}} \quad \sigma(\mathbf{X}, \mathbf{C}) = \sum_{i, j | c_{ij} \neq 0} w_{ij} (\|\mathbf{x}_i - \mathbf{x}_j\|_2 - \Theta(c_{ij}))^2$$

- $\mathbf{X}$  : 3D coordinates
- $\mathbf{C}$  : normalized contact counts.
- $w_{ij}$  are weights (set to  $\frac{1}{\Theta(c_{ij}^N)^2}$  in *pastis-MDS2*)
- $\Theta(c) = \beta c^\alpha$  : count-to-distance function

# Statistical approaches for inferring the 3D structure of the genome

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- MDS-based methods minimize an arbitrary stress function that measures the discrepancy between wish distances and 3D distances of the model.

## Statistical approach for stable inference of genome structure

- replace the arbitrary MDS loss function with a better-motivated likelihood function
- define a probabilistic model of contact counts parametrized by the 3D model.

# Inferring 3D models of genome architecture

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**The idea** Let's assume that  $c \sim \text{NegativeBinomial}(\beta d^\alpha, r)$ , where  $c$  is the interaction count,  $d$  the pairwise euclidean distance,  $r$  the dispersion parameter,  $\alpha$  unknown parameters, and  $\beta$  a scale coefficient.

## Likelihood

$$\ell(\mathbf{X}, C) = \prod_{i,j} \frac{\Gamma(c_{ij} + r)}{\Gamma(c_{ij} + 1)\Gamma(r)} \left(\frac{\beta d_{ij}^\alpha}{r + \beta d_{ij}^\alpha}\right)^{c_{ij}} \left(1 - \frac{\beta d_{ij}^\alpha}{r + \beta d_{ij}^\alpha}\right)^r \quad (1)$$

## The optimization problem

$$\max_{\alpha, \beta, \mathbf{X}} \mathcal{L}(\mathbf{X}, \alpha, \beta) = \sum_{i < j \leq n} c_{ij} \alpha \log d_{ij} - (c_{ij} + r) \log(r + \beta d_{ij}^\alpha) \quad (2)$$

# Estimating the dispersion $r$

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## Assumptions

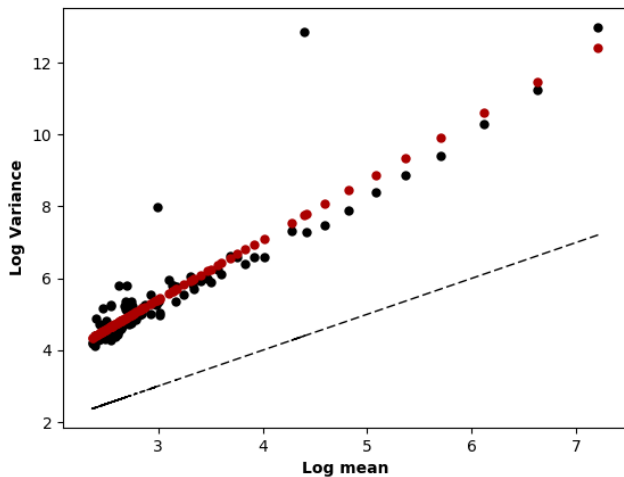
- Contact counts for pairs of loci are of the same order of magnitude.
- The variance is a smooth function of the mean.

## Estimating the dispersion $r$

- For each genomic distance  $l$ , compute the empirical mean and variance on normalized data:
  - $\hat{q}_l = \frac{1}{|I(l)|} \sum_{(i,j) \in I(l)} c_{ij}$
  - $\hat{v}_l = \frac{1}{|I(l)-1|} \sum_{(i,j) \in I(l)} (c_{ij} - \hat{q}_l)^2$
- Fit a polynomial function between  $\hat{q}$  and  $\hat{v}$
- Or estimate a constant dispersion parameter.

# Dispersion fit on *S. cerevisiae*

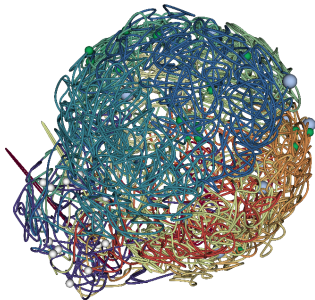
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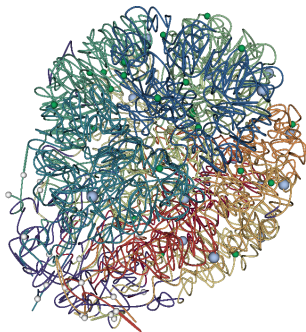
# MDS versus the Negative Binomial modeling: the case of Sporozoites *P. falciparum*

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**MDS**



**Negative Binomial**



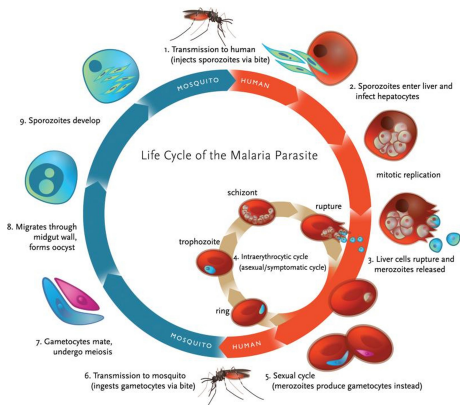
**Sporozoite stage**



**Three-dimensional modeling of the *P. falciparum* genome during reveals a strong connection between genome architecture and gene expression.**

*joint work with Evelien Bunnik, Kate Cook, Ferhat Ay, Sebastiaan Bol, Jacques Prudhomme, Jean-Philippe Vert, William S. Noble and Karine Le Roch.*

# 5 timepoints in the life cycle of *P. falciparum*



## Experiments

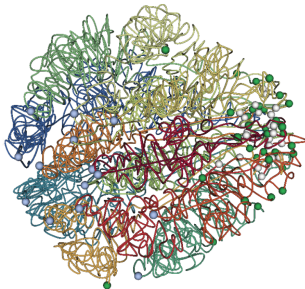
Parasite phenotype	Ring	Trophozoite	Schizont	Early and late gametocytes	Sporozoite

# 3D modeling recapitulates known organizational principles of *Plasmodium* genome

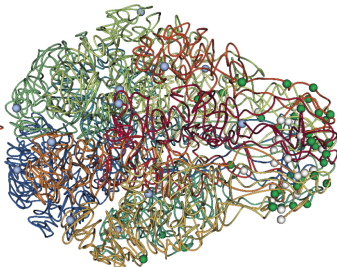
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We applied our method to the data sets thus obtaining 5 models

Stage IV/V gametocytes

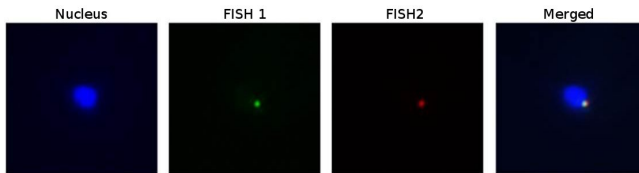
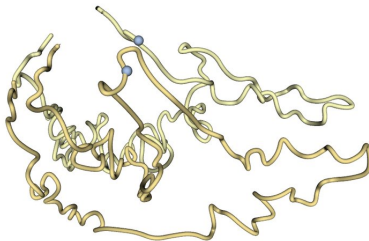


Sporozoites



# Colocalization of loci is validated with FISH

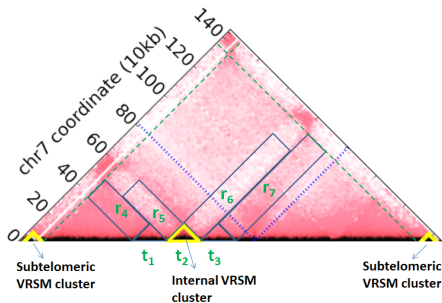
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Var genes on chromosomes VII and VIII colocalize

# Biological insights on the 3D architecture of the genome

- Virulence gene clusters on different chromosomes colocalize in 3D.
- Highly transcribed rDNA units colocalize in 3D during the ring stage.
- Transcriptionally active trophozoite stage exhibits an open chromatin structure.
- VRSM gene clusters form domain-like structures.



# Identifying links between gene expression profiles and 3D structure

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**Motivation:** Extract a gene expression profile  $v \in \mathbb{R}^p$  that is:

- representative of the gene expression profiles ;
- correlated with the 3D structure;

**Data:** For each gene  $g \in \mathcal{G}$

- Log expression profiles at 27 datapoints:  
 $e(g) = (e_1(g), \dots, e_p(g)) \in \mathbb{R}^p$  .
- Gene's 3D coordinates, extracted from the inferred 3D structure:  $x(g)$ .

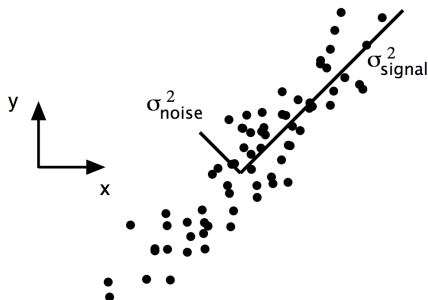
**Method:** KernelCCA [Vert and Kanehisa, 2003, Bach and Jordan, 2002]

# Extracting a vector $v$ representative of the gene expression profiles

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Find  $v \in \mathbb{R}^p$  to maximize:

$$V(v) = \frac{\sum_{g \in \mathcal{G}} (v^T e(g))^2}{\|v\|^2}$$



# Find $f$ such that $f$ is smooth with respect to the 3D structure

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Let  $f$  be a vector of scores assigned to each genes.

$$S(f) = \frac{f^\top K_{3D}^{-1} f}{\|f\|^2}$$

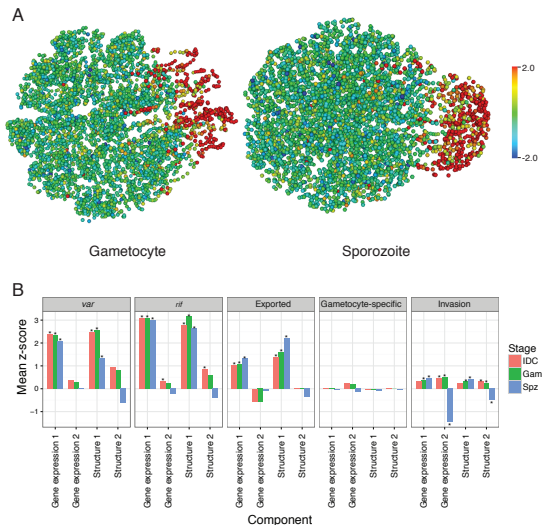
We want:

- $V(v)$  be large,
- $S(f)$  be small,
- $(v^\top e(g))_{g \in \mathcal{G}}$  and  $f$  be as correlated as possible

**This can be cast as a generalized eigenvalue problem**



# KernelCCA reveals a strong correlation between gene expression profiles and 3D structure



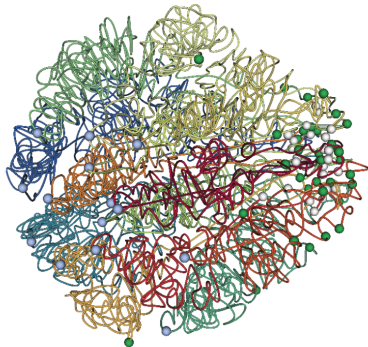
# Conclusion

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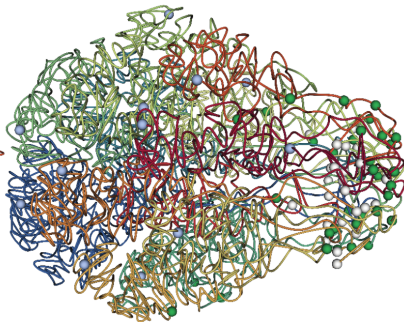
- We built high-resolution models of *P. falciparum*'s genome architecture at three time points.
- We observed :
  - strong clustering of centromeres, telomeres, virulence genes and rDNA, resulting in a **complex architecture**.
  - strong correlation between 3D genome architecture and gene expression.
- **Disruption of the parasite's genome organization** is likely to interfere with its life cycle, and could therefore be **lethal**.

# 3D models

Stage IV/V gametocytes



Sporozoites



# References I

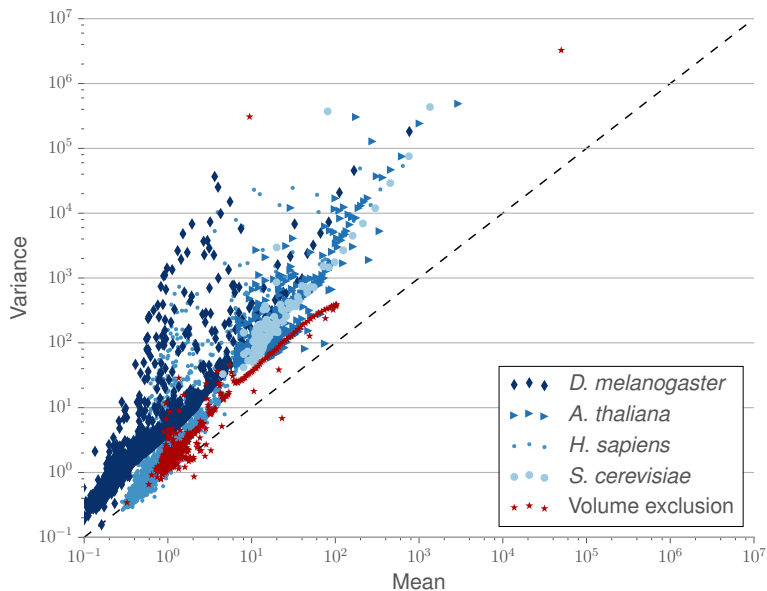
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# Contact counts are overdispersed I

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# Contact counts are overdispersed II



# Variation is greater between timepoints than between initial points I

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