

P. Pankajakshan, *et al.*

Wavefront sensing for aberration modeling in fluorescence MACROscopy

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Widefield fluorescence MACROscope

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Widefield fluorescence MACROscope





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Central points

- Low magnification objective lens is combined with apochromatic zoom lens,
- permits large object fields (up to 35mm) and large working distances (up to 97mm),



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Widefield fluorescence MACROscope

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Central points

- Low magnification objective lens is combined with apochromatic zoom lens,
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- parallax-free and precise imaging,



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Central points

- Low magnification objective lens is combined with apochromatic zoom lens,
- permits large object fields (up to 35mm) and large working distances (up to 97mm),
- parallax-free and precise imaging,
- multi-color fluorescence.



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Numerical aperture and working distance





Numerical aperture and working distance



Numerical aperture increases



Numerical aperture and working distance



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Best of two worlds

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Best of two worlds

Max	Maximum z	Maximum zoom po	Maximum zoom position
Max	Maximum z	Maximum zoom po	Maximum zoom position



Experimental impulse response





Experimental impulse response



Figure 1: 2.5μ m beads imaged using a Leica Widefield MacroFluoTMZ16 APO fit with $5\times$ objective and the $1.6\times$ zoom. ©Herbomel lab, Pasteur Institute.

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MACROscopes-Are they really the best of the two worlds?



Figure 2: Axial projection of the beads. ©Herbomel lab, Pasteur Institute.

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Forward problem: Characterizing the aberration



Incoherent scalar PSF model

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 If P(k_x, k_y, z) is the 2D complex pupil function and λ is the wavelength, the amplitude PSF can be calculated by just 2N_z number of 2D FFTs as

$$h_A(x, y, z; \lambda) = \int_{k_x} \int_{k_y} P(k_x, k_y, z) \exp(j(k_x x + k_y y)) dk_y dk_x$$



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► and the incoherent PSF is

$$h_{Th}(\mathbf{x}; \lambda_{\mathsf{ex}}, \lambda_{\mathsf{em}}) = C |h_A(\mathbf{x}; \lambda_{\mathsf{ex}})| \times |h_A(\mathbf{x}; \lambda_{\mathsf{em}})|$$

 \blacktriangleright $\lambda_{\rm ex}$ and $\lambda_{\rm em}$ are the excitation and emission peak wavelengths.



Limiting apertures overlapping





Limiting apertures overlapping





Optical vignetting

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Figure 3: Lens viewed from the front. (Photograph by Peter Boehmer.)



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Figure 3: Lens viewed from the front. (Photograph by Peter Boehmer.)



Figure 4: Lens viewed from the side. (Photograph by Peter Boehmer.)



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► For a MICROscope, the pupil function is



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► For a MICROscope, the pupil function is

$$P_m(k_x, k_y, z; \lambda) = \begin{cases} e^{jk_0\phi(\theta_i, \theta_s, z)}, & \text{ if } \sqrt{k_x^2 + k_y^2} < \frac{2\pi}{\lambda} \mathsf{NA}_{\mathsf{Obj}} \\ 0, & \text{ otherwise.} \end{cases}$$



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Modification for a MACROscope

$$P(k_x, k_y, z; \lambda) = \begin{cases} P_m, & \text{if } \sqrt{(k_x - r_x)^2 + (k_y - r_y)^2} < \frac{2\pi}{\lambda} \mathsf{NA}_{\mathsf{Zo}} \\ 0, & \text{otherwise.} \end{cases}$$



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► NA_{Obj} and NA_{Zo} are the objective and zoom lens NA; (r_x, r_y) are the relative displacements.

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Characterization in the lateral field

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Figure 5: We can characterize the behavior at any position in the lateral field.

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Out-of-focus highlights (OOFH)



Figure 6: Theoretically calculated MACROscope PSF in log scale. NA=0.5, lateral sampling 178.33nm, axial sampling 1000nm.

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Out-of-focus highlights (OOFH)



Figure 6: Theoretically calculated MACROscope PSF in log scale. NA=0.5, lateral sampling 178.33nm, axial sampling 1000nm.



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Inverse problem: Wavefront sensing from intensity data







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Summary

► The PSF of an imaging system can vary experimentally,



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- The PSF of an imaging system can vary experimentally,
- theoretically it can be calculated from the pupil function by a simple Fourier transform,



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- The PSF of an imaging system can vary experimentally,
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- the pupil function is not often available,



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- The PSF of an imaging system can vary experimentally,
- theoretically it can be calculated from the pupil function by a simple Fourier transform,
- the pupil function is not often available,
 - the wavefront can be sensed by using a Shack-Hartmann wavefront sensor,



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 - it can be also retrieved from the observed intensities,



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- the aberrations in the optics of the objective can be determined by studying this phase,



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 - the wavefront can be sensed by using a Shack-Hartmann wavefront sensor,
 - it can be also retrieved from the observed intensities,
- the aberrations in the optics of the objective can be determined by studying this phase,
- the estimated wavefront can be used to correct the aberrated optical path.



Wavefront sensing-a Bayesian interpretation

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For uncorrelated low photon count data the observation is:

$$i(\mathbf{x}) = \mathcal{P}\{|h_{\mathsf{A}}(\mathbf{x})|^2 + b(\mathbf{x})\}, \forall \mathbf{x} \in \Omega_s$$



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 Considering Poissonian photon counting statistics, the likelihood of obtaining image i(x) from a diffraction-limited point source:

$$\Pr(i|h_{\mathsf{A}}) = \prod_{\mathbf{x}\in\Omega_s} \frac{(h_{\mathsf{A}}+b)(\mathbf{x})^{i(\mathbf{x})}\exp(-(h_{\mathsf{A}}+b)(\mathbf{x})}{i(\mathbf{x})!}$$



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From the Bayes' theorem, the a posteriori is

$$\Pr(h_{\mathsf{A}}|i) = \frac{\Pr(i|h_{\mathsf{A}})\Pr(h_{\mathsf{A}})}{\Pr(i)}$$



Global idea for sensing

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Invariant to defocus



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Invariant to defocus

$$P(k_x, k_y, z) = \begin{cases} e^{jk_0(\varphi_{\mathsf{aberr}} + \phi_{\mathsf{defocus}}(\theta_i, z))}, & \text{ if } \sqrt{k_x^2 + k_y^2} < \frac{2\pi}{\lambda}\mathsf{N}\mathsf{A} \\ 0, & \text{ otherwise.} \end{cases}$$



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• Estimate the near-focus amplitude distribution, \hat{h}_A , by maximizing the *a posteriori* (MAP) or minimizing the cologarithm of the *a posteriori*

$$\hat{h}_{\mathsf{A}}(\mathbf{x};\varphi_{\mathsf{aberr}}) = \underset{h_{\mathsf{A}}(\mathbf{x})}{\operatorname{arg\,min}} - \log[\Pr(h_{\mathsf{A}}|i)], \mathsf{s. t. } k_{\mathsf{MAX}} < \frac{2\pi\mathsf{NA}}{\lambda_{\mathsf{ex}}}$$

 k_{MAX} is the pupil support,

this can be solved by using a fixed-point iterative algorithm.



Experiment on intensity data

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Figure 7: Radially projected 2.5μ m observed intensity volume. ©Imaging Center, IGBMC, France.



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Figure 7: Radially projected $2.5\mu m$ observed intensity volume. ©Imaging Center, IGBMC, France.

Figure 8: Axially projected 2.5μ m observed intensity volume. ©Imaging Center, IGBMC, France.



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Figure 9: OOFH radial section of the observed volume, $z = -57 \mu m$. ©Imaging Center, IGBMC, France.



Chopped defocus



Figure 9: OOFH radial section of the observed volume, $z = -57 \mu m$. ©Imaging Center, IGBMC, France.



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Figure 10: Retrieved unwrapped pupil phase from the intensity images $\tau = 0.9$ and the maximum number of iteration is 40.



Chopped pupil

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Figure 10: Retrieved unwrapped pupil phase from the intensity images $\tau = 0.9$ and the maximum number of iteration is 40.



Take home message



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 The MACROscope PSF varies as a function of the lateral field position,



Take home message



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- The MACROscope PSF varies as a function of the lateral field position,
- vignetting was observable only for smaller zooms (large FOV) and is negligible for larger zooms,



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- the sensed wavefront can be used to define the effective working zoom,
- ongoing work: restore the images by correcting for the field aberration and also the diffraction effects.



Acknowledgements

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 The first author wish to thank ANR DIAMOND for funding the postdoctoral research fellowship,



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Thank you!