ADVANCING SUPER RESOLUTION MICROSCOPY FOR QUANTITATIVE IN-VIVO IMAGING OF CHROMATIN NANODOMAINS

by

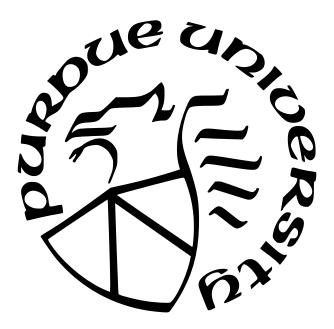
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A Dissertation

Submitted to the Faculty of Purdue University

In Partial Fulfillment of the Requirements for the degree of

Doctor of Philosophy



Department of Physics and Astronomy
West Lafayette, Indiana
December 2024

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LIST OF SYMBOLS

- m mass
- v velocity

ABBREVIATIONS

abbr abbreviation

bcf billion cubic feet

BMOC big man on campus

ABSTRACT

Single-molecule localization microscopy (SMLM) techniques, such as direct stochastic optical reconstruction microscopy (dSTORM), can be used to produce a pointillist representation of fluorescently-labeled biological structures at diffraction-unlimited precision. Direct STORM approaches leverage the deactivation of standard fluorescent tags, followed by spontaneous or photoinduced reactivation, allowing resolution of fluorophores at distances below the diffraction limit. This basic principle remains one of the method's primary limitations standard SMLM fitting routines require tight control of activation and reactivation to maintain sparse emitters, presenting a tradeoff between imaging speed and labeling density. Here, I present two parallel projects, which aim to push the current state of the art in SMLM and apply SMLM to the study of gene regulation. The former represents a novel localization technique for dense SMLM, based on deep probabilistic modeling and photon statistics. In the latter, conventional dSTORM is adapted for live cell imaging of chromatin nanodomains, demonstrating that BRD4 protein concentrates in nucleosome depleted regions.

1. INTRODUCTION

Single molecule localization microscopy (SMLM) relies on the temporal resolution of fluorophores in the sample whose spatially overlapping point spread functions would otherwise render them unresolvable at the detector. SMLM techniques, such as stochastic optical reconstruction microscopy (STORM) and photo-activated localization microscopy (PALM) remain desirable for super-resolution imaging of many cellular structures, due to their cost-effective implementation and diffraction unlimited resolution (Schermelleh 2019). Common strategies for the temporal separation of molecules involve transient intramolecular rearrangements to switch from dark to fluorescent states or the exploitation of non-emitting molecular radicals. For direct STORM (dSTORM), rhodamine derivatives can undergo intersystem crossing to a triplet state, which can be reduced by thiols to form a dark radical species. The dark state can then be quenched by oxidative processes, driving the fluorophore back to its ground state (Figure 1a). Long dark state lifetimes are commonly used in STORM imaging in order to maintain sparse activation and high resolution.

1.0.1 The definition of resolution in SMLM

The distribution of a particular biomolecule in the cell can be described as a probability density over a two-dimensional space, casting super-resolution as a density estimation problem. Intuitively, the spatial resolution of SMLM images then increases as we draw more samples from this density - a concept which is made mathematically precise by the so-called Fourier ring correlation or FRC. Using FRC, one can compute image resolution as the spatial frequency at which a correlation function in the frequency domain drops below a threshold, typically taken to be 1/7 (See Supplement). According to this theory, reducing localization uncertainty while increasing the number of samples, results in an increase in image resolution (Nieuwenhuizen 2013). However, there remains a fundamental limit to the the minimal localization uncertainty which can be obtained.

Localization uncertainty, typically the RMSE of a maximum likelihood or similar statistical estimator, is bounded from below by the inverse of the Fisher information matrix, known as the Cramer-Rao lower bound (Chao 2016). Localization uncertainties in sparse

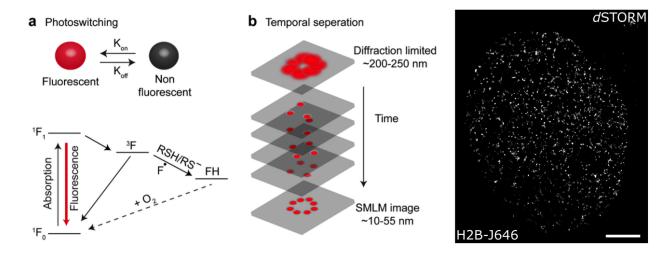


Figure 1.1. Stochastic optical reconstruction microscopy (STORM). (A) Single molecules are resolved by separating their fluorescent emission in time, using fluorophores with multiple photophysical states (B) Example super-resolution image of H2B protein in a living Hela cell nucleus at 37C, 5 percent CO2. Image reconstructed from 10³ 10ms frames. Scalebar 5um.

conditions are often tens of nanometers, although recent work on integration of Bayesian priors with modulation enhanced SMLM (meSMLM) or structured illumination with MIN-FLUX, has reduced spatial resolution below to a few nanometers (Kalisvaart 2022, Gwosh 2020). Nevertheless, managing the increase in localization uncertainty at high labeling density remains a major bottleneck to SMLM. Static uncertainty due to molecular crowding can be partially amelioriated by using pairwise or higher-order temporal correlations within a pixel neighborhood, known as stochastic optical fluctuation imaging or SOFI (Dertinger 2009). Other approaches such as stimulated emission and depletion (STED) imaging bring control over the photophysical state of a chosen subset of the sample, yet the need for laser scanning prevents widespread application in live-cell studies. The spatial resolution and relative simplicity of SMLM techniques remains unmatched, inciting an effort to increase the resolution of SMLM techniques and explore avenues towards time resolved SMLM.

1.0.2 Novel contributions to the field

Here, we present single photon counting enhanced SMLM and a technique for superresolution imaging of chromatin nanodomains in-vivo. We address dense SMLM by utilizing a high-speed single photon avalanche diode (SPAD) camera. SPAD cameras provide the necessary functionality for both single molecule localization and counting. Then, we present a more conventional dSTORM approach to study spatial organization of nucleosomes in living cells, with a particular focus on the structure of phase separated condensates containing bromodomain protein 4 (BRD4) protein.

A. CITATIONS AND REFERENCES

1 \chapter{CITATIONS AND REFERENCES}

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B. PHYSICS

B.0.1 Spatial coherence for an isolated emitter

Photoswitching fluorescent molecules are described in the density matrix formalism

$$\rho = \sum_{k} \xi_k |\alpha_k\rangle \langle \alpha_k| \quad \sum_{k} \xi_k = 1$$

where $|\alpha_k\rangle$ is a coherent state with amplitude α_k i.e., $\langle n\rangle = \langle \alpha_k | n | \alpha_k \rangle = |\alpha_k^2|$. Typically ξ_k and $\langle n_k \rangle$ are heterogeneous. We consider a simplified model consisting of a single mode field

$$E^{+}(r_{i}) = h(r_{i} - s_{0})\hat{a}_{n}$$

$$g_{ij}^{(2)}(0) = \frac{\langle E^{-}(r_i)E^{-}(r_j)E^{+}(r_i)E^{+}(r_j)\rangle}{\langle E^{-}(r_i)E^{+}(r_i)\rangle\langle E^{-}(r_i)E^{+}(r_i)\rangle} = \frac{\operatorname{Tr}(E^{-}(r_i)E^{-}(r_j)E^{+}(r_i)E^{+}(r_j)\rho)}{\operatorname{Tr}(E^{-}(r_i)E^{+}(r_i)\rho)\operatorname{Tr}(E^{-}(r_i)E^{+}(r_i)\rho)}$$

Terms related to point spread function will cancel. It is instructive to compute

$$\operatorname{Tr}(a^{\dagger}a^{\dagger}aa\left(\xi_{k} | \alpha_{k}\right) \langle \alpha_{k}|) = \operatorname{Tr}\left(\xi_{k} e^{-|\alpha|^{2}} \sum_{n,m}^{\infty} \frac{\alpha^{n}}{n!} | n \rangle \langle m|\right)$$

$$= \operatorname{Tr}\left(\xi_{k} e^{-|\alpha|^{2}} \sum_{n}^{\infty} \frac{|\alpha|^{2n}}{n!} n(n-1)\right)$$

$$= \operatorname{Tr}\left(\xi_{k} e^{-|\alpha|^{2}} \sum_{n=2}^{\infty} \frac{|\alpha|^{2n}}{(n-2)!}\right)$$

$$= \xi_{k} |\alpha_{k}|^{4}$$

Similarly,

$$\operatorname{Tr}(a^{\dagger}a\left(\xi\left|\alpha\right\rangle\left\langle\alpha\right|)\right) = \operatorname{Tr}\left(\xi e^{-|\alpha|^{2}} \sum_{n,m}^{\infty} \frac{\alpha^{n}(\alpha^{m})^{*}}{\sqrt{n!}\sqrt{m!}} a^{\dagger}a\left|n\right\rangle\left\langle m\right|\right)$$

$$= \xi e^{-|\alpha|^{2}} \sum_{n=0}^{\infty} \frac{(|\alpha|^{2})^{n}}{n!} n$$

$$= \xi e^{-|\alpha|^{2}} \sum_{n=1}^{\infty} \frac{(|\alpha|^{2})^{n}}{(n-1)!}$$

$$= \xi e^{-|\alpha|^{2}} \left(|\alpha|^{2} + \frac{|\alpha|^{4}}{1!} + \frac{|\alpha|^{6}}{2!} + \ldots\right)$$

$$= \xi e^{-|\alpha|^{2}} |\alpha|^{2} \left(1 + \frac{|\alpha|^{2}}{1!} + \frac{|\alpha|^{3}}{2!} + \ldots\right)$$

$$= \xi e^{-|\alpha|^{2}} e^{|\alpha|^{2}} |\alpha|^{2} = \xi |\alpha|^{2}$$

$$\operatorname{Tr}(aa^{\dagger}(\xi | \alpha) \langle \alpha |)) = \operatorname{Tr}\left(\xi e^{-|\alpha|^{2}} \sum_{n,m}^{\infty} \frac{\alpha^{n}(\alpha^{m})^{*}}{\sqrt{n!}\sqrt{m!}} aa^{\dagger} | n \rangle \langle m |\right)$$

$$= \xi e^{-|\alpha|^{2}} \sum_{n=0}^{\infty} \frac{(|\alpha|^{2})^{n}}{n!} (n+1)$$

$$= \xi e^{-|\alpha|^{2}} \left(\sum_{n=1}^{\infty} \frac{(|\alpha|^{2})^{n}}{(n-1)!} + e^{|\alpha|^{2}}\right)$$

$$= \xi e^{-|\alpha|^{2}} \left(|\alpha|^{2} e^{|\alpha|^{2}} + e^{|\alpha|^{2}}\right) = \xi(|\alpha|^{2} + 1)$$

Putting it all together yields a simple expression for the two-point coherence function

$$g_{ij}^{(2)}(0) = \frac{\sum_{k} \xi_{k} |\alpha_{k}|^{4}}{(\sum_{k} \xi_{k} |\alpha_{k}|^{2}) (\sum_{k} \xi_{k} |\alpha_{k}|^{2})}$$

For example, if we have a two-level system consisting of a fluorescent state with amplitude α and the vacuum state, this becomes

$$g_{ij}^{(2)}(0) = \frac{\xi |\alpha|^4}{\xi^2 |\alpha|^4} = \frac{1}{\xi}$$

As $\xi \to 1$ (always on) we recover a coherent state. As $\xi \to 0$ we observe $g_{ij}^{(2)}(0) > 1$ i.e., bunching.

B.0.2 Generalization to nonzero background

$$E_0^+ \sim \sum_{j=1}^M \delta(s - s_j) a_j \ E^+(r_i) = \int d^2 s E_0^+ = \sum_n h(r_i - s_n) a_n$$

$$\rho_S = \xi |\alpha\rangle \langle \alpha| + (1 - \xi) |0\rangle \langle 0| \quad \rho_B = |\beta\rangle \langle \beta| \quad \rho = \rho_S \otimes \rho_B$$

$$E(r_i)^+ = E_S(r_i)^+ + E_B(r_i)^+ = h(r_i - s_n)a_S + a_B$$

$$G_{ij}^{2}(0) = \langle (E_{S}^{\dagger} + E_{B}^{\dagger})(E_{S}^{\dagger} + E_{B}^{\dagger})(E_{S} + E_{B})(E_{S} + E_{B})\rangle$$

$$= h_{i}^{2}h_{j}^{2}\langle a_{S}^{\dagger}a_{S}^{\dagger}a_{S}a_{S}\rangle + h_{i}^{2}\langle a_{S}^{\dagger}a_{B}^{\dagger}a_{S}a_{B}\rangle + h_{j}^{2}\langle a_{B}^{\dagger}a_{S}^{\dagger}a_{B}a_{S}\rangle + \langle a_{B}^{\dagger}a_{B}^{\dagger}a_{B}a_{B}\rangle$$

$$= \xi(h_{i}^{2}h_{j}^{2}|\alpha|^{4} + h_{i}^{2}|\alpha|^{2}|\beta|^{2} + h_{j}^{2}|\alpha|^{2}|\beta|^{2}\rangle + |\beta|^{4})$$

$$= \xi(h_{i}^{2}h_{j}^{2}|\alpha|^{4} + |\alpha|^{2}|\beta|^{2}(h_{i}^{2} + h_{j}^{2}) + |\beta|^{4})$$

The normalized second order coherence function then reads

$$g_{ij}^{2}(0) = \frac{\xi h_{i}^{2} h_{j}^{2} N_{0}^{2} + \xi N_{0} B_{0}(h_{i}^{2} + h_{j}^{2}) + B_{0}^{2}}{\xi^{2} h_{i}^{2} h_{j}^{2} N_{0}^{2} + \xi N_{0} B_{0}(h_{i}^{2} + h_{j}^{2}) + B_{0}^{2}}$$

Notice the PSF factor h_i appears squared. This squared value can be seen as the probability of photon detection at a point s_i , while h_i is the amplitude of the electric field.

B.0.3 Ergodicity of photoswitching

In general, Markov jump processes are non-ergodic, meaning that their time averages and ensemble averages are not equal. The use of ξ as a probability is only valid when the observation duration (exposure time) is much longer than the characteristic switching time. However, the use of ξ above is quite convenient, so we look to determine how long our exposure must be for the emitter to be considered in equilibrium. For a two state process

$$P(t) = e^{Wt}P(0) \to \dot{P}(t) = We^{Wt}P(0)$$

$$e^{Wt} = I + W \frac{1 - e^{-2\lambda t}}{2\lambda}$$

where $\lambda = (\lambda_1 + \lambda_2)/2$. Now,

$$\dot{P}(t) = W\left(I + W\frac{1 - e^{-2\lambda t}}{2\lambda}\right)P(0)$$

It can then be shown that the individual gradients are

$$\begin{pmatrix} \frac{\lambda_1\lambda_2\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2} - \lambda_1\left(1-\frac{\lambda_1\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2}\right) & \lambda_2\left(1-\frac{\lambda_2\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2}\right) - \frac{\lambda_1\lambda_2\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2} \\ \lambda_1\left(1-\frac{\lambda_1\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2}\right) - \frac{\lambda_1\lambda_2\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2} & \frac{\lambda_1\lambda_2\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2} - \lambda_2\left(1-\frac{\lambda_2\left(1-\mathrm{e}^{-t(\lambda_1+\lambda_2)}\right)}{\lambda_1+\lambda_2}\right) \end{pmatrix}$$

B.0.4 Details of the Gaussian PSF

We will derive the gradients for the integrated astigmatic Gaussian, since it is the more general case. As before, define $i_0 = g_k \gamma \Delta t N_0$ such that $\mu'_k = i_0 \lambda_k$

$$J_{x_0} = \beta_k \lambda_y \frac{\partial \lambda_x}{\partial x_0} \quad J_{y_0} = \beta_k \lambda_x \frac{\partial \lambda_y}{\partial y_0} \quad J_{z_0} = \frac{\partial \mu_k'}{\partial \sigma_x} \frac{\partial \sigma_x}{\partial z_0} + \frac{\partial \mu_k'}{\partial \sigma_y} \frac{\partial \sigma_y}{\partial z_0}$$

$$J_{x_0} = \beta_k \lambda_y \frac{\partial \lambda_x}{\partial x_0}$$

$$= \frac{\beta_k \lambda_y}{2} \frac{\partial}{\partial x_0} \left(\operatorname{erf} \left(\frac{x_k + \frac{1}{2} - x_0}{\sqrt{2}\sigma_x} \right) - \operatorname{erf} \left(\frac{x_k - \frac{1}{2} - x_0}{\sqrt{2}\sigma_x} \right) \right)$$

$$= \frac{\beta_k \lambda_y}{\sqrt{2\pi}\sigma_x} \left(\exp \left(\frac{(x_k - \frac{1}{2} - x_0)^2}{2\sigma_x^2} \right) - \exp \left(\frac{(x_k + \frac{1}{2} - x_0)^2}{2\sigma_x^2} \right) \right)$$

$$J_{y_0} = \beta_k \lambda_x \frac{\partial \lambda_y}{\partial y_0}$$

$$= \frac{\beta_k \lambda_x}{2} \frac{\partial}{\partial y_0} \left(\operatorname{erf} \left(\frac{y_k + \frac{1}{2} - y_0}{\sqrt{2}\sigma_y} \right) - \operatorname{erf} \left(\frac{y_k - \frac{1}{2} - y_0}{\sqrt{2}\sigma_y} \right) \right)$$

$$= \frac{\beta_k \lambda_x}{\sqrt{2\pi}\sigma_y} \left(\exp \left(\frac{(y_k - \frac{1}{2} - y_0)^2}{2\sigma_y^2} \right) - \exp \left(\frac{(y_k + \frac{1}{2} - y_0)^2}{2\sigma_y^2} \right) \right)$$

$$J_{\sigma_x} = \beta_k \lambda_y \frac{\partial \lambda_x}{\partial \sigma_x}$$

$$= \frac{\beta_k \lambda_y}{2} \frac{\partial}{\partial \sigma_x} \left(\operatorname{erf} \left(\frac{x_k + \frac{1}{2} - x_0}{\sqrt{2}\sigma_x} \right) - \operatorname{erf} \left(\frac{x_k - \frac{1}{2} - x_0}{\sqrt{2}\sigma_x} \right) \right)$$

$$= \frac{\beta_k \lambda_y}{\sqrt{2\pi}} \left(\frac{\left(x - x_0 - \frac{1}{2} \right) e^{-\frac{\left(x - x_0 - \frac{1}{2} \right)^2}{2\sigma_x^2}}}{\sigma_x^2} - \frac{\left(x - x_0 + \frac{1}{2} \right) e^{-\frac{\left(x - x_0 + \frac{1}{2} \right)^2}{2\sigma_x^2}}}{\sigma_x^2} \right)$$

$$J_{\sigma_y} = \beta_k \lambda_x \frac{\partial \lambda_y}{\partial \sigma_y}$$

$$= \frac{\beta_k \lambda_x}{2} \frac{\partial}{\partial \sigma_y} \left(\operatorname{erf} \left(\frac{y_k + \frac{1}{2} - y_0}{\sqrt{2}\sigma_y} \right) - \operatorname{erf} \left(\frac{y_k - \frac{1}{2} - y_0}{\sqrt{2}\sigma_y} \right) \right)$$

$$= \frac{\beta_k \lambda_x}{\sqrt{2\pi}} \left(\frac{\left(y - y_0 - \frac{1}{2} \right) e^{-\frac{\left(y - y_0 - \frac{1}{2} \right)^2}{2\sigma_y^2}}}{\sigma_y^2} - \frac{\left(y - y_0 + \frac{1}{2} \right) e^{-\frac{\left(y - y_0 + \frac{1}{2} \right)^2}{2\sigma_y^2}}}{\sigma_y^2} \right)$$

Luckily, computing the Hessian matrix for (2.9) is tractable, and is actually quite simple when one takes advantage of the chain rule for Hessian matrices. Looking at (2.9), the likelihood is a hierarchical function that maps a vector space Θ to a vector space Λ to a scalar value. Formally, we define $T:\Theta\to\Lambda$ and $W:\Lambda\to\mathbb{R}$. The parameter vector $(x_0,y_0,z_0,\sigma_0,N_0)\in\Theta$, the Poisson rate vector $\vec{\lambda}\in\Lambda$ and $\ell\in\mathbb{R}$. Note that we choose to optimize σ_x and σ_y directly and compute z_0 to simplify the computation of the Hessian. To get the Hessian, we need the chain-rule for Hessian matrices, which can be quickly computed in terms of the jacobian and hessian of T and W.

$$H_{\ell} = J_{\mu}^{T} H_{\ell} J_{\mu} + (J_{\ell} \otimes I_{n}) H_{\mu}$$

where we have used J_{μ} to represent the jacobian of T and J_{ℓ} for the jacobian of W. Similar notation is used for the corresponding Hessian matrices. In the 3D case, the Hessian matrix is not directly separable since $\mu \propto \lambda_x(x_0, \sigma_0, \sigma_x)\lambda_y(y_0, \sigma_0, \sigma_y)$. To see this, an abstract representation of the Hessian reads

B.0.5 Fisher information for 2D integrated gaussian

For the 2D integrated gaussian point spread function, the Hessian only contains separable second order derivatives, so the Fisher information matrix takes on a convenient form

$$I_{ij}(\theta) = \mathbb{E}\left(\frac{\partial \ell}{\partial \theta_i} \frac{\partial \ell}{\partial \theta_j}\right)$$
 (B.1)

For an arbitrary parameter then we have

$$\frac{\partial \ell}{\partial \theta_{i}} = \frac{\partial}{\partial \theta_{i}} \sum_{k} x_{k} \log x_{k} + \mu'_{k} - x_{k} \log (\mu'_{k})$$
$$= \sum_{k} \frac{\partial \mu'_{k}}{\partial \theta_{i}} \left(\frac{\mu'_{k} - x_{k}}{\mu'_{k}} \right)$$

$$I_{ij}(\theta) = \mathbb{E}\left(\sum_{k} \frac{\partial \mu_{k}'}{\partial \theta_{i}} \frac{\partial \mu_{k}'}{\partial \theta_{j}} \left(\frac{\mu_{k}' - x_{k}}{\mu_{k}'}\right)^{2}\right) = \sum_{k} \frac{1}{\mu_{k}'} \frac{\partial \mu_{k}'}{\partial \theta_{i}} \frac{\partial \mu_{k}'}{\partial \theta_{j}}$$

To compute the bound, it turns out all we need is the jacobian $\frac{\partial \mu'_k}{\partial \theta_i}$.

VITA

 $[{\rm Put\ a\ brief\ autobiographical\ sketch\ here.}]$

PUBLICATION(S)

The following is based on information in template1, template2, template3.

In a publication or publications section you can

- list a single publication
- include a single publication
- list multiple publications
- include multiple publications

Use

```
\begin{publication}...\end{publication}
```

or

```
\begin{publications}...\end{publications}
```

to skip to the next page and put the appropriate heading on the top of the page.

To List a Single Publication

```
\begin{publication}
    ...list a single publication here...
    ...IMPROVE THIS LATER to show how to do that...
\end{publication}
```

To Include a Single Publication

```
\begin{publication}
    ...put a single publication here...
    ...IMPROVE THIS LATER to show how to do that...
\end{publication}
```

To List Multiple Publications

```
\begin{publications}
...list multiple publications here...
...IMPROVE THIS LATER to show how to do that...
\end{publications}
```

To Include Multiple Publications

```
\begin{publications}
    ...put the multiple publications here...
    ...IMPROVE THIS LATER to show how to do that...
\end{publications}
```