# Three-dimensional structured illumination microscopy using Lukosz bound apodization reduces pixel negativity at no resolution cost

Christiaan H. Righolt, 1,2 Sabine Mai, Lucas J. van Vliet, 1 and Sjoerd Stallinga 1,\*

<sup>1</sup>Department of Imaging Physics, Delft University of Technology, Lorentzweg 1, 2628 CJ Delft, Netherlands <sup>2</sup>Manitoba Institute of Cell Biology, University of Manitoba, Cancer Care Manitoba, 675 McDermot Avenue, Winnipeg MB, R3E 0V9 Canada

\*s.stallinga@tudelft.nl

**Abstract:** The quality of the reconstructed image in structured illumination microscopy (SIM) depends on various aspects of the image filtering process. To optimize the trade-off between resolution and ringing artifacts, which lead to negative intensities, we extend Lukosz-bound filtering to 3D SIM and derive the parametrization of the 3D SIM cut-off. We compare the use of the Lukosz-bound as apodization filter to triangular apodization and find a tenfold reduction in the most negative pixel value with a minimal resolution loss. We test this algorithm on experimental SIM images of tubulin filaments and DAPI stained DNA structure in cancer cells and find a substantial reduction in the most negative pixel value and the percentage of pixels with a negative value. This means that there is no longer a need to clip the final image to avoid these negative pixel values.

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### 1. Introduction

Structured illumination microscopy (SIM) extends the resolution of fluorescence microscopy with a factor of up to two beyond the diffraction limit. It was first described theoretically in the 1960s [1] and demonstrated as both an optical sectioning and a superresolution method in the 1990s and 2000s [2–7]. Recent years have seen various improvements in the technology [8–15]. The standard way to reconstruct a SIM image has been through the inversion of the linear combination of frequency bands, followed by a Wiener style multi-band filter. In this Wiener style filter, the inverse signal-to-noise ratio has been replaced by a single regularization parameter. An apodization filter is then used as a final post-processing step, in order to suppress negative pixel value and ringing artifacts [11, 16–18]. It turns out, however, that apodization filters used in practice often suffer from severe negative pixel value artifacts, which are typically solved by clipping the negative pixel values to zero. This practice leads to unreliable final images for laymen users, and ultimately to biased biological results. Recently we described a method for SIM reconstruction in 2D [19] with the aim to reduce the number of free parameters and to ensure a non-negative Point Spread Function (PSF) by using the upper bound to the Modulation Transfer Function (MTF) as derived by Lukosz [20,21] as apodization function.

In this paper we focus on the extension of the Lukosz bound apodization method to 3D-SIM. This method cannot be trivially extended from 2D to 3D due to the severe anistropy in

the MTF of the microscope. In particular we will introduce, for the first time, the (numerical) parameterization of the anisotropic 3D cut-off of the SIM OTF, the generalization of the Lukosz bound to 3 (or any higher number of) dimensions, a method to compute the Lukosz bound for any arbitrary frequency cut-off surface with two possible ways to deal with anisotropic transfer functions, and the unification of all reconstruction steps in a generalized Tikhonov-Miller filter. This transforms SIM reconstruction from a patchwork of individual, unconnected steps to a unified, tunable optimization problem for quantitatively reliable super-resolution microscopy.

The content of the paper is as follows. In section 2 we will shortly summarize the main reconstruction steps of [19], derive the parametrization of the 3D SIM cut-off, give an extension of the Lukosz bound to 3D, and compare 3D Lukosz bound and triangular apodization. Experimental validation will be presented in section 3, were the methods developed throughout the paper will be applied to images of actin filaments and DAPI-stained DNA structure in a cell. The paper will conclude with a discussion in section 4.

# 2. Theory

# 2.1. Overview of 3D SIM reconstruction

The spatial coordinates  $(x,y,z)^T$ , with coordinate z along the optical axis, are normalized to  $\mathbf{u} = (xn_{\text{med}}\sin\alpha/\lambda, yn_{\text{med}}\sin\alpha/\lambda, z4n_{\text{med}}\sin^2(\alpha/2)/\lambda))^T$ , with  $n_{\text{med}}$  the refractive index of the medium in object space,  $\lambda$  the wavelength of the emitted light and  $\alpha$  the marginal ray angle in object space. The spatial frequencies are denoted by  $\mathbf{v}$ , for which we use the Fourier transform convention:

$$\hat{H}(\mathbf{v}) = \int H(\mathbf{u}) \exp(-2\pi i \mathbf{u} \cdot \mathbf{v}) d\mathbf{u}, \tag{1}$$

in which  $H(\mathbf{u})$  is the PSF and  $\hat{H}(\mathbf{v})$  the optical transfer function (OTF). The illumination function  $W(\mathbf{u})$  is periodic and its spectrum consists of delta-peaks:

$$\hat{W}(\mathbf{v}) = \sum_{m} w_{m} \delta(\mathbf{v} - \mathbf{q}_{m}), \qquad (2)$$

where  $w_m$  denotes the relative weight of the  $m^{th}$  frequency component. In 3D SIM, the zeroth and first orders of a diffraction grating are projected on the back focal plane of the objective. The interference pattern of the three resulting plane waves in object space leads to a periodic illumination pattern with a spectrum that has seven peaks,  $\mathbf{q}_m \in \{(0,0,0)^T, (\pm 2q_r,0,0)^T, (\pm q_r,0,\pm q_z)^T\}$ , where  $q_r = 1/p$  and

$$q_z = \frac{1}{4\sin^2(\alpha/2)} \left( 1 - \sqrt{1 - \frac{\sin^2 \alpha}{p^2}} \right)$$
 (3)

with p the normalized grating period. Images are recorded for a set of rotations  $\mathbf{R}_l$ ,  $(l = 1, 2, ..., N_r)$  and translations  $\mathbf{u}_n = n(p/N_t, 0)$   $(n = 1, 2, ..., N_t)$ . For 3D SIM typically  $N_r \in \{3, 5\}$  rotations and  $N_t = 5$  translations are used. In the remainder of the paper we will use  $N_r = 3$ .

The images are linear combinations of shifted band-passed images. These band-passed images can be recovered by an inversion algorithm [22–24]:

$$\hat{I}_{\text{bands},lm}(\mathbf{v}) = w_m \hat{H}(\mathbf{v}) \hat{T}(\mathbf{v} - \mathbf{R}_l \cdot \mathbf{q}_m), \qquad (4)$$

where the object spectrum is denoted by  $\hat{T}(\mathbf{v})$ . A final reconstructed image can now be obtained by moving each band in frequency space and filtering it with kernel  $\hat{F}_{lm}(\mathbf{v})$ :

$$\hat{I}_{gen}(\mathbf{v}) = \sum_{lm} s_m \hat{F}_{lm}(\mathbf{v}) \hat{I}_{band,lm}(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m), \qquad (5)$$

where  $s_m$  is an additional weight factor. The generalized Tikhonov-Miller functional to be minimized in SIM reconstruction [19] is given by:

$$\mathcal{L}_{gen} = \int \left[ \sum_{lm} \hat{g}(\mathbf{v}_{lm}) \left| \hat{B}_{lm}(\mathbf{v}) \hat{I}_{gen}(\mathbf{v}) - s_m \hat{C}(\mathbf{v}) \hat{I}_{band,lm}(\mathbf{v}_{lm}) \right|^2 + \kappa \left| \hat{A}(\mathbf{v}) \hat{I}_{gen}(\mathbf{v}) \right|^2 \right] d\mathbf{v}. \quad (6)$$

In Eq. (6),  $\mathbf{v}_{lm} = \mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m$  is the band-shifted spatial frequency vector,  $\hat{B}_{lm}(\mathbf{v}) = s_m w_m \hat{H}(\mathbf{v}_{lm})$  denotes the ideal or desired response of the imaging system,  $\kappa$  the regularization parameter,  $\hat{A}(\mathbf{v})$  the regularization weight function,  $\hat{C}(\mathbf{v})$  the apodization function, and  $\hat{g}(\mathbf{v})$  a data misfit weight function, that can be used to suppress the hexagonal imprint artifact that is typical for SIM [23]. A Gaussian data misfit function can be chosen:

$$\hat{g}(\mathbf{v}) = 1 - \alpha \exp\left[-\frac{v_x^2 + v_y^2}{2\sigma_l^2} - \frac{v_z^2}{2\sigma_a^2}\right],\tag{7}$$

where  $\alpha$  is the amplitude,  $\sigma_l$  the lateral frequency width and  $\sigma_a$  the axial frequency width. The functional is minimized with respect to the reconstructed image  $\hat{I}_{gen}(\mathbf{v})$ , leading to the filter kernels:

$$\hat{F}_{lm}(\mathbf{v}) = \frac{\hat{C}(\mathbf{v})\hat{g}(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m)\hat{B}_{lm}(\mathbf{v})^*}{\kappa \hat{A}(\mathbf{v})^2 + \sum_{lm} \hat{g}(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m) \left|\hat{B}_{lm}(\mathbf{v})\right|^2}$$
(8)

and the effective OTF for the reconstructed image:

$$\hat{H}_{gen}(\mathbf{v}) = \frac{\hat{C}(\mathbf{v}) \sum_{lm} s_m w_m \hat{g}(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m) \hat{B}_{lm}(\mathbf{v})^* \hat{H}(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m)}{\kappa \hat{A}(\mathbf{v})^2 + \sum_{lm} \hat{g}(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m) \left| \hat{B}_{lm}(\mathbf{v}) \right|^2}.$$
(9)

# 2.2. The 3D SIM cut-off

In order to compute the Lukosz-bound for 3D SIM, the anisotropic SIM cut-off needs to be derived first. This cut-off will be derived in spherical coordinates  $(\rho, \theta, \phi)$ , such that  $v_x = \rho \cos \phi \sin \theta$ ,  $v_y = \rho \sin \phi \sin \theta$  and  $v_z = \rho \cos \theta$ . The SIM OTF of Eq. (9) is non-zero when  $\sum_{lm} \hat{H}(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m)$  is non-zero. The cut-off of the SIM OTF is, therefore, given by the outer boundary segments of the incoherent OTFs moved to positions  $\{\mathbf{R}_l \cdot \mathbf{q}_m\}$ . The cut-off of the incoherent OTF has been derived by Sheppard [25]. The cross-section of the incoherent OTF with a meridional plane has a two-lobed shape, each lobe consisting of two elliptical arcs. The SIM OTF cut-off  $\rho(\theta, \phi)$ , is thus given by the intersection of the radial line in direction  $(\theta, \phi)$  and one of the moved elliptical arcs of the incoherent OTF. Using the translation  $\mathbf{q}_m = (q_{mx}, q_{my}, q_{mz})$ , see Eq. (3), and using our coordinate normalization, we then find:

$$\sin^{2}\alpha \left(\sqrt{(v_{x}-q_{m,x})^{2}+(v_{y}-q_{m,y})^{2}}-1\right)^{2}+16\sin^{4}\left(\frac{\alpha}{2}\right)\left(v_{z}-q_{m,z}+\frac{\cos\alpha}{4\sin^{2}(\alpha/2)}\right)^{2}=1.$$
(10)

Substituting the lateral coordinates in Eq. (10) with  $v_x = v_z \cos \phi \tan \theta$  and  $v_y = v_z \sin \phi \tan \theta$  and squaring this equation leads to a quartic equation in  $v_z$  after several rearrangements:

$$a_4 v_z^4 + a_3 v_z^3 + a_2 v_z^2 + a_1 v_z + a_0 = 0. (11)$$

The polynomial coefficients are:

$$a_0 = Q_C^2 - 4q_{m,x}^2 - 4q_{m,y}^2, (12)$$

$$a_1 = 4Q_B Q_C + 8q_{m,x} \cos \phi \tan \theta + 8q_{m,y} \sin \phi \tan \theta, \qquad (13)$$

$$a_2 = -2Q_A Q_C + 4Q_B^2 - 4\tan^2\theta, (14)$$

$$a_3 = -4Q_A Q_B, (15)$$

$$a_4 = Q_A^2, (16)$$

with the constants:

$$Q_A = 4\tan^2\left(\frac{\alpha}{2}\right) + \tan^2\theta, \tag{17}$$

$$Q_B = q_{m,x}\cos\phi\tan\theta + q_{m,y}\sin\phi\tan\theta + (1 + 4q_{m,z})\tan^2\left(\frac{\alpha}{2}\right) - 1, \tag{18}$$

$$Q_C = \frac{1}{\sin^2 \alpha} - \left(\frac{1}{\tan \alpha} - 2q_{m,z} \tan\left(\frac{\alpha}{2}\right)\right)^2 - q_{m,x}^2 - q_{m,y}^2 - 1.$$
 (19)

Equation (11) has four solutions  $v_{z,m,j}(\theta,\phi)$  for each peak in the illumination pattern. The *z*-solutions can be transformed to spherical coordinates leading to:

$$\rho_{m,j}(\theta,\phi) = \nu_{z,m,j}(\theta,\phi)\sqrt{1 + \tan^2 \theta}.$$
 (20)

The *j*th solution represents an actual intersection when it is positive. Negative solutions match positive solutions for a different *m* because of symmetry in the system, and imaginary solutions occur when there is no intersection. The maximum of all solutions, excluding imaginary and negative numbers yields the SIM cut-off:

$$\rho\left(\theta,\phi\right) = \max_{m,j} \left\{ \rho_{m,j}\left(\theta,\phi\right) > 0 \right\}. \tag{21}$$

The above derivation is valid, except when  $\theta \in \{0, \pi/2, \pi\}$ . For  $\theta = \pi/2$ ,  $v_z = 0$ ; for standard 3D SIM this leads to:

$$\rho\left(\frac{\pi}{2},\phi\right) = 2q_r\cos\phi_r + 2\sqrt{1 - q_r^2\sin^2\phi_r}.$$
 (22)

For  $\theta \in \{0, \pi\}$ ,  $v_x = v_y = 0$  and Eq. (10) can be solved directly as

$$\rho(0,\phi) = \rho(\pi,\phi) = q_z + \frac{\sqrt{1 - (q_r - 1)^2 \sin^2 \alpha} - \cos \alpha}{4 \sin^2 (\alpha/2)}.$$
 (23)

The resulting parametrization of the SIM cut-off is shown in Fig. 1.

# 2.3. The 3D SIM Lukosz-bound

Lukosz presented several conditions a transfer function should satisfy for its corresponding PSF to be non-negative in 1D and 2D [20,21]. Note that these conditions are necessary but not sufficient. Previously we discussed how this set of rules can be used to derive the Lukosz-bound for 2D SIM in order to get a non-negative SIM PSF [19].

The 1D bound is given by connecting the tips of Lukosz' "stair case":

$$\hat{\Lambda}_{1}(v, q_{c}) = \cos\left(\frac{\pi |v|}{|v| + q_{c}}\right) \text{ for } |v| \leq q_{c} \text{ and } \hat{\Lambda}_{1}(v) = 0 \text{ elsewhere}, \tag{24}$$

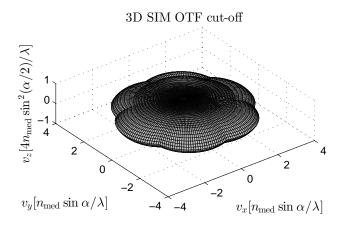


Fig. 1. Plot of the frequency cut-off in normalized frequency coordinates for standard 3D SIM in which the pattern is created by the zeroth and first order diffraction orders with three equally spaced rotation angles for  $q_r = 0.7$  and  $q_z = 0.2$ .

where  $q_c$  is the cut-off frequency. For the 2D case, the product of two 1D bounds along two orthogonal directions has to be taken. The actual bound is the minimum over all possible orientations of the 2D coordinate system. This formalism can be extended into 3D:

$$\hat{\Lambda}_{3}(\mathbf{v}) = \min \left\{ \prod_{j=1}^{3} \hat{\Lambda}_{1} \left( v'_{j}, q_{c,j}(\mathscr{R}) \right) | \mathscr{R} \in \mathscr{G} \right\}, \tag{25}$$

where the minimum is taken over a set  $\mathscr{G}$  of rotations  $\mathscr{R}$  of the 3D coordinate frame, where  $\mathbf{v}'$  are the spatial frequencies in the rotated frame and  $q_{c,j}(\mathscr{R})$  are the cut-offs along the rotated cardinal directions. The axial and lateral directions are fundamentally different, which requires care in implementing this recipe. We have devised two implementations that work well in practice. In the first, only rotations around the optical axis are considered; in the second, the full group of 3D rotations, SO(3), is considered in combination with an anisotropic pre-strechting of the spatial frequency space.

The first option of dealing with the anisotropy between the lateral and axial directions is to ensure that one of the cardinal directions remains aligned with the optical axis. This leaves one degree of freedom, the rotation  $\phi$  around the  $v_z$ -axis. The coordinate system is now described by orthogonal axes  $(\cos\phi,\sin\phi,0), (-\sin\phi,\cos\phi,0)$  and (0,0,1). A spatial frequency vector  $\mathbf{v}=(v\sin\xi\cos\psi,v\sin\xi\sin\psi,v\cos\xi)$  will have coordinates  $\mathbf{v}'=(v\sin\xi\cos(\psi-\phi),v\sin\xi\sin(\psi-\phi),\cos\xi)$  in the rotated frame. The minimum for all rotations for this bound is then given by:

$$\hat{\Lambda}_{3,\phi}\left(\mathbf{v}\right) = \min\left\{\hat{\Lambda}_{1}\left(v_{x}^{\prime},\rho\left(\frac{\pi}{2},\phi\right)\right)\hat{\Lambda}_{1}\left(v_{y}^{\prime},\rho\left(\frac{\pi}{2},\phi+\frac{\pi}{2}\right)\right)\hat{\Lambda}_{1}\left(v_{z}^{\prime},\rho\left(0,\phi\right)\right)|\phi\in\left[0,2\pi\right)\right\},\tag{26}$$

where the respective cut-offs are described by Eqs. (22) and (23). The 3D bound should still adhere to the 1D bound in the direction  $(\xi, \psi)$ . The 3D Lukosz-bound for rotations around the

optical axis,  $\hat{\Lambda}_{OAR}(\mathbf{v})$ , therefore becomes:

$$\hat{\Lambda}_{OAR}(\mathbf{v}) = \min \left\{ \hat{\Lambda}_{3,\phi}(\mathbf{v}), \hat{\Lambda}_{1}(\nu, \rho(\xi, \psi)) \right\}. \tag{27}$$

Note that  $\hat{\Lambda}_{OAR}(v_x, v_y, 0)$  is the 2D bound we described previously in [19]. The second option is to use all possible 3D rotations:

$$\hat{\Lambda}_{FR}(\mathbf{v}) = \min \left\{ \prod_{j=1}^{3} \hat{\Lambda}_{1} \left( v'_{j}, q_{c,j}(\mathscr{R}) \right) | \mathscr{R} \in SO(3) \right\}.$$
 (28)

Straightforward application of this recipe leads to a bound that is too pessimistic (too low) because of the severe anisotropy of the underlying spatial frequency cut-off. It appears that this can largely be solved by first applying an anisotropic stretch such that the bounding box of the cut-off surface has a cubic shape. To achieve this we rescale the  $v_r$ -axis by a factor:

$$\zeta = \frac{q_{\text{max},r}}{q_{\text{max},z}} = \frac{q_r + 2}{q_z + 1/2}.$$
 (29)

Note that this scale factor can be changed accordingly when a different spatial normalization is chosen; 2 and 1/2 are the lateral and axial frequency cut-offs of the lens OTF for our normalization. This scaling leads to several changed coordinates, marked with a tilde:

$$\tilde{v}_z = \zeta v_z, \quad \tilde{v} = \sqrt{v_x^2 + v_y^2 + \zeta^2 v_z^2} \quad \text{and} \quad \tilde{\theta} = \arccos\left(\frac{\zeta v_z}{\sqrt{v_x^2 + v_y^2 + \zeta^2 v_z^2}}\right).$$
 (30)

The cut-off can also be obtained in the rescaled coordinate system:

$$\tilde{\rho}\left(\tilde{\theta},\phi\right) = \rho\left(\theta,\phi\right)\sqrt{\sin^2\theta + \zeta^2\cos^2\theta},\tag{31}$$

All of the cardinal direction cut-offs  $q_{c,j}(\mathcal{R})$  and spatial frequency components  $v'_j$  in the rotated frame can be calculated using a parametrization of the rotation matrices  $\mathcal{R}$  with the three Euler angles, and can then be plugged into Eq. (28). This then yields  $\hat{\Lambda}_{FR}(\mathbf{v})$ , the 3D Lukosz-bound based on the full set of SO(3) rotations.

# 2.4. Comparison of apodization functions

The two options presented for the 3D SIM Lukosz-bound can be used as apodization function. A triangular apodization is typically used in practice, which is commonly created using a distance transform [11, 23]. In this case the apodization function becomes:

$$\hat{C}_{DT}(\mathbf{v}) = \frac{\mathscr{D}\left\{\sum_{lm}\hat{H}\left(\mathbf{v} + \mathbf{R}_{l} \cdot \mathbf{q}_{m}\right) > 0\right\}}{\max \mathscr{D}\left\{\sum_{lm}\hat{H}\left(\mathbf{v} + \mathbf{R}_{l} \cdot \mathbf{q}_{m}\right) > 0\right\}},$$
(32)

in which  $\mathscr{D}$  denotes the distance transform and  $\sum_{lm} \hat{H}\left(\mathbf{v} + \mathbf{R}_l \cdot \mathbf{q}_m\right) > 0$  is the footprint of the SIM OTF. The triangle can alternatively be formed over every line originating in the origin to the cut-off in that direction. This leads to an apodization function:

$$\hat{C}_{line}(\mathbf{v}) = 1 - \frac{v}{\rho(\theta, \phi)} \text{ for } v \le \rho(\theta, \phi) \text{ and } \hat{C}_{line}(\mathbf{v}) = 0 \text{ elsewhere.}$$
 (33)

We calculated these four apodization functions, the two triangle-based functions and the two Lukosz-bound functions, in the frequency domain for a  $256 \times 256 \times 128$  grid using Matlab

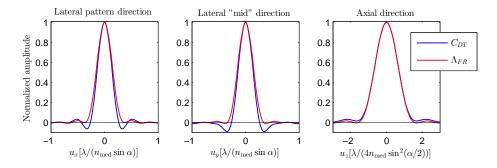


Fig. 2. Cross-sectional plots of the PSF's corresponding to  $C_{DT}(\mathbf{u})$  in blue and  $\Lambda_{FR}(\mathbf{u})$  in red for 3D-SIM with three rotation angles. The PSF's are the inverse Fourier transforms of  $\hat{C}_{DT}(\mathbf{v})$  and  $\hat{\Lambda}_{FR}(\mathbf{v})$ . The left plot depicts the line through the origin in the lateral  $v_z = 0$  plane in one of the pattern directions of  $\mathbf{R}_l$ . The middle plot shows a similar line in the  $v_z = 0$  plane in the direction midway between the directions of  $\mathbf{R}_l$  and  $\mathbf{R}_{l+1}$ . The right plot shows the values of the PSF's along the optical axis.

(MathWorks, Natick, USA). We chose the pixel size such that the footprint of the SIM OTF fitted exactly. We used  $q_r = 0.65$  and  $q_z = 0.16$  for three equidistant rotation angles. The corresponding PSF's, the inverse Fourier transforms of the respective functions were computed using the chirp z-transform algorithm [26]. The relative minimum values of these PSF's are summarized in Table 1 together with the full width at half maximum (FWHM) in the cardinal directions. Lines through the PSF's are plotted for the distance transform and SO(3)-rotated Lukosz-bound in Fig. 2.

Table 1. Comparison of the PSF's corresponding to the four different apodization options for 3D-SIM with three rotation angles. In addition to the absolute values of the measures, a normalized measure with respect to the value for  $C_{DT}(\mathbf{u})$  is included as well. The relative minimum value of the PSF (compared to the maximum) is listed as  $\min(\text{PSF})/\max(\text{PSF})$ . The lateral and axial full width at half maximum (FWHM lat. and FWHM ax.) are tabulated in the last four columns. The lateral anisotropy leads to a lateral FWHM difference between the  $\mathbf{R}_l$  and  $(\mathbf{R}_l + \mathbf{R}_{l+1})/2$  directions of a half percent or less in all four cases. Note that the lateral and axial coordinates are normalized differently.

	$\frac{\min(PSF)}{\max(PSF)}$	$\div C_{DT}$	FWHM lat.	$\div C_{DT}$	FWHM ax.	$\div C_{DT}$
$C_{DT}\left(\mathbf{u}\right)$	-0.0905	1	0.286	1	1.44	1
$C_{line}\left(\mathbf{u}\right)$	-0.0483	0.534	0.308	1.077	1.35	0.935
$\Lambda_{OAR}\left(\mathbf{u}\right)$	-0.0158	0.175	0.321	1.120	1.53	1.062
$\Lambda_{FR}\left(\mathbf{u}\right)$	-0.0109	0.120	0.320	1.118	1.44	0.997

The traditional apodization filter,  $C_{DT}(\mathbf{u})$ , yields negative values of -9% of the value at the origin. Negative PSF values lead to ringing artifacts in the reconstructed SIM image. The minimum occurs around  $u_x \approx u_y \approx 0.25$ , which causes these artifacts to be at *the* length scale of interest for this type of microscopy. Using a different method to create the triangle, namely by parametrizing the SIM cut-off, and drawing a triangle in each direction leads to a twofold reduction of the negativity. An order of magnitude improvement can, however, be gained by switching to the 3D SIM Lukosz-bound with the full SO(3) rotation group. This comes at the price of a small (around  $\approx 10\%$ ) increase in the lateral FWHM. The axial FWHM is hardly affected. Note that, as for the 2D case, these 3D Lukosz-bounds have a small residual amount

of negativity, because the conditions on which they are based are necessary conditions, not sufficient conditions.

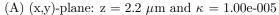
All performance measures are a bit worse for the Lukosz-bound implementation with only rotations around the optical axis. The computational load of finding the Lukosz-bound for only one rotational degree of freedom is, however, significantly lower. The calculation of  $\hat{\Lambda}_{FR}(\mathbf{v})$  is computationally intensive because it needs to be performed in a 6D-space (three spatial frequency components, three Euler angles parametrizing all rotations). Fortunately, this function does not change from image to image when the imaging system parameters, in particular the pitches and rotation angles of the illumination pattern, remain constant.

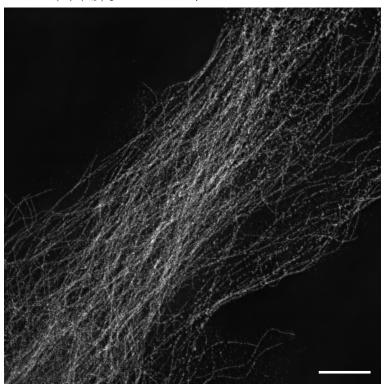
# 3. Experiments

In order to investigate the performance of the 3D-Lukosz bound as apodization function we recorded raw SIM image series of two test samples. The first sample is of tubulin from bovine pulmonary artery endothelial (BPEA) cells on a Molecular Probes test slide (F14781), the second sample contains the DAPI (4,6-diamidino-2-phenylindole)-stained DNA structure of a fixed, complex, multinucleated Reed-Sternberg cell from the Hodgkin's lymphoma HDLM-2 cell line [27]. Images were acquired using a Zeiss Elyra SIM microscope with a 63x/1.40 NA Plan-Apochromat objective lens and  $n_{\rm med} = 1.518$  immersion oil equipped with an Andor EMCCD iXon 885 camera with 8  $\mu$ m pixel size. The back-projected pixel size was 79 nm, because a 1.6x tube lens was used as well. The tubulin image was acquired using 488 nm laser excitation, a projected grating period of 444 nm and an emission passband of 495–575 nm. The DAPI image was acquired using 405 nm laser excitation, a projected grating period of 365 nm and a 420–480 nm passband for the emitted light.

Image reconstruction of the 3D image was performed in two steps. First, we extracted the spatial frequency bands using the methods developed by Wicker and Heintzmann [22–24]. Second, we obtained the final image using the filters described in section 2.1 with the 3D Lukosz-bound based on SO(3) rotations as apodization function  $\hat{C}(\mathbf{v})$ . We used conventional total energy regularization,  $\hat{A}(\mathbf{v}) = 1$  and the same amplitude  $\alpha = 0.95$  for the Gaussian data misfit function as for the 2D reconstruction. We fit a Gaussian to the lens PSF, Fourier transformed this Gaussian and took half the widths in Fourier space as the widths for the data misfit function:  $\sigma_l = 0.9$  and  $\sigma_a = 0.3$ . The images were reconstructed for a range of regularization parameters  $\kappa$  without boosting the side-bands, i.e.  $s_m = 1$  for all m. A slice of the 3D volume depicting tubulin filaments is shown for  $\kappa = 1 \times 10^{-5}$  in Fig. 3. A slice depicting the DNA structure in the multinucleated cell for  $\kappa = 1 \times 10^{-3}$  is shown in Fig. 4. We choose these values because they visually led to the best trade-off between noise suppression and image sharpness and contrast. The influence of the regularization parameter on these slices is shown in Media 1 and Media 3 for both images. A z-stack movie, going through the slices along the optical axis is visible in Media 2 and Media 4 for both samples. Note that no clipping of negative pixel values has been applied to any of the reconstructions.

We compared these images with the reconstruction obtained by Zeiss' ZEN software (release version 8.0) using triangular apodization, also without clipping negative pixel values to zero. The regularization parameter was set to the same values,  $10^{-5}$  and  $10^{-3}$  respectively, as for Lukosz-bound apodization. The ratio of the minimum to the maximum pixel value and the percentage of negative pixels were determined for both reconstructions as follows. First, we estimated the zero-level in the ZEN reconstructions from the background, because the ZEN software maps the images to 16-bit files. The foreground object was subsequently found by a coarse segmentation – an isodata threshold followed by a binary dilation – for the same slices as shown in Figs. 3 and 4. We took the minimum as the 1st and the maximum as the 99th percentile over the region of interest (ROI) thus found in order to be independent of outliers.



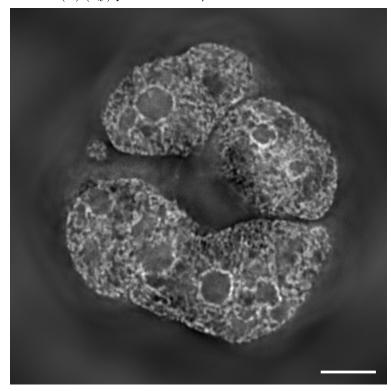


(B) (x,z)-plane: y = 20.1  $\mu$ m and  $\kappa$  = 1.00e-005



Fig. 3. 3D SIM image of tubulin filaments from a BPEA cell sample reconstructed using the full 3D SIM Lukosz-bound as apodization function. The intensity of the reconstructed image is linearly stretched, without clipping, between gray values 0 and 255 for display purposes. Shown is the result for a lateral *z*-slice (A) and an axial *y*-slice (B) for regularization parameter  $\kappa=10^{-5}$ . The lateral slice is shown for a range of values for the regularization parameters in Media 1 and for all *z*-positions in Media 2. The scale bar is 5  $\mu$ m.

(A) (x,y)-plane:  $z = 3.5 \mu m$  and  $\kappa = 1.00e-003$ 



(B) (x,z)-plane: y = 18.9  $\mu m$  and  $\kappa$  = 1.00e-003

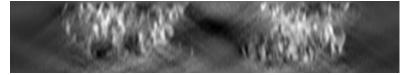


Fig. 4. 3D SIM image of the DNA structure (DAPI) in a multinucleated Reed-Sternberg cell reconstructed using the 3D SIM Lukosz-bound as apodization function. The intensity of the reconstructed image is linearly stretched, without clipping, between gray values 0 and 255 for display purposes. Shown is the result for a lateral z-slice (A) and an axial y-slice (B) for regularization parameter  $\kappa = \times 10^{-3}$ . The lateral slice is shown for a range of values for the regularization parameters in Media 3 and for all z-positions in Media 4. The scale bar is 5  $\mu$ m.

We also estimated the number of pixels in the ROI with value below the background level. The distribution of normalized pixel values over the ROI's are plotted in Fig. 5.

For the tubulin sample, the ZEN reconstruction led to a minimum pixel value of -7.8% of the maximum and 20% of the pixels in the ROI had a negative value; our reconstruction (Fig. 3) led to a minimum pixel value of -1.8% of the maximum and 3.9% negative pixels in the ROI. For the DAPI image, the ZEN reconstruction led to a minimum pixel value of -46% of the maximum with 42% negative pixels in the ROI. Our reconstruction (Fig. 4) had a minimum value of -18% the maximum; 9.3% of the pixels were negative in the ROI. Note that the depths of the minima do not solely depend on the choice of apodization function. The data misfit function, regularization term, noise properties and frequency content of the sample play a role as well. The experiments confirm our simulation results and show that using the Lukosz-bound as apodization function substantially reduces the negativity in the reconstructed image. For our images this led to a two to fivefold reduction in negativity. The sharpness of the Lukosz-bound reconstruction is visually the same as the sharpness of the ZEN reconstruction (data not shown). The reduction of negativity does, however, improve image contrast when seen on a stretched linear scale.

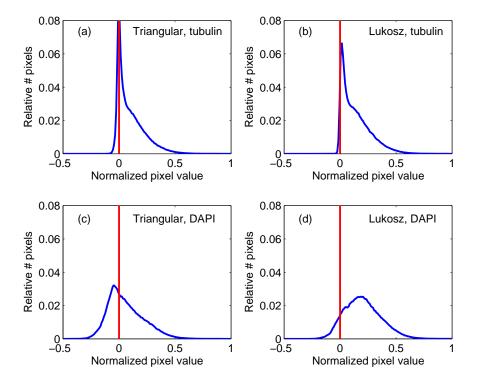


Fig. 5. Probability density function (pdf) of the normalized pixel values in reconstructed 3D SIM images. For each image, the object pixels are normalized by linearly scaling the background level to 0 and maximum pixel value to 1. The blue lines denote the pdf's, the bin-widths are 0.01, the red lines indicate the zero-intensity levels. This is done for: the ZEN reconstruction with triangular apodization of the tubulin sample (a); our reconstruction with Lukosz-bound apodization of the tubulin sample (b); the triangular apodized ZEN reconstruction of the DAPI cell sample (c); and our Lukosz-bound apodization in the reconstruction of the DAPI cell sample (d).

### 4. Discussion

In summary, we presented an extension of the Lukosz-bound to 3D-SIM and incorporated this bound as apodization function in 3D-SIM image reconstruction. The reconstruction is formulated in the framework of Tikohonov-Miller filtering, an approach that unifies the linear combination of bands shifted in spatial frequency space with filtering steps for noise suppression, regularization and apodization. We presented a formalism to calculate the 3D Lukosz-bound for any given surface that describes the cut-off of a 3D OTF in spatial frequency space. An analytical description of this cut-off surface for 3D SIM, valid for unequally spaced rotation angles and rotation-dependent grating periods, was derived as well. We compared the 3D SIM Lukosz-bound to other choices for the apodization function that are currently used in SIM reconstructions. The Lukosz-bound results in a PSF with an order of magnitude reduction of negativity at the expense of a moderate increase in width when compared to the distance transform based triangle function. Switching to Lukosz-bound based apodization reduces ringing artifacts and avoids the need for clipping the final image as shown for the tubulin and DAPI results in Figs. 3 and 4.

Apodization with the 3D Lukosz-bound does not require fine-tuning of any free parameters in the apodization function, such as taking it to a power other than one, as is sometimes done with the triangular apodization function [23]. Note that raising the 3D Lukosz-bound to a power smaller than one would violate the non-negativity criteria. Raising it to a power larger than one causes an unnecessary reduction of the "effective" width in frequency space, leading to a wider PSF.