

Improving Resolution in Wide-field Fluorescence Microscopy Using **Deconvolution Techniques** Charlene Cuellar¹, Nikhil Chacko², Dr. Michael Liebling²

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A. Motivation

- Fluorescence microscopy enables the study of specific regions of interest in live cell imaging.



C. Project Outline

1. Model the available microscope (Leica DMI 6000B) Determine the PSF characteristic to the microscope in the lab

2. Deconvolve data from single-view observation Use PSF to deblur 3-D data acquired by the microscope

3. Deconvolve data from multi-view observations

Acquire data from multiple angles and perform deblurring using a

C3. Multi-view Deconvolution

FIG .5: Two observations (xy-plane) of the zebrafish blurred along different directions. Since a single-view observation loses information along an axis, it can be compensated by imaging the object at different angles and then spatially aligning the images



multi-channel deconvolution algorithm.

C1. PSF Measurement

. PSF Measurement

- Fluorescent beads having a diameter less than the spatial resolution of the device approximate point sources.
- The blurred observation of any single bead hints to the PSF of the microscope.



FIG. 3: The xz-plane views of (a) the measured PSF, (b) the average of 10 measured PSFs with less noise content and (c) the Richards and Wolf PSF model generated.



- Though spatial registration of observations blurred at different angles is an unsolved problem, simulated results prove to be superior than single-view deconvolution results.



- Optical Sectioning

3D data stacks are obtained by acquiring multiple 2D images (xy) at different focal planes along the optical axis (z).

- Drawback of optical sectioning

Objects outside each focal plane being imaged also receive illumination and become fluorescent, undesirably interfering with the imaging of all focal planes. This manifests as blurring along the z-axis during 3D data reconstruction.

- Pursued solution

Use a computational technique (deconvolution) to reverse the blurring process.

B. Deconvolution: problem statement

- A linear shift-invariant system is completely characterized by its response to a point source.



- The measured PSF exhibit a close proximity to the theoretical models generated, except for their amplitude. - Their similarity advocates the use of theoretical PSF models, which are free of noise, for deconvolution algorithms.

C2. Single-view Deconvolution

200um

Any signal response

can be represented

as a linear

combination of

many PSFs.



(b) Deconvolution algorithm: Richardson-Lucy, **PSF:** Richards & Wolf model



xy-plane

xy-plane



FIG. 6: Four observations (yz-plane) of a mouse blood-brain barrier, blurred (simulated) along different directions (shown inlet is the PSF responsible for the respective blurs)





D. Conclusion and Future Perspective



 $f(x) = \int f(\xi) \cdot \delta(x - \xi) d\xi \quad g(x) = \int f(\xi) \cdot h(x - \xi) d\xi$

- Fourier Transforms simplify convolutions to multiplications

$$F(\omega) \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} f(\xi) \cdot e^{-j\omega x} dx \qquad G(\omega) = F(\omega) \cdot H(\omega)$$

- Given g(x) [blurred output] and h(x) [PSF], find f(x) [original] object]: **deconvolution** problem



(c) Deconvolution all gorithm: Thresholded Landweber, PSF: Richards & Wolf model



The deconvolution algorithm reduces the blur both in the lateral planes (xy) and the axial direction (z) in our model system, the zebrafish.

- The theoretical PSF models generated, according to the instrument and experimental parameters, were close approximations to the actual measured data and produced less noisier deconvolved results

- The faithful registration of actual 3D datasets blurred along different angles still remains a problem that is unsolved, though the simulation results for the multi-angle deconvolution look promising.

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