Bridging Mesoscale Nucleosome Organization and Dynamics with Super Resolution Microscopy

Clayton W. Seitz

July 12, 2023

Introduction

Genome organization in eukaryotes

- ▶ The eukaryotic genome has hierarchical structure
- ▶ This structure is highly variable and often abberrant in disease

Finn et al., Science 365, 998 (2019)

A phase separation model for transcriptional control

- ► Liquid-liquid phase separation (LLPS) is a major organizer of cellular biochemistry
- ► Recent work highlights the importance of CTCF-dependent transcriptional condensates in determining cell fates

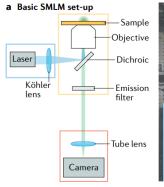
Int. J. Mol. Sci. 2022, 23(14), 8039;

Formulate the basic research question and introduce the approach using major results from section 3

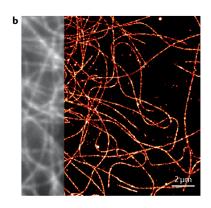
Methods

Direct stochastic optical reconstruction microscopy

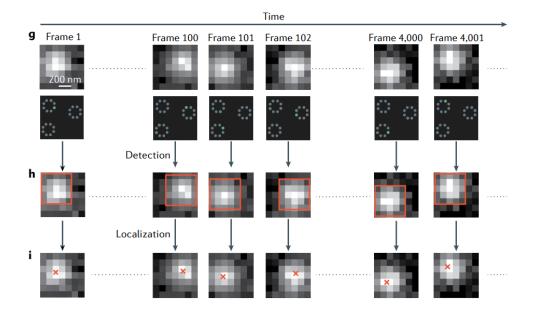
Direct stochastic optical reconstruction microscopy





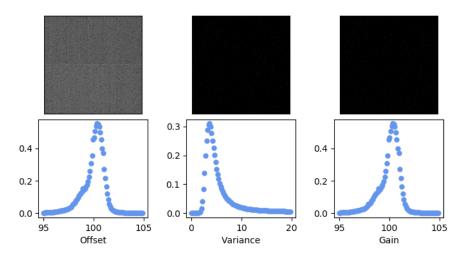


Direct stochastic optical reconstruction microscopy



Readout noise of sCMOS cameras

Hamamatsu ORCA v3 CMOS, air cooled to -10C



Measured signal: $H_k = S_k + \xi_k$, $S_k \sim \text{Poisson}(\mu_k), \xi_k \sim \mathcal{N}(o_k, \sigma_k^2)$

Maximum likelihood localization of an isolated fluorescent emitter

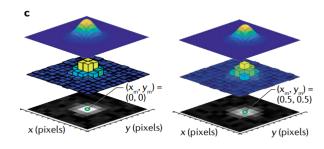
Localization:
$$\theta^* = \operatorname*{argmax}_{\theta} \prod_k P(H_k | \theta) = \operatorname*{argmin}_{\theta} - \sum_k \log P(H_k | \theta)$$

$$\mu_k = g_k \frac{\eta}{N_0} \Delta \int_{\text{pixel}} G(x, y) dA$$

 η – quantum efficiency

 N_0 – emission rate

 Δ – exposure time

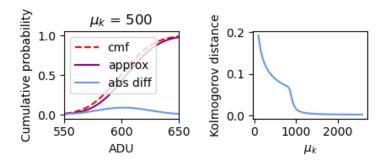


$$P(H_k|\theta) = A \sum_{r=0}^{\infty} \frac{1}{q!} e^{-\mu_k} \mu_k^q \frac{1}{\sqrt{2\pi}\sigma_k} e^{-\frac{(H_k - g_k q - o_k)}{2\sigma_k^2}}$$

 $P(H_k|\theta)$ can be approximated as Poisson at high signal-to-noise (SNR)

Quality of the Poisson approximation depends on SNR

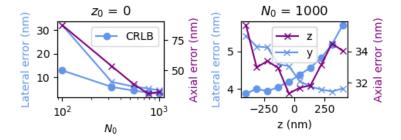
$$P(H_k|\theta) \approx \mathrm{Poisson}(\mu_k + \sigma_k^2)$$
 for $N_0 > 500$ asssuming $\Delta = 100$ ms



Using the approximation we can write

$$\ell(\vec{H}|\theta) = -\log \prod_{k} \frac{e^{-\left(\mu_{k}'\right)} \left(\mu_{k}'\right)^{n_{k}}}{n_{k}!} = \sum_{k} \log n_{k}! + \mu_{k}' - n_{k} \log \left(\mu_{k}'\right)$$

Estimator precision sets the resolution limit in localization microscopy



MLE can approach the CRLB on simulated isolated emitter data

Resolution is dependent on photoswitching kinetics

Define α as the detection probability with threshold $\delta << \Delta$

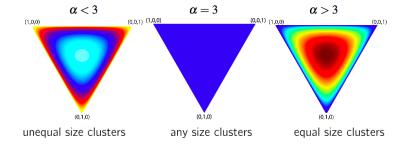
$$\alpha = \int_{\delta}^{\Delta} \left(\sum_{n=0}^{\infty} Q(N=n) \psi(t|n; \vec{k}) \right) dt \approx \underset{t \sim P}{\mathbb{E}} \left(\mathbb{I}[t > \delta] \right)$$

The number of molecules within the diffraction limit is $K\left(\frac{\lambda}{2\mathrm{NA}}\right)$. then $\alpha K\left(\frac{\lambda}{2\mathrm{NA}}\right)$ are detected, on average. Higher density gives more errors but decreases the SR frame duration Δ_{SR} . Thus we define

$$D = lpha K \left(rac{\lambda}{2 \mathrm{NA}}
ight) \ \ T = \left(\Delta_{SR} + rac{2N}{\log(1 - lpha)}
ight)^2$$

Deep learning enables accurate 3D localization and single molecule tracking

Dirichlet process Gaussian mixture model (DPGMM)



Results

GMM cluster analysis of H2B

Number of clusters is unknown apriori - Bayesian nonparametrics

Mesoscale nucleosome organization and dynamics

Mesoscale nucleosome organization and dynamics

BRD4 associates with the small clusters

Besag's L-Function

Besag's L-Function and 3D diffusion