

Super-resolution microscopy using speckle illumination

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Outline

- 1 Image formation model
- 2 From SIM to blind-SIM
- 3 Marginal estimation

Outline

1 Image formation model

2 From SIM to blind-SIM

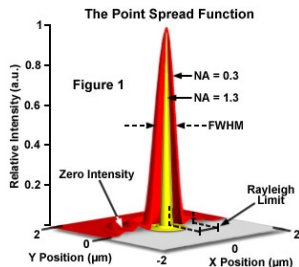
3 Marginal estimation

Image formation model

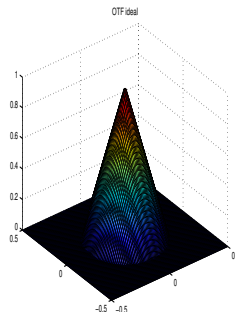
A diffraction limited system

Conventional microscopy is a diffraction limited system:

$$y = h \otimes (\rho \times I) + \epsilon$$



(a) Point spread function h



(b) \tilde{h} for circular pupil

Outline

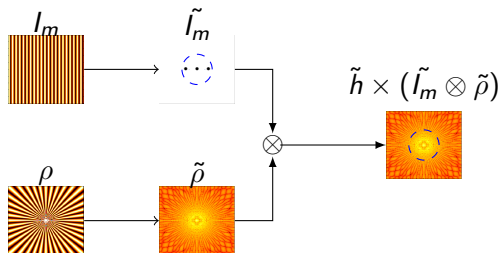
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Surpassing the resolution limit

Structured illumination microscope

The image formation model of SIM [2][4] is given by:

$$y = h \otimes (\rho \cdot I_{\text{structured}}) + \epsilon$$



From SIM to Blind-SIM

- Sensitive to the distortions of the Illumination
- Blind-SIM [3] is a alternative to SIM proposed in 2012. It replaces structured illumination by M random speckles I_1, \dots, I_M

$$\begin{cases} y_I = h \otimes (\rho \cdot I_1) + \epsilon_1 \\ \vdots \\ y_M = h \otimes (\rho \cdot I_M) + \epsilon_M \end{cases}$$

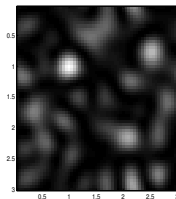


Figure : speckle illumination

Joint estimation in blind-SIM

The super resolution image is obtained by minimizing the **joint least square criterion**:

$$J(\rho, l_1, \dots, l_M) = \sum_{m=1}^M \|y_m - h \otimes (\rho \cdot l_m)\|^2$$

subject to:

$$\begin{cases} \rho \geq 0, & l_m \geq 0, & \forall m \\ \frac{1}{M} \sum_{m=1}^M l_m = l_0 \end{cases} \quad (1)$$

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Marginal approach

Maximum Likelihood estimation

- Good asymptotic properties
- Known super resolution capacity (2 times OTF support)[5]

$$y_m = h \otimes (\rho \cdot I_m) + \epsilon_m$$

in which:

$$I_m \sim \mathcal{N}(I_0, \Sigma_i)$$

$$\epsilon_m \sim \mathcal{N}(0, vI)$$

Marginal approach

Maximum Likelihood estimation

$$y_m \sim \mathcal{N}(h \otimes (\rho \cdot l_0), \Sigma_y)$$

in which:

$$\Sigma_y = HR\Sigma_iRH^t + vI$$

Then the **marginal (anti-log) likelihood** function is a function of ρ and does not depend on the illumination values

$$J(\rho) = - \sum_{m=1}^M \log(f(y_m|\rho))$$

Marginal approach

Patch Model

Dealing with covariance matrix Σ_y is difficult for real size images, hence we introduce a patch model:

$$y_{mp} = H_p R I_m \quad (2)$$

In this case, the anti-log likelihood function $J(\rho)$ becomes:

$$J(\rho) = \sum_p J_p(\rho) \quad (3)$$

Simulations

Parameters setting

- A star like object is used as the true sample: $\rho \propto (1 + \cos(40\theta))$
- The point spread function is given by the Bessel function

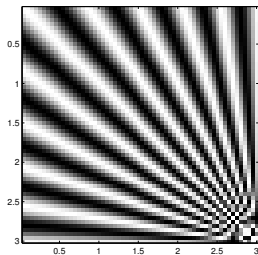
$$h(r, \theta) = \left(\frac{J_1(NAk_0r)}{k_0r} \right)^2 \frac{k_0^2}{\pi} \quad (4)$$

in which NA is the objective numerical aperture set to 1.49 and $k_0 = \frac{2\pi}{\lambda}$ is the free-space wavenumber with λ the emission and the excitation wavelengths. The sampling step is $\frac{\lambda}{20}$.

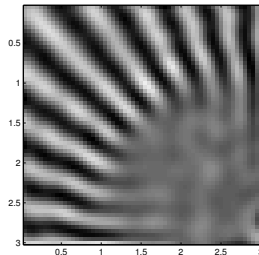
- Covariance speckle $\Sigma_I = H$
- Noise level 40dB

Simulations

Original object



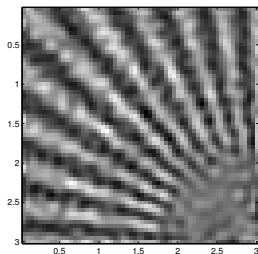
(a) Original object



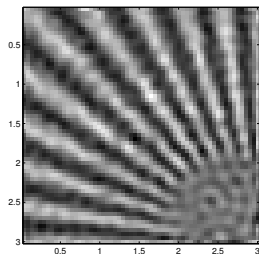
(b) Deconvolution of wide-field image

Simulations

Marginal estimator without patch



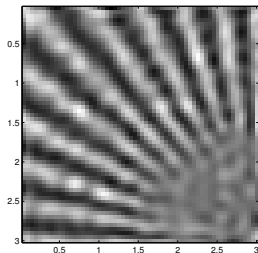
(a) 100 speckle patterns



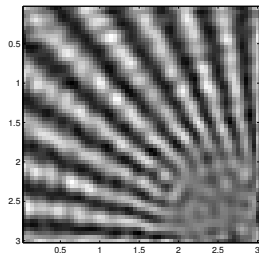
(b) 300 speckle patterns

Simulations

200 Illuminations



(a) Patch size 2x2



(b) Patch size 10x10

Conclusion

- Introduce the marginal approach to image reconstruction using speckle patterns
- Demonstrate the super-resolution capacity of marginal approach and its patch version
- Future work
 - A more efficient numerical optimization algorithm
 - 3D image reconstruction problem will be explored

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Idier Jérôme

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