Precise Nanometer Localization Analysis for Individual Fluorescent Probes

based on a paper by R.E. Thompson, D.R. Larson and W.W. Webb

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Outline

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1. Motivation

Minimal resolution of light microscopes

 $r \approx \frac{\lambda}{2} \approx 250$ nm (diffraction limit)

 (Left to right) A mammalian cell, a bacterial cell, a mitochondrion, an influenza virus, a ribosome, the green fluorescent protein, and a small molecule (thymine).



Source:

Huang et al., Breaking the Diffraction Barrier: Super-Resolution Imaging of Cells, Cell 143 (2010), p. 1047-1058



- Stochastical optical reconstruction
- microscopy (STORM)
- Reconstruction of spot centers
- Assumption: punctate and well-separated objects

Source:

T.J. Gould, S.T. Hess, Nanoscale Biological Fluorescence Imaging: Breaking the Diffraction Barrier, Methods in Cell Biology 89 (2008), p. 329-358



- Aggravated by noise
 - Background noise
 - Photon shot noise



2. Algorithms

- Given: matrix of photon counts for each pixel
- 2.a. Spot recognition
 - Find pixel values above some threshold (e.g. "8 standard deviations away from the mean")
 - Look for local maxima
 - Cut out a surrounding regions according to the potential spot
 - Mean remaining cells to obtain the background mean
 - Subtract the background mean from the spots
 - Fit the spot centers

2.b. Fitting algorithms

• Given: matrix of photon counts S_{ij}

Notations

- (i, j) location of pixel center in local coordinates
- Spot center (x_0, y_0) unknown
- Number of photons in spot N unknown
- \circ Standard deviation σ known

$$p_G(i,j) = \frac{1}{2\pi\sqrt{\sigma}} \exp(-\frac{(i-x_0)^2}{2\sigma^2} - \frac{(j-y_0)^2}{2\sigma^2})$$

• We want: approximation of the center of *S_{ij}* by fitting with a <u>pixelated</u> Gaussian, i.e.

$$G_{ij} = Np_G(i,j)$$



Least squares approach, i.e. we want to minimize $\chi^2 = \sum_{i,j} \frac{(S_{ij} - G_{ij})^2}{\psi_{ij}^2}$ (here $\psi_{ij}^2 = \sigma^2 + b^2$) For every minimum of χ^2 : $\frac{d}{dx_0} \sum \left(S_{ij} - G_{ij} \right)^2 = 0$ $\Leftrightarrow \sum 2(S_{ij} - G_{ij}) \frac{(i - x_0)}{\sigma^2} G_{ij} = 0$ $\Leftrightarrow \sum S_{ij}G_{ij}(i-x_0) + \sum G_{ij}^2(i-x_0) = 0$

Full least squares

$$x_0 = \frac{\sum i(S_{ij} - G_{ij})G_{ij}}{\sum (S_{ij} - G_{ij})G_{ij}}$$

Odd symmetry: $\sum G_{ij}^{2}(i - x_{0}) \approx 0$ Gaussian mask fitting $x_{0} = \frac{\sum i S_{ij} G_{ij}}{\sum S_{ij} G_{ij}} = \frac{\sum i S_{ij} p_{G}(i,j)}{\sum S_{ij} p_{G}(i,j)}$ Number of photons within spot $N = \frac{\sum S_{ij} p_{G}(i,j)}{\sum p_{C}(i,i)^{2}}$



3. Error estimation

- Split into two cases
 - Photon shot noise-limited (a)
 - Background noise-limited (b)

Case (a)

• No pixelation:
$$\langle (\Delta x)^2 \rangle = \frac{Var(x)}{N} = \frac{\sigma^2}{N}$$

• Pixelation noise: $\langle (\Delta x)^2 \rangle = \frac{\sigma^2 + \frac{a^2}{12}}{N}$

• pixel size a, $\frac{a^2}{12}$ is the variance of a top-hat distribution of size a



Case (b)

• We minimized
$$\chi^2(x) = \sum \frac{(S_{ij} - G_{ij}(x))^2}{\psi_{ij}^2}$$

• 1-dimensional
$$\chi^2(x) = \sum \frac{(S_k - G_k(x))^2}{\psi_k^2}$$

- Background noise only $\psi_k = b \ (b \in R)$
- Taylor expansion $G_k(x) = G_k(x_0) + \underbrace{(x - x_0)}_{\Delta x} G'_k(x_0) + O((\Delta x)^2)$ • Denote $\Delta S_k = G_k(x_0) - S_k$



$$0 = \frac{d}{dx}\chi^{2} = \sum 2 \frac{(S_{k} - G_{k}(x))}{b^{2}}G'_{k}(x)$$

$$\Leftrightarrow 0 = \sum (-\Delta S_{k} - \Delta x G'_{k}(x_{0}))G'_{k}(x_{0}) + O((\Delta x)^{2})$$

$$\Leftrightarrow 0 = \sum \Delta S_{k}G'_{k}(x_{0}) + \sum G'_{k}(x_{0})^{2}\Delta x + O((\Delta x)^{2})$$

$$\Rightarrow \Delta x \approx -\frac{\sum \Delta S_{k}G'_{k}(x_{0})}{\sum G'_{k}(x_{0})^{2}}$$
Now as $G'_{k}(x_{0})$ are constants
$$\langle (\Delta S_{k})^{2} \rangle = Var(S_{k}) = b^{2}, \langle \Delta S_{k} \rangle = 0$$

$$\Rightarrow \langle (\sum \Delta S_{k}G'_{k}(x_{0}))^{2} \rangle = (\sum \langle (\Delta S_{k})^{2} \rangle G'_{k}(x_{0})^{2}) = \sum b^{2}G'_{k}(x_{0})^{2}$$

$$\Rightarrow \langle (\Delta x)^2 \rangle = \frac{D^2}{\sum G'_k(x_0)^2}$$



From
$$G_k(x) = \frac{N}{\sqrt{2\pi\sigma}} \exp\left(-\frac{(k-x)^2}{2\sigma}\right)$$
 we derive
 $G'_k(x) = \frac{N}{\sqrt{2\pi\sigma^2}} (k-x) \exp\left(-\frac{(k-x)^2}{2\sigma}\right)$
 $\Rightarrow G'_k(x)^2 = \frac{N^2}{2\pi\sigma^4} (k-x)^2 \exp\left(-\frac{(k-x)^2}{\sigma}\right)$

Finally we use

•
$$\sum G'_k(x_0)^2 \approx a \int G'_k(x_0)^2 dk =$$

 $\frac{aN^2}{2\pi\sigma^4} \int (k-x)^2 \exp\left(-\frac{(k-x)^2}{\sigma^2}\right) dk = \frac{aN^2}{4\sqrt{\pi}\sigma^3}$
to obtain

$$\langle (\Delta x)^2 \rangle = \frac{4b^2 \sqrt{\pi} \sigma^3}{aN^2}$$





4. Monte Carlo sample generation

Basic algorithm

- Fix some (arbitrary) spot center
- Generate random photon positions (Gaussian distributed)
- Collect photons on coarse grid
- Add (Poisson distributed) background for every pixel
- Parameters
 - Spot diameter, pixel size
 - Number of photons in spot
 - Background noise level

Dependence on the photon number N_t ≈ 1317 photons









- Underestimation due to
 - First order Taylor approximation
 - Interpolation between two limited cases
 - Transformation of a discrete sum into an integral
 - Assumption of $\frac{s}{a} \gg 1$
 - Simplification of Least Squares
- Error in N



5. Summary & Outlook

- Basic idea of achieving nanometer resolution in microscopy
- Construction of fitting Algorithms
- Simple, but rough estimates for the errors
- Application to experimental data, comparison with common software
- Check for potential improvements
- • •



THANK YOU FOR YOUR ATTENTION!



Source: xkcd.com/435/

