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## Artifact-free whole-slide imaging with structured illumination microscopy and Bayesian image reconstruction --Manuscript Draft--

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<b>Abstract:</b>	<p><b>Background</b></p> <p>Structured illumination microscopy (SIM) is a method which can be used to image biological samples and can achieve both optical sectioning and super-resolution effects. Optimization of the imaging setup and data processing methods results in high quality images without artifacts due to mosaicking or due to the use of SIM methods. Reconstruction methods based on Bayesian estimation can be used to produce images with a resolution beyond that dictated by the optical system.</p> <p><b>Findings</b></p> <p>Five complete datasets are presented including large panoramic SIM images of human tissues in pathophysiological conditions. Cancers of the prostate, skin, ovary, and breast, as well as tuberculosis of the lung, were imaged using SIM. The samples are available commercially and are standard histological preparations stained with hematoxylin and eosin.</p> <p><b>Conclusion</b></p> <p>The use of fluorescence microscopy is increasing in histopathology. There is a need for methods which reduce artifacts when employing image stitching methods or optical sectioning methods such as SIM. Stitched SIM images produce results which may be useful for intraoperative histology. Releasing high quality, full slide images and related data will aid researchers in furthering the field of fluorescent histopathology.</p>	
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<b>Response to Reviewers:</b>	Dear Editor: We would like to re-submit our manuscript entitled "Artifact-free whole-slide imaging with structured illumination microscopy and Bayesian image reconstruction" for	

	<p>consideration in GigaScience as a data note. We made the following revisions.</p> <ol style="list-style-type: none"> <li>1. The reviewer noticed that two of the data files was incomplete or corrupted. We reloaded these two files.</li> <li>2. We included a table showing the organization of the data files along with their sizes.</li> <li>3. We included a universal public domain dedication text file with the dataset.</li> <li>4. We added a statement about ethics approval and consent to participate.</li> <li>5. We moved all of the information from the supplement into the main paper. We removed the example of the cellular-scale imaging.</li> </ol> <p>Please note that the figures in the automatically generated PDF look terrible as usual. Two of the figures look like blank black boxes. There is nothing I can do about this. Please click the links embedded in the PDF to view the figures.</p> <p>We hope that our paper will now be acceptable for publication in GigaScience.</p> <p>Sincerely,</p> <p>Guy M. Hagen</p>
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# Artifact-free whole-slide imaging with structured illumination microscopy and Bayesian image reconstruction

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## Abstract

**Background** Structured illumination microscopy (SIM) is a method which can be used to image biological samples and can achieve both optical sectioning and super-resolution effects. Optimization of the imaging setup and data processing methods results in high quality images without artifacts due to mosaicking or due to the use of SIM methods. Reconstruction methods based on Bayesian estimation can be used to produce images with a resolution beyond that dictated by the optical system.

**Findings** Five complete datasets are presented including large panoramic SIM images of human tissues in pathophysiological conditions. Cancers of the prostate, skin, ovary, and breast, as well as tuberculosis of the lung, were imaged using SIM. The samples are available commercially and are standard histological preparations stained with hematoxylin and eosin.

**Conclusion** The use of fluorescence microscopy is increasing in histopathology. There is a need for methods which reduce artifacts when employing image stitching methods or optical sectioning methods such as SIM. Stitched SIM images produce results which may be useful for intraoperative histology. Releasing high quality, full slide images and related data will aid researchers in furthering the field of fluorescent histopathology.

**Keywords** Structured illumination microscopy, SIM, image stitching, Bayesian methods, MAP-SIM, SIMToolbox, histopathology, cancer.

## 24 **Data description**

### 25 **Context**

26           Structured illumination microscopy (SIM) is a method in optical fluorescence microscopy which can  
27 achieve both optical sectioning (OS-SIM) [1] and resolution beyond the diffraction limit (SR-SIM) [2,3]. SIM has  
28 been used for super-resolution imaging of both fixed and live cells [4–7] and has matured enough as a method that it  
29 is now available commercially. In SIM, a set of images is acquired using an illumination pattern which shifts  
30 between each image. As SIM has developed, diverse strategies have been proposed for creation of the SIM  
31 pattern [1,8–13]. Several different approaches for processing the data have also been introduced [3,7,8,14–16].

32           Recently, microscope systems capable of imaging with high resolution and a large field of view (FOV)  
33 have been developed [17–21], some using custom-made microscope objectives. However, stitching together images  
34 acquired with a higher magnification objective to create a large mosaic remains a valid and popular approach. Some  
35 published results involving stitched images suffer from pronounced artifacts in which the edges of the individual  
36 sub-images are visible, usually as dark bands which outline each sub-image [22–24]. On the other hand, several  
37 studies have proposed methods for stitching of microscope images with reduced artifacts [25–32].

38           The combination of SIM with image stitching methods allows collection of large FOV images with both  
39 optical sectioning and super-resolution properties. Here, we demonstrate methods and provide complete datasets for  
40 five different samples. The samples are hematoxylin and eosin (H&E) stained histological specimens which provide  
41 examples of human diseases (ovarian cancer, breast cancer, prostate cancer, skin cancer, and tuberculosis), and  
42 which are also available commercially for those who wish to reproduce our work. We used freely available optical  
43 designs [6,10,33] and open source software [33] for SIM imaging, along with freely available software for image  
44 stitching (Microsoft Image Composite Editor (ICE) [34], or a well validated plugin [26] for ImageJ [35]).  
45 Combining this with deconvolution methods, we produced stitched images which are free of noticeable artifacts from  
46 stitching or from SIM reconstruction.

47           Fluorescence microscopy is becoming more important in histopathology. Traditional bright field  
48 microscopy diagnostic methods require a time-consuming process, involving chemical fixation and physical  
49 sectioning. The use of optical sectioning fluorescence microscopy allows high-quality images to be captured without  
50 the need for physical sectioning. Consequently, it has been shown that imaging can be performed on large human

51 tissue samples within 1 hour after excision [36]. Additionally, other studies have shown the results of fluorescence  
 52 imaging to be usable and accurate in diagnosis of various medical conditions [37–42]. Previously, it was noted that  
 53 obvious stitching artifacts significantly decrease the usability of large fluorescence images in medical diagnosis. In  
 54 one case, such artifacts resulted in the rejection of over half of the images acquired [38]. The setup we describe here  
 55 allows for fast, artifact-free, high-resolution imaging of fluorescent samples, and is compatible with samples stained  
 56 with most fluorescent dyes.

## 57 **Methods**

### 58 *Samples*

59 All samples used in this study are available from Carolina Biological, Omano, or Ward’s Science. The  
 60 samples are approximately 7  $\mu\text{m}$  thick and are stained with hematoxylin and eosin. The commercial source, product  
 61 number, and other SIM imaging parameters for each sample are detailed in Table 1. Table 2 details imaging  
 62 parameters for acquisitions of each sample with a color camera.

63 Table 1: Imaging parameters for the SIM datasets

Sample	Source company and product no.	SIM pattern no. of phases	Exposure time, ms	No. of tiles	Objective mag/NA	Acquisition time, s	Stitching software
Carcinoma of Prostate	Carolina, 318492	5	50	23 $\times$ 11	20 $\times$ /0.45	315	Microsoft ICE
Basal Cell Carcinoma	Ward’s Science, 470183-256	6	75	29 $\times$ 18	30 $\times$ /1.05	821	FIJI
Adenocarcinoma of Ovary	Carolina, 318628	5	100	25 $\times$ 14	10 $\times$ /0.4	595	Microsoft ICE
Adenocarcinoma of Breast	Carolina, 318766	8	200	12 $\times$ 8	10 $\times$ /0.4	278	FIJI
Lung Tuberculosis	Omano, OMSK-HP50	5	100	20 $\times$ 16	30 $\times$ /1.05	541	FIJI

64 Table 2: Parameters for the color images

Sample	No. of tiles	Objective mag/NA
Carcinoma of Prostate	6 $\times$ 5	4 $\times$ /0.16
Basal Cell Carcinoma	5 $\times$ 5	4 $\times$ /0.16
Adenocarcinoma of Ovary	11 $\times$ 11	4 $\times$ /0.16
Adenocarcinoma of Breast	6 $\times$ 6	4 $\times$ /0.16
Lung Tuberculosis	8 $\times$ 10	10 $\times$ /0.4

### 65 *Microscope setup and data acquisition*

66 We used a home-built SIM setup based on the same design as described previously [6,10,15] (Figure 1).  
 67 The SIM system is based on an IX83 microscope (Olympus) equipped with a Zyla 4.2+ sCMOS camera (Andor)  
 68 under the control of IQ3 software (Andor). We used the following Olympus objectives: UPLSAPO 4 $\times$ /0.16 NA,

69 UPLSAPO 10×/0.4 NA, LUCPLFLN 20×/0.45 NA, and UPLSAPO 30×/1.05 NA silicone oil immersion. For color  
70 images we used an aca1920-40uc color camera (Basler) under control of Pylon software (Basler). We used a MS-  
71 2000 motorized microscope stage (Applied Scientific Instrumentation) to acquire tiled SIM images. In all datasets,  
72 the stage scanning was configured such that all image edges overlapped by 20%.

73 Briefly, the SIM system uses a ferroelectric liquid crystal on silicon (LCOS) microdisplay (type SXGA-  
74 3DM, Forth Dimension Displays). This device has been used previously in SIM and related methods in fluorescence  
75 microscopy [5,10,15,33,43–47] and allows one to produce patterns of illumination on the sample which can be  
76 reconfigured at will by changing the image displayed on the device. The light source (Lumencor Spectra-X) is  
77 toggled off between SIM patterns and during camera readout. Figure 1 shows a simplified diagram of the  
78 microscope system.

#### 79 **INSERT FIGURE 1-system diagram**

#### 80 *Motorized stage and illumination control*

81 Close synchronization between the camera acquisitions, light source, and microdisplay ensures rapid image  
82 acquisition, helps reduce artifacts, and reduces light exposure to the sample. As shown in Figure 2, Andor IQ  
83 software controls the SIM system.

#### 84 **INSERT FIGURE 2-connection diagram**

85 While the camera and XYZ stage receive signals directly from Andor IQ software and an input/output computer  
86 card (Measurement Computing DDA06/16), the illumination signals which the software generates must be altered  
87 before being sent to the light source. Firstly, the microdisplay used in our setup will not produce an image on the  
88 sample if it is illuminated with a constant light source. Rather, a meaningful illumination pattern will only form if  
89 the light source is synchronized with an enable signal output from the microdisplay control board. Therefore, the  
90 channel signals output by IQ are first modulated with the microdisplay enable signal (this is performed by the  
91 leftmost AND gates pictured in Figure 2). Additionally, to reduce unnecessary light exposure to the sample, the light  
92 source is shut off whenever the camera sensor is not being exposed. This is accomplished by performing a second  
93 logical AND of the result of the previous AND with the ‘FIRE’ signal output from the camera. This process is  
94 illustrated in Figure 3.

95 **INSERT FIGURE 3-trigger**

96 *SIM data processing*

97 SIM reconstructions were performed in the same way as previously described using SIMToolbox, an open-  
98 source and freely available program that our group developed for processing SIM data [33]. We generated optically  
99 sectioned, enhanced resolution images using a Bayesian estimation method, maximum *a posteriori* probability SIM  
100 (MAP-SIM) [15]. MAP-SIM works using maximum *a posteriori* probability methods, which are well known in  
101 microscopy applications [48,49], to enhance high spatial frequency image information. We then combine this  
102 information, in the frequency domain, with low spatial frequency image information obtained by OS-SIM methods,  
103 then produce the final image by an inverse Fourier transform [15]. We typically measure the final resolution  
104 obtained by analyzing the frequency spectrum of the resulting image, as is discussed below.

105 The illumination patterns used here are generated such that the sum of all positions in each pattern set  
106 results in homogenous illumination. As such, a widefield (WF) image can be reconstructed from SIM data simply by  
107 performing an average intensity projection of the patterned images. This can be described by

108 
$$I_{WF} = \frac{1}{N} \sum_{n=1}^N I_n,$$

109 where N is the number of pattern phases,  $I_n$  is the image acquired on the  $n^{\text{th}}$  illumination position, and  $I_{WF}$  is the WF  
110 reconstruction. This is the method we used to generate WF images throughout this study.

111 *Vignetting correction*

112 Following SIM reconstruction, vignetting artifacts remain in each tile. If not removed prior to stitching, this  
113 vignetting introduces a distracting grid pattern in the final stitched image. We performed vignette removal by  
114 dividing each tile of the mosaic by an image representing the vignetting profile common to all tiles. Other studies  
115 have used an image of a uniformly fluorescent calibration slide as a reference for vignette removal [36], where  
116 information concerning non-uniform illumination is captured. However, we found that SIM processing introduces  
117 vignetting artifacts beyond those due to non-uniform illumination. Additionally, these artifacts vary somewhat  
118 depending on properties of the sample being imaged. As such, performing pre-acquisition calibration on a uniformly

119 fluorescent slide is not sufficient to remove vignetting artifacts from SIM reconstructions. Instead, an estimate of the  
120 vignetting profile is found through analysis of the mosaic tiles after SIM reconstruction.

121           A blurred average intensity projection of the tiles is a good approximation of the vignetting profile, as an  
122 average intensity projection merges the tiles into a single image with averaged foreground information while  
123 preserving the vignetting profile. Subsequent blurring with an appropriate radius and edge-handling method also  
124 eliminates the high spatial frequency foreground without impacting the low spatial frequency illumination profile.  
125 To eliminate errors during the blurring step due to the blurring area extending outside the original image, we used an  
126 edge handling method in which the blurring area is reduced near the edges of the image such that no values outside  
127 the image border are sampled. Unlike edge handling methods in which the image is padded with a uniform value (or  
128 mirrored and tiled) to accommodate a blurring area which extends beyond the original image limits, this method is  
129 free from major artifacts, such as erroneous brightness of the image edges, see figure 4. While the average intensity  
130 projection removes most foreground information from the images, some coarseness remains in the vignette  
131 estimation after this step (figure 4a), especially for stitches with fewer than 50 tiles. The blurring step serves to  
132 eliminate only this non-vignette information and must preserve the illumination profile. As shown in figure 4b, use  
133 of the default Gaussian blurring function in ImageJ introduces a bright glow near to borders of the image, a  
134 significant artifact which does not reflect the original vignetting profile. Use of the ‘border limited mean’ filter, on  
135 the other hand, does not introduce this aberration, as shown in figure 4c.

#### 136 **INSERT FIGURE 4-edge method**

137           Approximating the illumination profile works especially well for histological samples, as such samples are  
138 non-sparse and require many tiles, factors which improve the accuracy of this approach. We performed all steps of  
139 this devignetting process using built-in functions and the ‘Fast Filters’ plugin in ImageJ [50]. The effect of  
140 devignetting is illustrated in Figure 5.

#### 141 **INSERT FIGURE 5-devignetting**

142 *Image Stitching*

143 With visible vignetting removed, we then stitched together a composite image from the tiles. The pre-  
144 processing allows for stitching to be done in various stitching applications; Microsoft ICE and Preibisch's plugin for  
145 FIJI [26] were used to stitch the data presented here.

146 The data processing procedure described here is summarized in Figure 6. The total time for processing each  
147 dataset was about 30 min.

## 148 **INSERT FIGURE 6-flowchart**

### 149 *Color image data processing methods*

150 We created color overview images by stitching devignetted brightfield acquisitions. Devignetting was  
151 performed simply by adding the inverse of an empty brightfield acquisition to each color tile using ImageJ. For this  
152 method to produce optimal results, the empty brightfield image must be acquired in conditions identical to those of  
153 the raw tile data, such that the illumination profile in the empty image matches that of the unprocessed tiles. This  
154 simple operation removes nearly all visible vignetting and color balance artifacts within each tile. The results after  
155 devignetting were then stitched using Preibisch's plugin for FIJI [26].

### 156 *Resolution measurement*

157 We evaluated our results by measuring image resolution using SR Measure Toolbox. SR Measure  
158 Toolbox [51] measures the resolution limit of input images through analysis of the normalized, radially averaged  
159 power spectral density ( $PSD_{ca}$ ) of the images, as previously described [6]. Briefly, the resolution limit in real space  
160 is determined by evaluating the cutoff frequency in Fourier space. The cutoff frequency is estimated by calculating  
161 the spatial frequency at which the  $PSD_{ca}$  (after noise correction) drops to zero.

162 Focusing on the basal cell carcinoma sample, we selected 125 (out of 522 total) image tiles, calculated the  
163 PSD and resolution for each tile, and averaged the results. We found that, in the case of this sample, the image  
164 resolution was  $593 \pm 20$  nm for WF and  $468 \pm 2.5$  nm for MAP-SIM (average  $\pm$  standard deviation). This data was  
165 acquired with a UPLSAPO 30 $\times$ /1.05 NA silicone oil immersion objective. Figure 7 shows an example measurement  
166 for one image tile. Figure 8 shows a plot of  $PSD_{ca}$  for this image tile.

## 167 **INSERT FIGURE 7-resolution**

168 **INSERT FIGURE 8-PSD**

## 169 **Results**

170 Figure 9 shows images of a prepared slide containing a human prostate carcinoma sample stained with  
171 H&E. Figure 9a shows a stitched color overview, and Figure 9d shows a zoom-in of the region indicated in Figure  
172 9a, acquired separately using a UPLSAPO 20×/0.75NA objective. Figure 9b shows a stitched widefield fluorescence  
173 image, and Figure 9c shows a stitched SIM image. Figures 9e and 9f each show zoom-ins of the stitches shown in  
174 Figures 9b and 9c, respectively. Using the acquisition and processing methods described, whole-slide images are  
175 produced without any visible stitching artifacts. Additionally, the MAP-SIM reconstruction method produces  
176 resolution superior to that of the widefield data. Figures 10-13 show similar comparisons for basal cell carcinoma,  
177 ovary adenocarcinoma, breast adenocarcinoma, and tuberculosis of the lung, respectively.

178 The data shown in figures 9-13 is freely available through Giga DB[52]. This dataset includes all color  
179 overviews as well as WF and MAP-SIM stitches at full resolution. In addition, all image tiles used to create the WF  
180 and MAP-SIM stitches of the samples are provided.

181 **INSERT FIGURE 9-prostate**

182 **INSERT FIGURE 10-basal cell**

183 **INSERT FIGURE 11-ovary**

184 **INSERT FIGURE 12-breast**

185 **INSERT FIGURE 13-lung**

## 186 **Discussion**

187 Many past studies into stitching of SIM mosaics have suffered from noticeable image artifacts, arising from  
188 flaws in the optical setups used as well as imperfections in the SIM reconstruction and image stitching processes.  
189 While these artifacts are sometimes minimal enough to remain uncorrected, certain artifacts seriously inhibit the  
190 usefulness of the final stitched image. In [23], the authors note that issues in triggering and evenly illuminating the  
191 microdisplay being used for illumination resulted in striping and vignetting artifacts; similarly, in [22,24,36,53],  
192 stitching artifacts are apparent in the images.

193 Here, optimization of the optical setup, camera-microdisplay synchronization, and image processing  
194 methods yielded whole-slide images free from visible SIM or image stitching artifacts. In addition to the elimination  
195 of artifacts, our use of SIMToolbox to perform SIM reconstruction on the data allows for a variety of reconstruction  
196 algorithms to be used, including super-resolution algorithms such as MAP-SIM. This too presents an improvement  
197 over previous works.

198 Another advantage of the acquisition and processing methods demonstrated here is the minimization of  
199 user intervention, and in turn, reductions in acquisition and processing time. Stage movement, sample focusing,  
200 image acquisition, and SIM pattern advancement are controlled automatically. Loading of the sample, definition of  
201 the mosaic edges, and manual focus on 3-5 positions in the sample are the only steps needed to be taken by the user  
202 before acquisition can begin. Recent developments in autofocus technology for SIM may allow for the manual focus  
203 step to be shortened or omitted [53]. These automated steps during acquisition allow for large mosaics to be  
204 acquired. The quality of the final stitched images does not degrade for larger mosaics – in fact, the quality of the  
205 deghosting process improves with larger datasets, as more data is available to produce an accurate estimation of  
206 the illumination profile. SIMToolbox (version 2.0), which is capable of utilizing the processing power of modern  
207 consumer graphics cards during MAP-SIM processing, also reduces the time spent during the data processing phase.  
208 Finally, unlike other super-resolution reconstruction methods such as SR-SIM, MAP-SIM is able to produce artifact-  
209 free results without tuning of reconstruction parameters by the user, a process which is difficult to automate and  
210 requires significant user experience.

211 One drawback of the method presented here is the inability to image the entire volume of samples thicker  
212 than ~0.5 mm. However, this limitation does not prevent large, unsectioned samples from being imaged, as is the  
213 case with bright field microscopy, where samples must be thin enough for transmitted light to reach the objective.  
214 Rather, as the light which illuminates the sample in fluorescence microscopy emanates from the objective, all  
215 surface regions of a large sample may be imaged. Additionally, due to the optical sectioning exhibited by SIM, light  
216 from out-of-focus regions of the sample is almost completely attenuated. Consequently, imaging the surfaces of  
217 large samples with SIM produces high quality images without the need for physical sectioning, as previously  
218 demonstrated [23,36].

219 Here, we demonstrated our imaging techniques on traditionally prepared histopathological samples in order  
220 to provide a comparison between bright field imaging and SIM, but the same techniques can be used to image a  
221 wide variety of fluorescently labelled samples. The ability to seamlessly image the entire surface region of large  
222 samples has multiple potential applications in histopathology. SIM presents unique advantages in analyzing the  
223 surgical margins of large tissue excisions, as demonstrated by Wang, et al [36]. Confocal imaging of core needle  
224 biopsy samples has been previously demonstrated to produce images suitable for medical diagnosis [42], a practice  
225 easily adapted to SIM. The speed at which sample preparation and image acquisition can be performed in  
226 fluorescence microscopy presents opportunities for intra-operative analysis of tissue samples using SIM techniques,  
227 as mentioned by multiple other studies [23,36,54,55].

### 228 **Reuse potential and example of data reuse**

229 The data provided here presents various opportunities for reuse. With the multiple high-resolution color  
230 overviews and stitched SIM images, comparison of structures visible in the brightfield and fluorescent images could  
231 be performed to further study the use of fluorescence microscopy in histopathology. The unstitched image tiles of  
232 the basal cell carcinoma sample provided in the dataset, which still contain vignetting artifacts, may be used to  
233 reproduce the results of our devignetting process, as well as to further develop new devignetting approaches suited  
234 for SIM. These tiles might also be used to create or modify existing stitching software for global minimization of  
235 stitching artifacts. Note that the image tiles from the other samples in the dataset are provided after devignetting.

236 Gone uncorrected, vignetting in the image tiles used to stitch a larger image can cause a noticeable grid  
237 pattern in the final stitched image. While readily noticeable upon viewing of the image, quantification of this pattern  
238 is useful for evaluation of methods which remove it. Since this stitching artifact arises from an illumination profile  
239 common to each tile, the period of this pattern in the stitched image can simply be represented by the spacing  
240 between tiles used during acquisition:

$$241 \quad T_{stitch} = l \cdot (1 - r),$$

242 where  $l$  is the image width and  $r$  is the proportional overlap between image tiles. The parameters for the dataset  
243 visualized in Figure 14 are  $l = 2048$  px and  $r = 0.2$ , thus  $T_{stitch} = 1638$  px, or  $\sim 355$   $\mu\text{m}$  (as the dataset has a pixel size  
244 of  $\sim 216.7$  nm). In our setup, the camera sensor is square, so  $T_{stitch}$  is the same both the horizontally and vertically in  
245 the final image. As a pattern with a very consistent period, this grid artifact manifests in the FFT of an uncorrected

246 stitch as series of bright peaks. As shown in Fig 14e, the location of the peaks corresponding to the fundamental  
 247 frequency of the grid pattern agrees very well with the calculated  $T_{stitch}$ . Figs. 14d and 14e show that the FFT of a  
 248 properly corrected image contains no trace of the peaks evident in the uncorrected image.

249 **INSERT FIGURE 14-analysis**

250 **Availability of source code and requirements**

251 Project name: SIMToolbox version 2.12  
 252 Project home page: <http://mmtg.fel.cvut.cz/SIMToolbox/>  
 253 Operating system: platform independent  
 254 Programming language: MATLAB  
 255 License: GNU General Public License v3.0

256 **Detailed software compatibility notes**

257 The SIMToolbox GUI was compiled with MATLAB 2015a and tested in Windows 7 and 8. The GUI is a stand-alone  
 258 program and does not require MATLAB to be installed. To use the MATLAB functions within SIMToolbox (i.e.,  
 259 without the GUI), MATLAB must be installed. The functions were mainly developed with 64 bit MATLAB versions  
 260 2012b, 2014a, 2015a in Windows 7. When using SIMToolbox functions without the GUI, the MATLAB “Image  
 261 Processing Toolbox” is required. SIMToolbox also requires the “MATLAB YAML” package to convert MATLAB  
 262 objects to/from YAML file format. Note that this package is installed automatically when using the GUI.

263 **Availability of data**

264 All raw and analyzed data is available on GigaDB [52]. All files and data are distributed under the Creative Commons  
 265 CC0 waiver, with a request for attribution. The data is organized into five main folders for the five different samples,  
 266 see table 3.

267 Table 3: Description of the data files

Folder	Files	File Size
01-Prostate-carcinoma	Prostate-carcinoma-WF-tiles.zip	3.6 GB
	Prostate-carcinoma-MAPSIM-tiles.zip	3.5 GB
	Prostate-carcinoma-WF-stitch.tif	695 Mb
	Prostate-carcinoma-MAPSIM-stitch.tif	695 Mb
	Prostate-carcinoma-Color-stitch.tif	118 MB
02-Basal-cell-carcinoma	Basal-cell-carcinoma-MAPSIM-tiles.zip	7.5 GB

	Basal-cell-carcinoma-WF-tiles.zip	7.5 GB
	Basal-cell-carcinoma-MAPSIM-stitch.tif	6.3 GB
	Basal-cell-carcinoma-WF-stitch.tif	5.7 GB
	Basal-cell-carcinoma-color-stitch.tif	123 MB
03-Ovary-adenocarcinoma	Ovary-adenocarcinoma-MAPSIM-tiles.zip	891 MB
	Ovary-adenocarcinoma-WF-tiles.zip	849 MB
	Ovary-adenocarcinoma-MAPSIM-stitch.tif	916 MB
	Ovary-adenocarcinoma-WF-stitch.tif	979 MB
	Ovary-adenocarcinoma-Color-stitch.tif	610 MB
04-Breast-adenocarcinoma	Breast-adenocarcinoma-MAPSIM-tiles.zip	1.4 GB
	Breast-adenocarcinoma-WF-tiles.zip	1.4 GB
	Breast-adenocarcinoma-MAPSIM-stitch.tif	1.1 GB
	Breast-adenocarcinoma-WF-stitch.tif	1.1 GB
	Breast-adenocarcinoma-Color-stitch.tif	129 MB
05-Lung-tuberculosis	Lung-tuberculosis-MAPSIM-tiles.zip	4.8 GB
	Lung-tuberculosis-WF-tiles.zip	4.6 GB
	Lung-tuberculosis-MAPSIM-stitch.tif	3.8 GB
	Lung-tuberculosis-WF-stitch.tif	3.6 GB
	Lung-tuberculosis-Color-stitch.tif	341 MB

268

269 **Abbreviations**

270 Av Int Proj, average intensity projection; FOV, field of view; H&E, hematoxylin and eosin; ICE, Image Composite  
271 Editor; MAP-SIM, maximum *a posteriori* probability SIM; NA, numerical aperture; LCOS, liquid crystal on silicon;  
272 PSDca, circularly averaged power spectral density; SIM, structured illumination microscopy; WF, wide field.

273 **Ethics approval and consent to participate**

274 Because the samples were acquired commercially, and because they are completely de-identified (meaning that there  
275 is no way to connect these particular samples to the original donor), this is not considered human subject research,  
276 and approval is not required to work with these samples.

277 **Consent for publication**

278 Not applicable

279 **Competing interests**

280 The authors declare that they have no competing interests.

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289 **Author Contributions**

290 KJ: acquired data, analyzed data, wrote the paper

291 GH: conceived project, acquired data, analyzed data, supervised research, wrote the paper

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431

## 432 **FIGURE CAPTIONS**

433 Figure 1: Simplified diagram of SIM system. LCOS, liquid crystal on silicon

434 Figure 2. Simplified connection diagram.

435 Figure 3. Illumination trigger signal logic. This logical operation occurs for each illumination wavelength at the  
436 rightmost AND gates pictured in Figure 2.

437 Figure 4: Edge handling during the blurring process. (a) is the result after performing an average intensity projection  
438 on a set of MAP-SIM tiles. (b) shows (a) after being blurred using ImageJ's 'Gaussian blur' with a radius of 200  
439 pixels, while (c) shows (a) blurred using the 'border limited mean' filter.

440 Figure 5: Vignetting artifacts and their removal. (a) shows the result of stitching images without applying the  
441 devignetting process, while (b) shows a stitch of the same data after devignetting has been applied. (c) shows the  
442 average intensity projection of the images used to stitch (a), which estimates the vignette profile of each frame. This  
443 estimate can be refined by application of an edge-limited blurring filter, as shown in (d). (e) shows the average  
444 intensity projection of the data used in (b), after devignetting has been applied. The uniform brightness of (e)  
445 indicates that no major vignetting artifacts remain in the devignetted data.

446 Figure 6. Panoramic SIM data processing workflow. Devignetting was performed after SIM reconstruction. Note the  
447 vignette profile differs between reconstruction methods, necessitating separate projection, blurring and division  
448 steps. Av Int Proj, average intensity projection.

449 Figure 7. Evaluating image resolution. (a) and (b) show a tile from the data in Figure 10 (basal cell carcinoma  
450 sample) after widefield and MAP-SIM reconstruction, respectively. (c) and (d) each show a zoom-in of (a) and (b),  
451 respectively. (e) and (f) each show the FFT of (a) and (b), respectively. The dotted lines in (e) and (f) indicate the  
452 resolution of each image according to the resolution measurement described.

453 Figure 8. Normalized, radially averaged power spectral density ( $PSD_{ca}$ ) and resolution analysis measured on the tiles  
454 shown in Figs. 7a and 7b.

455 Figure 9: Carcinoma of human prostate. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a region  
456 of the sample indicated in (a). (e) and (f) each show a zoom-in of (b) and (c), respectively, in the region indicated in  
457 (a).

458 Figure 10: Basal Cell Carcinoma. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a region of the  
459 sample indicated in (a). (e) and (f) each show a zoom-in of (b) and (c), respectively, in the region indicated in (a).

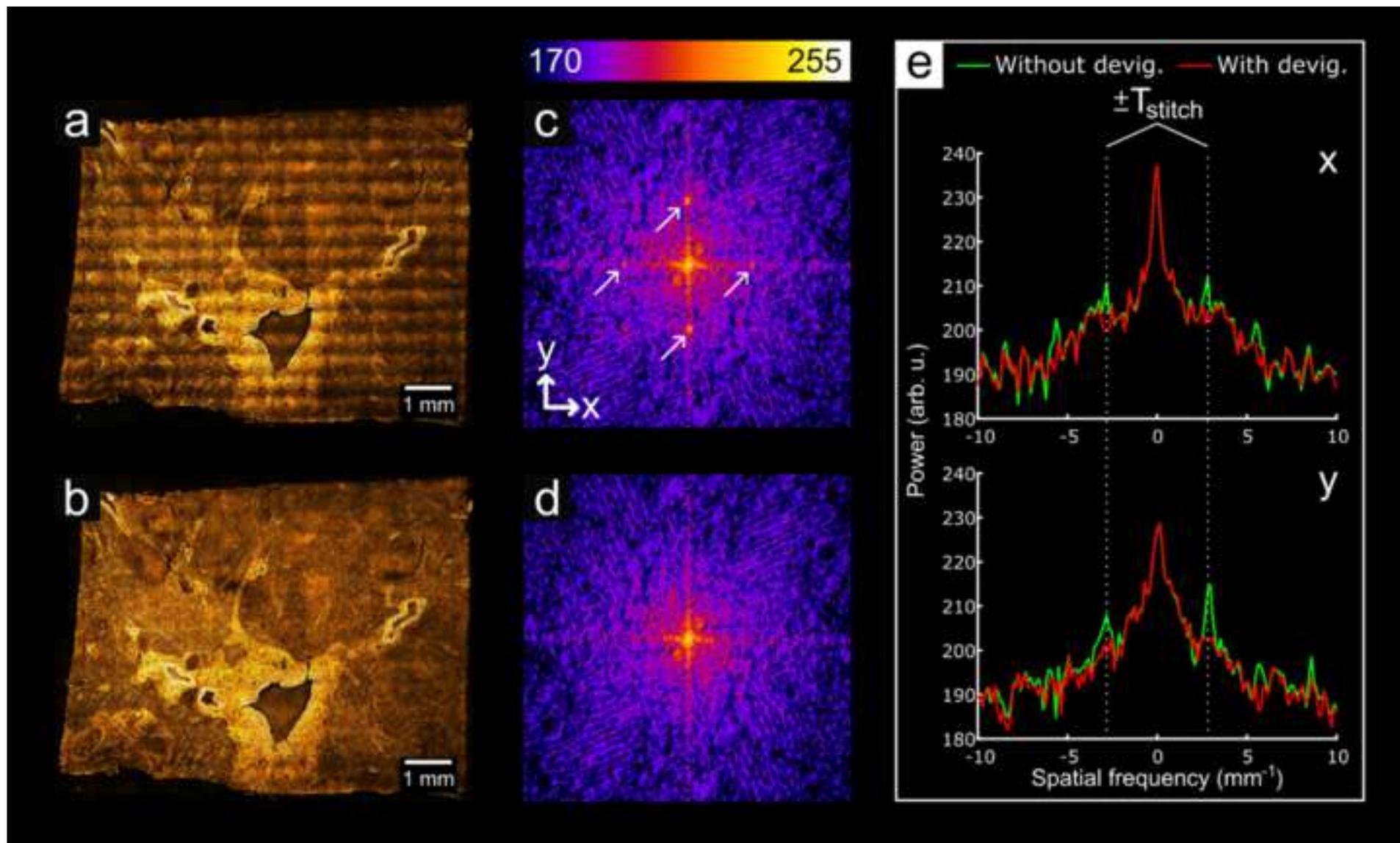
460 Figure 11: Adenocarcinoma of human ovary. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d), (g) show a  
461 region of the sample indicated in (a), acquired separately from (a) using a 10 $\times$  objective. (e) and (h) show a zoom-in  
462 of (b), while (f) and (i) show a zoom-in of (c), all in the regions indicated in (a).

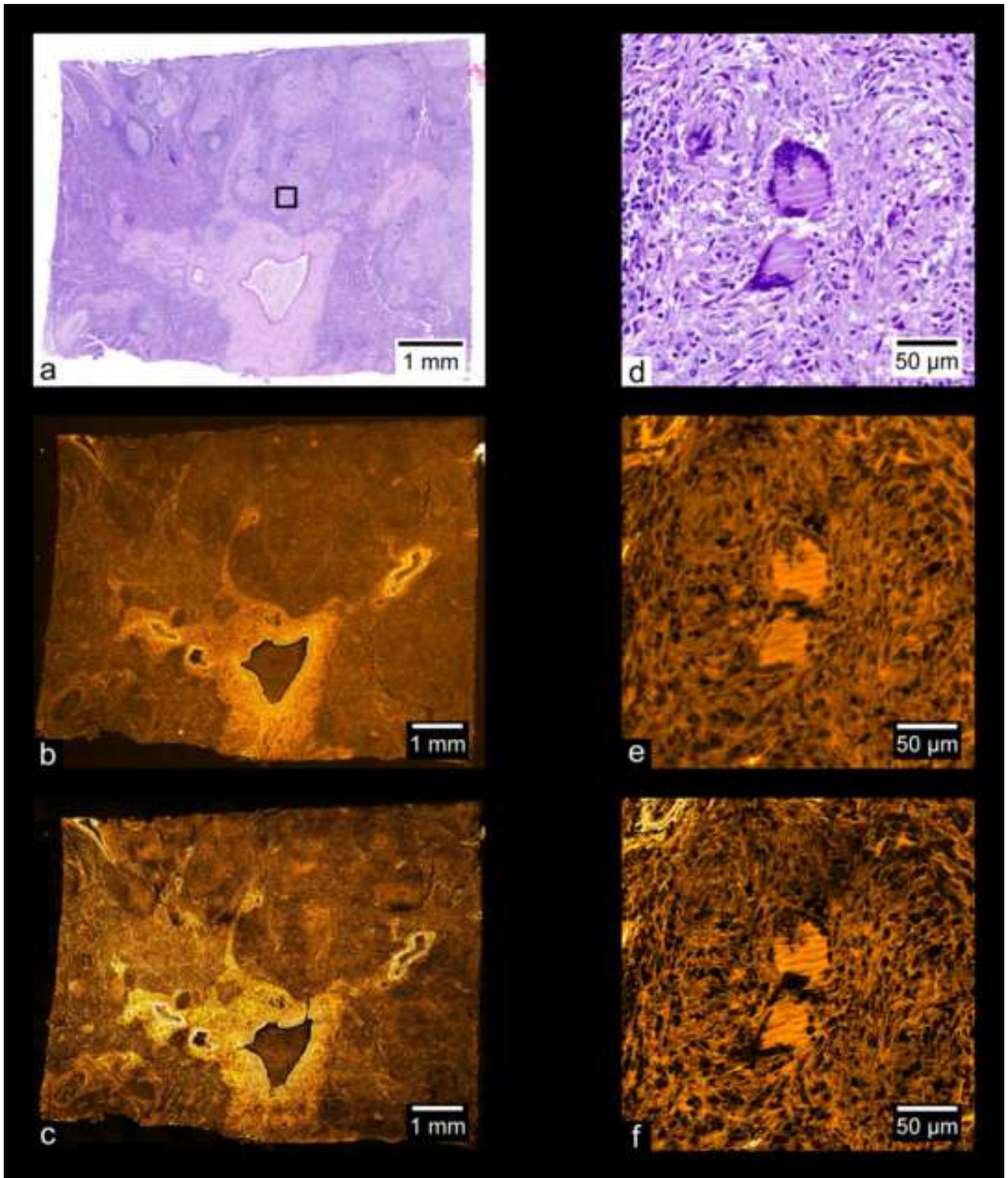
463 Figure 12: Adenocarcinoma of human breast. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a  
464 region of the sample indicated in (a), acquired separately from (a) using a 10 $\times$  objective. (e) and (f) each show a  
465 zoom-in of (b) and (c), respectively, in the region indicated in (a).

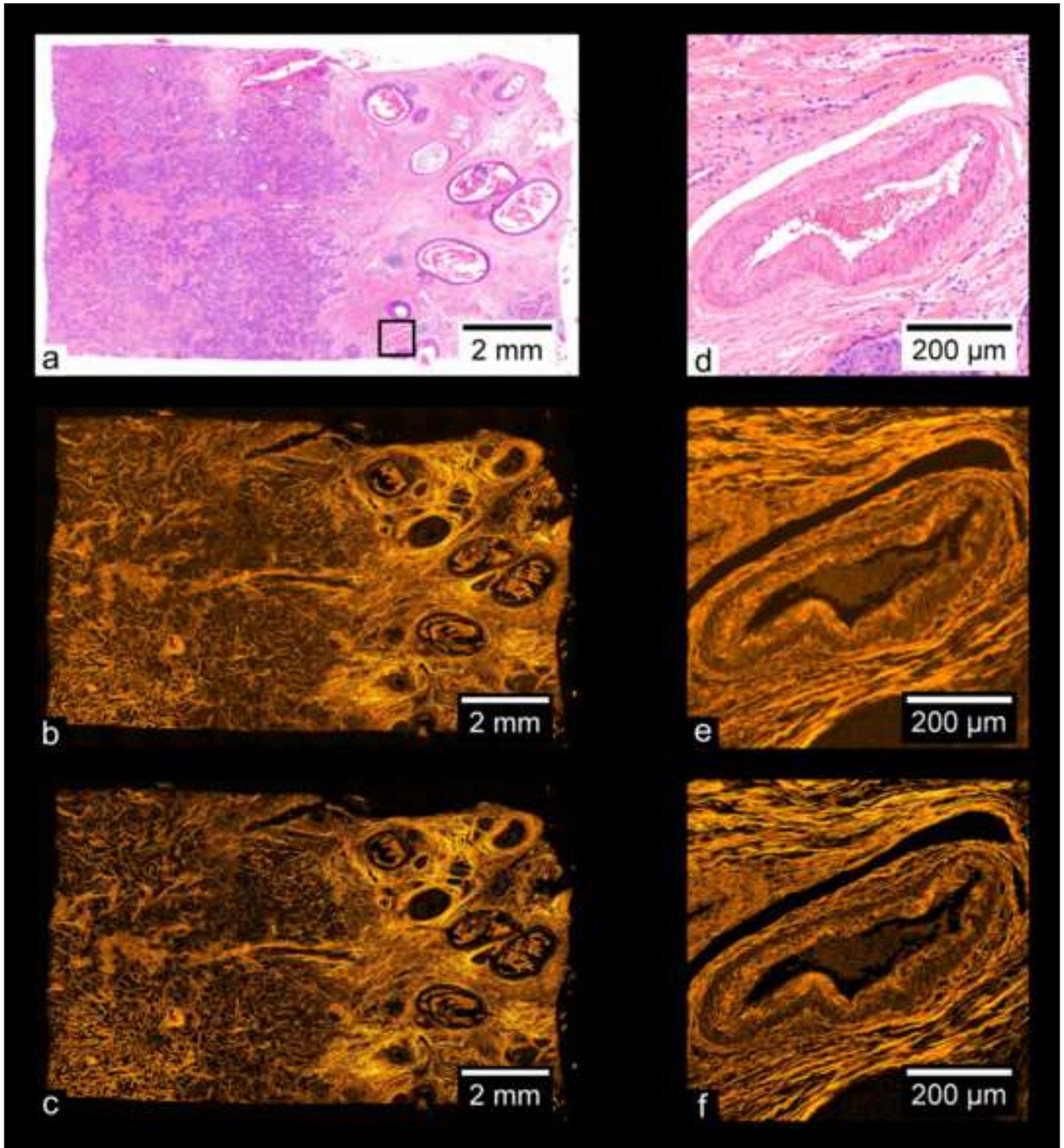
466 Figure 13: Tuberculosis of human lung. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a region  
467 of the sample indicated in (a), acquired separately from (a) using a 20 $\times$  objective. (e) and (f) each show a zoom-in of  
468 (b) and (c), respectively, in the region indicated in (a).

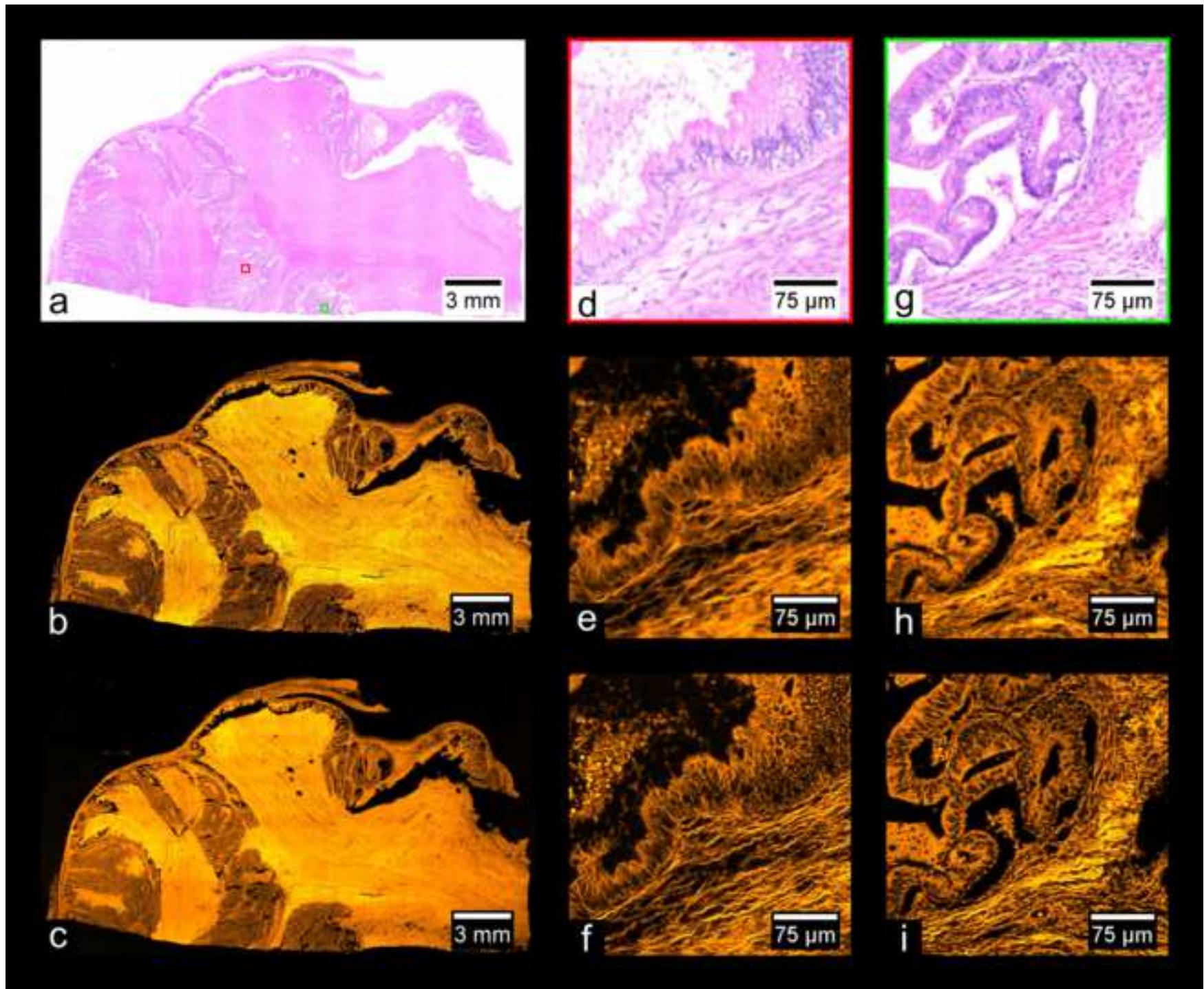
469 Figure 14: Analysis of periodic stitching artifacts in the frequency domain. (a) shows a stitched image where  
470 devignetting has not been applied to the tiles, while (b) shows a stitch of devignetted data. (c) and (d) show the

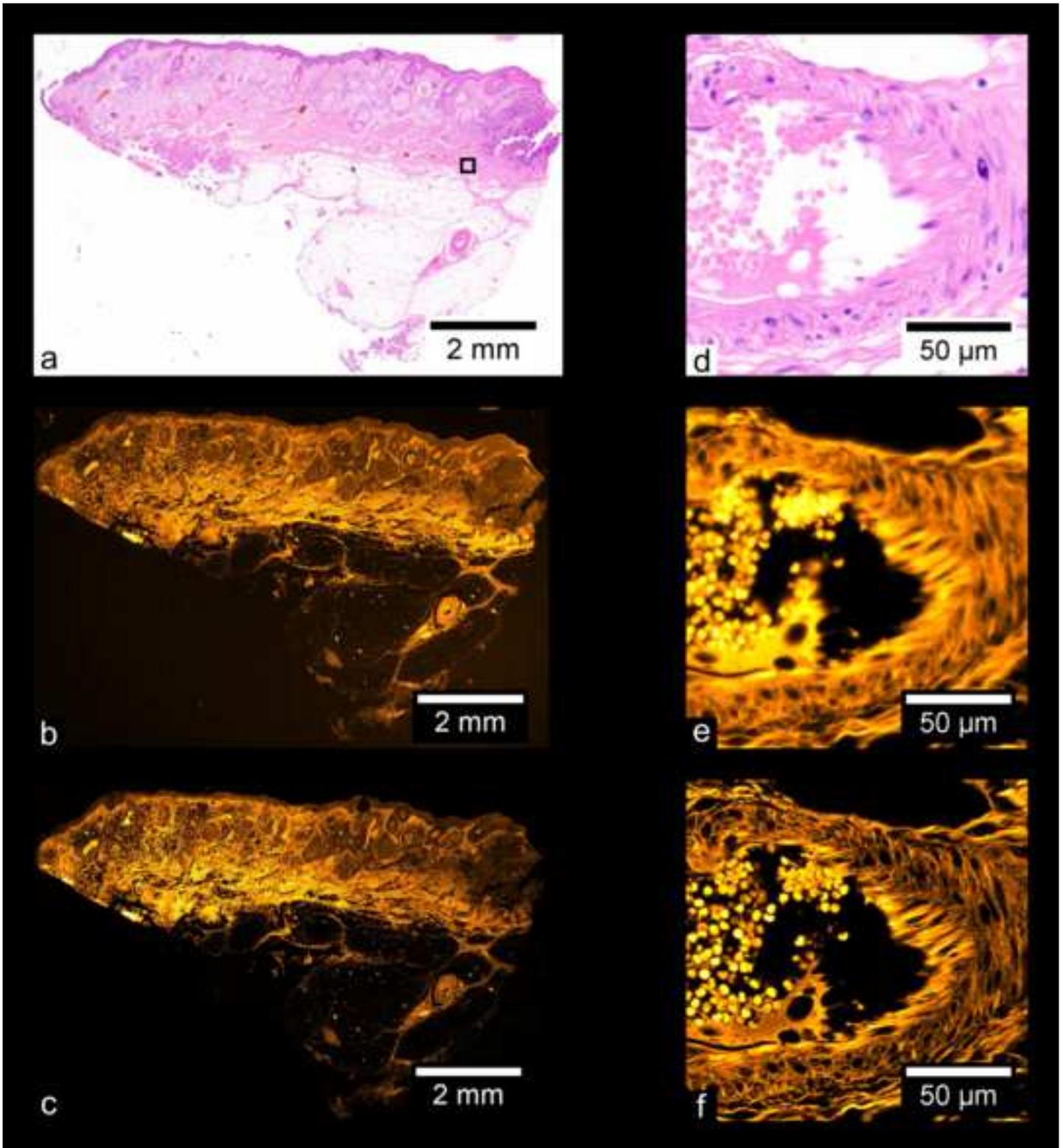
471 central region of the FFT's of (a) and (c), respectively. The arrows in (c) point to the fundamental peaks of the grid  
472 artifact. (e) shows plots of (c) and (d) through the center of each image along the x and y axes. The calculated value  
473 for  $T_{stitch}$  aligns very well with the peaks along both axes.

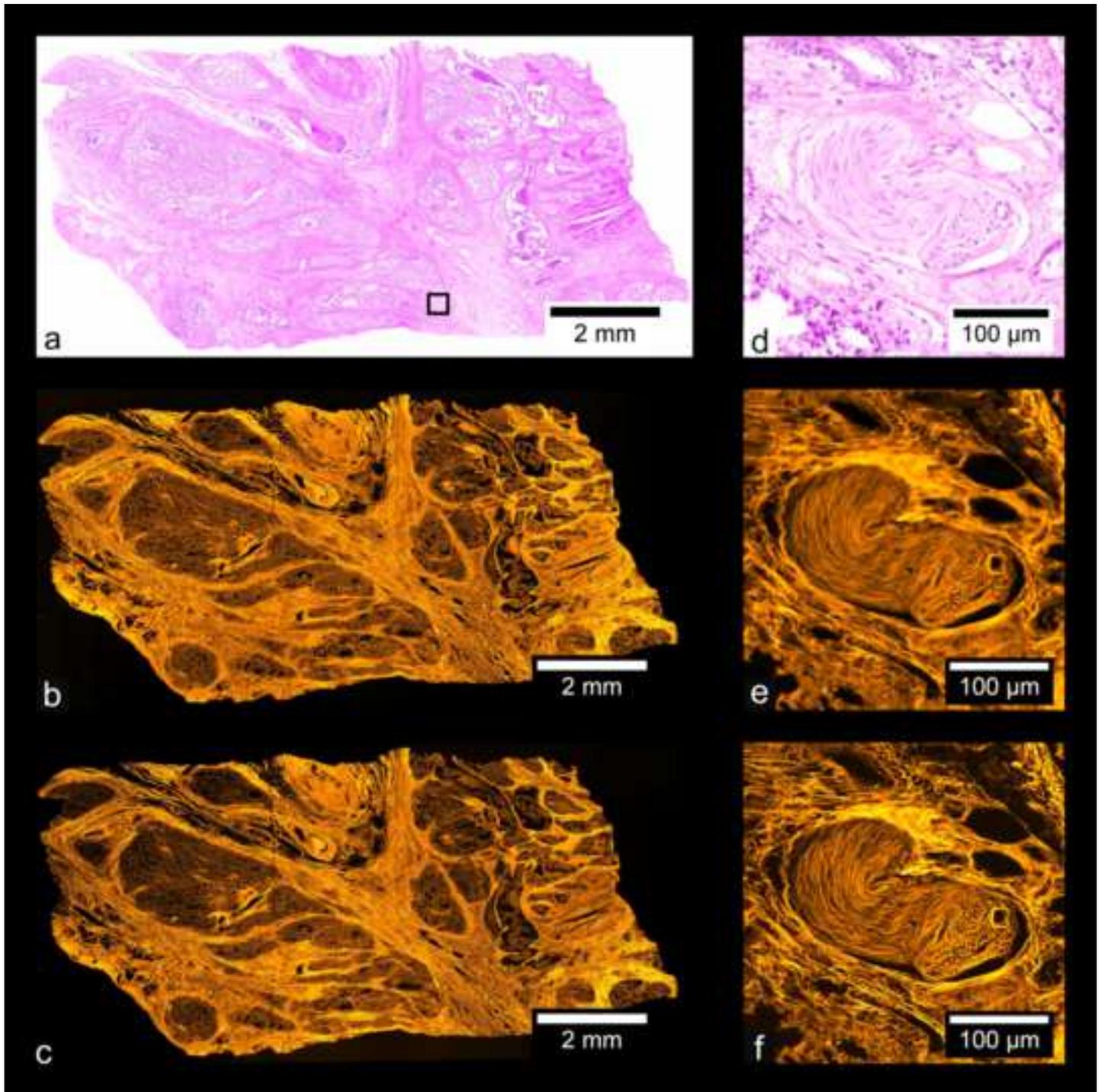


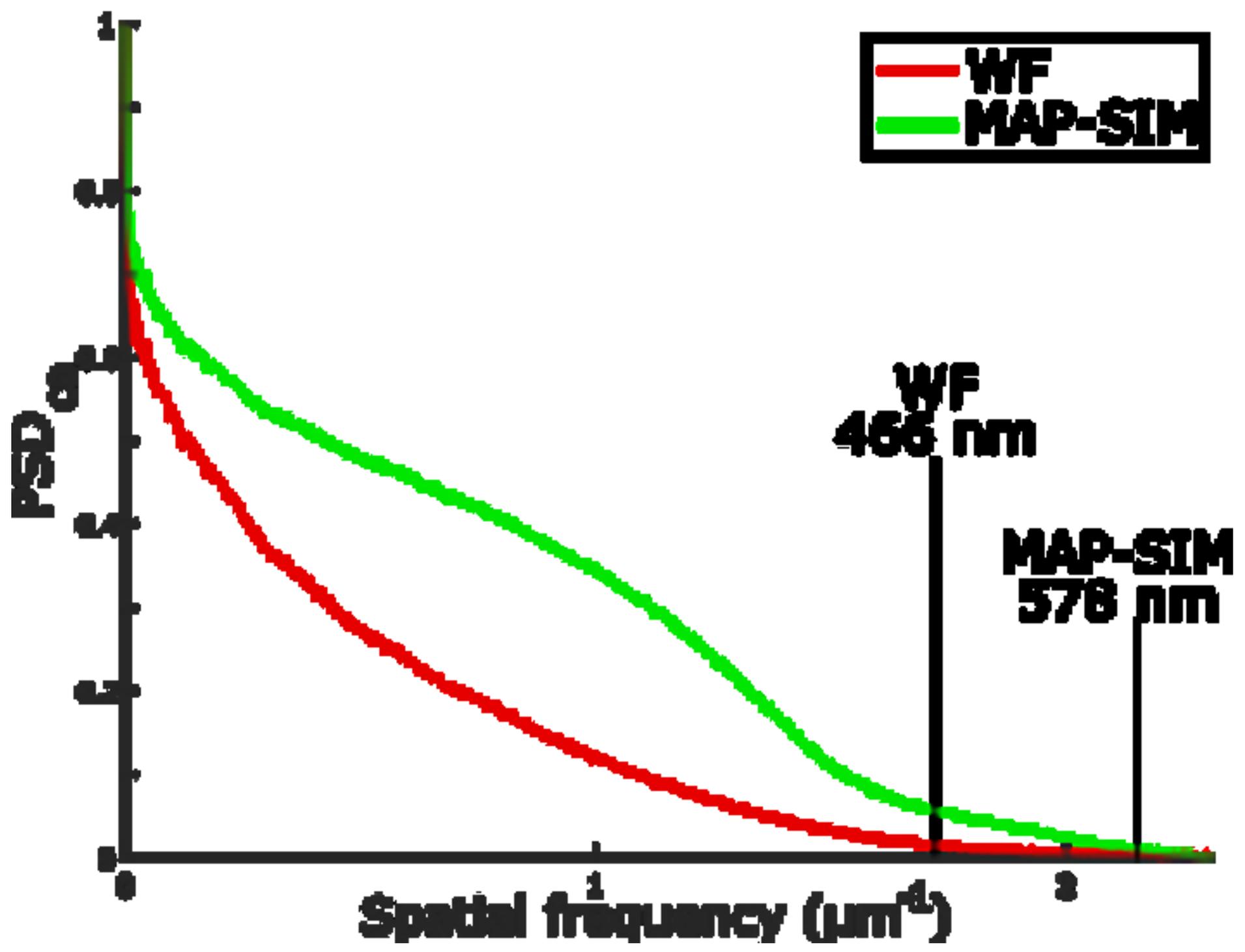


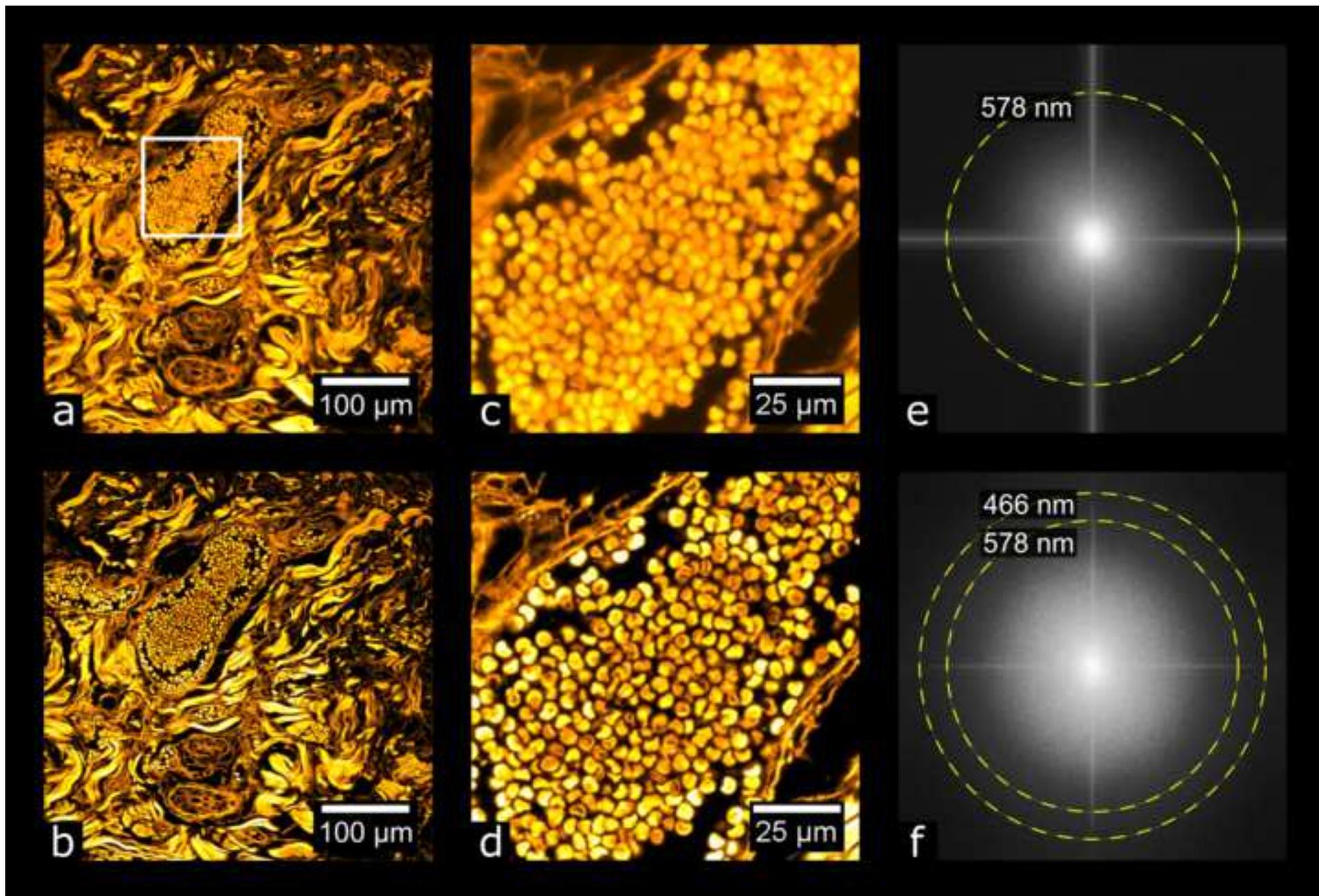


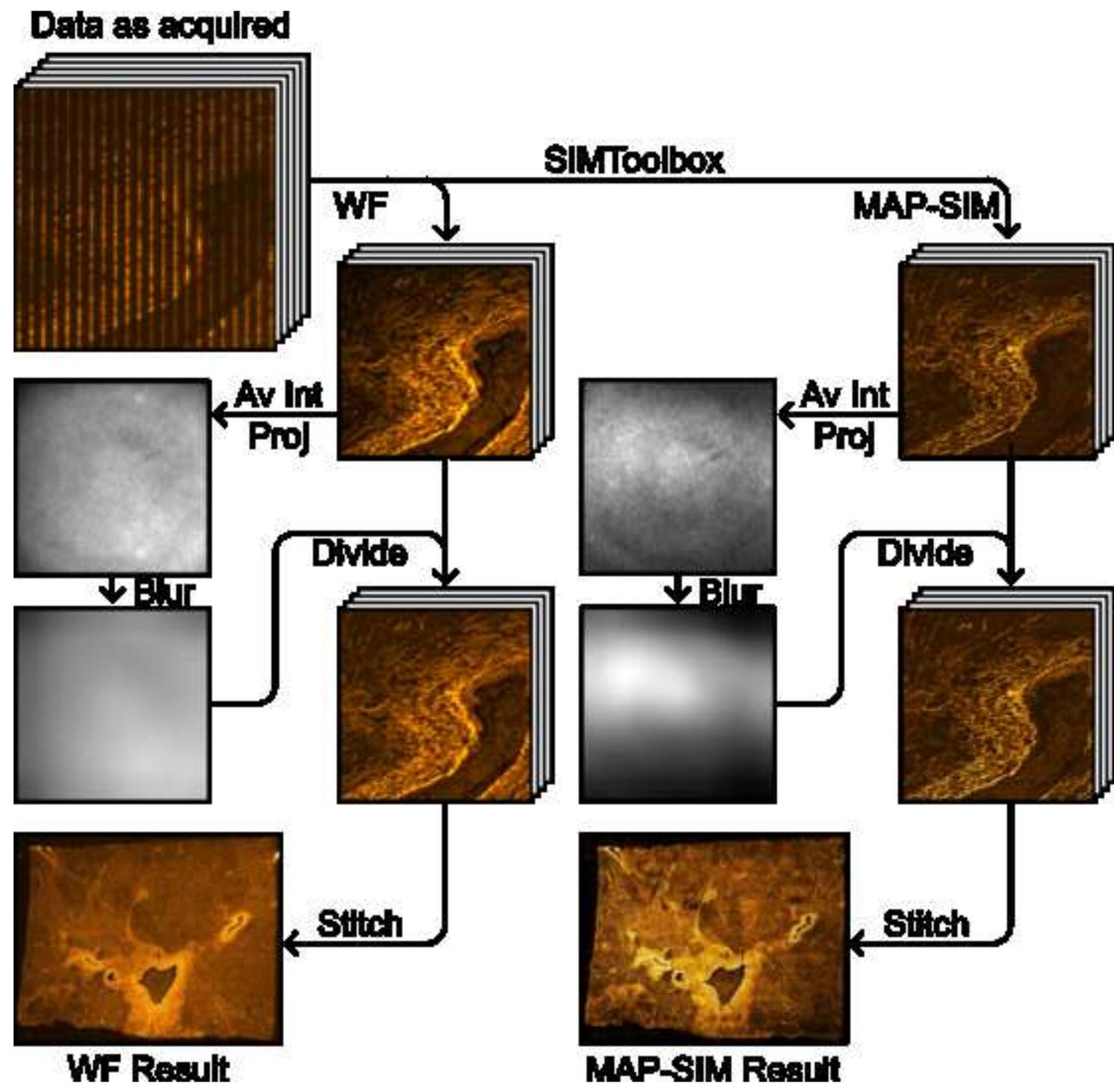


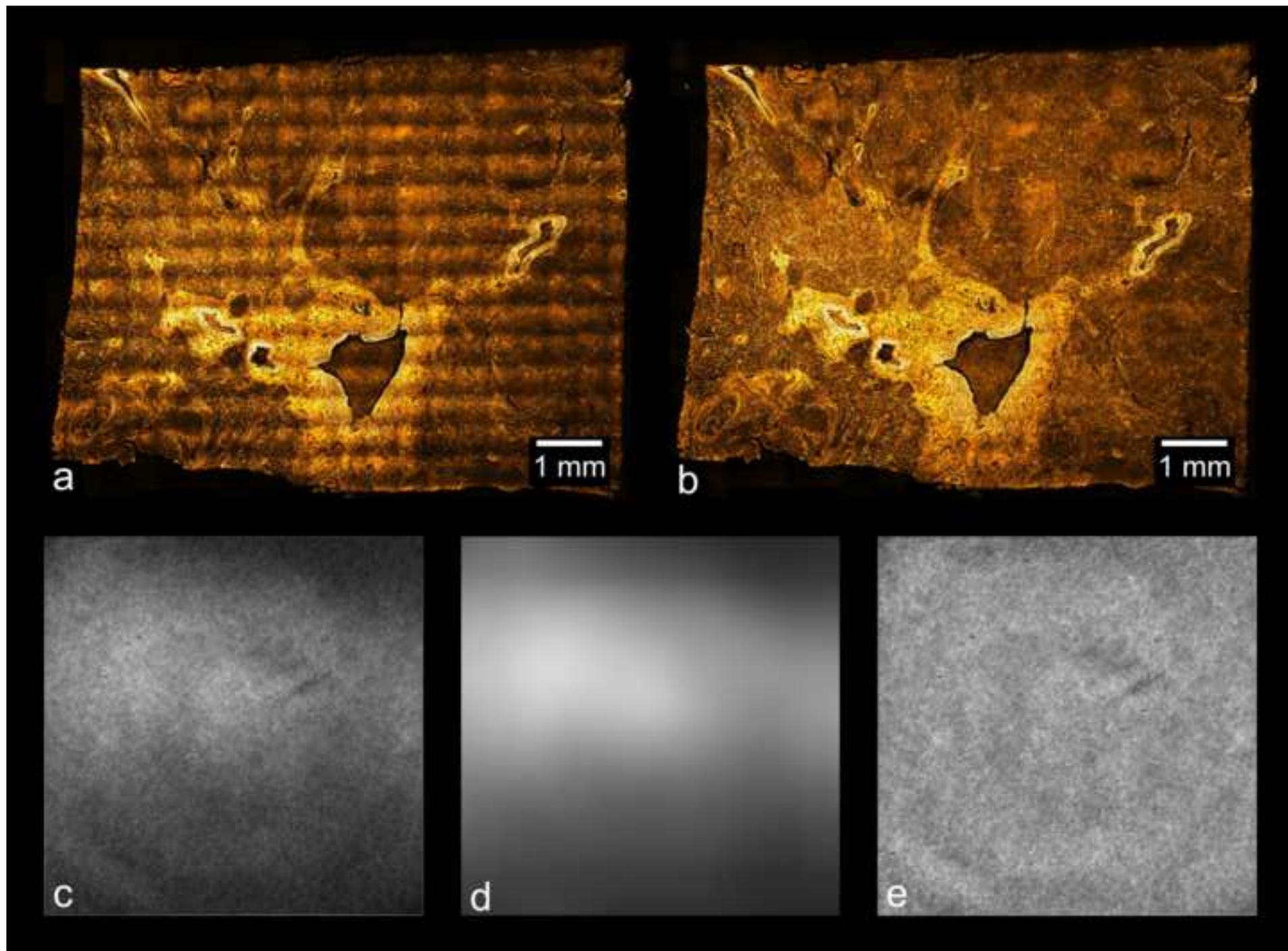


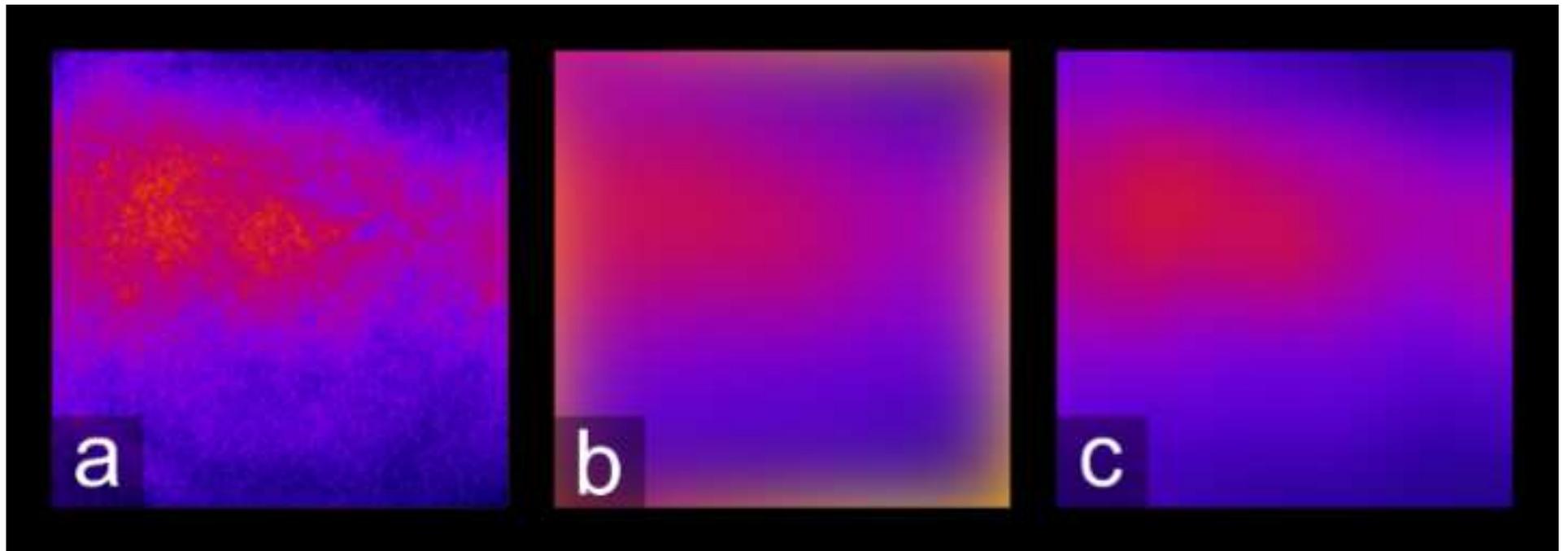




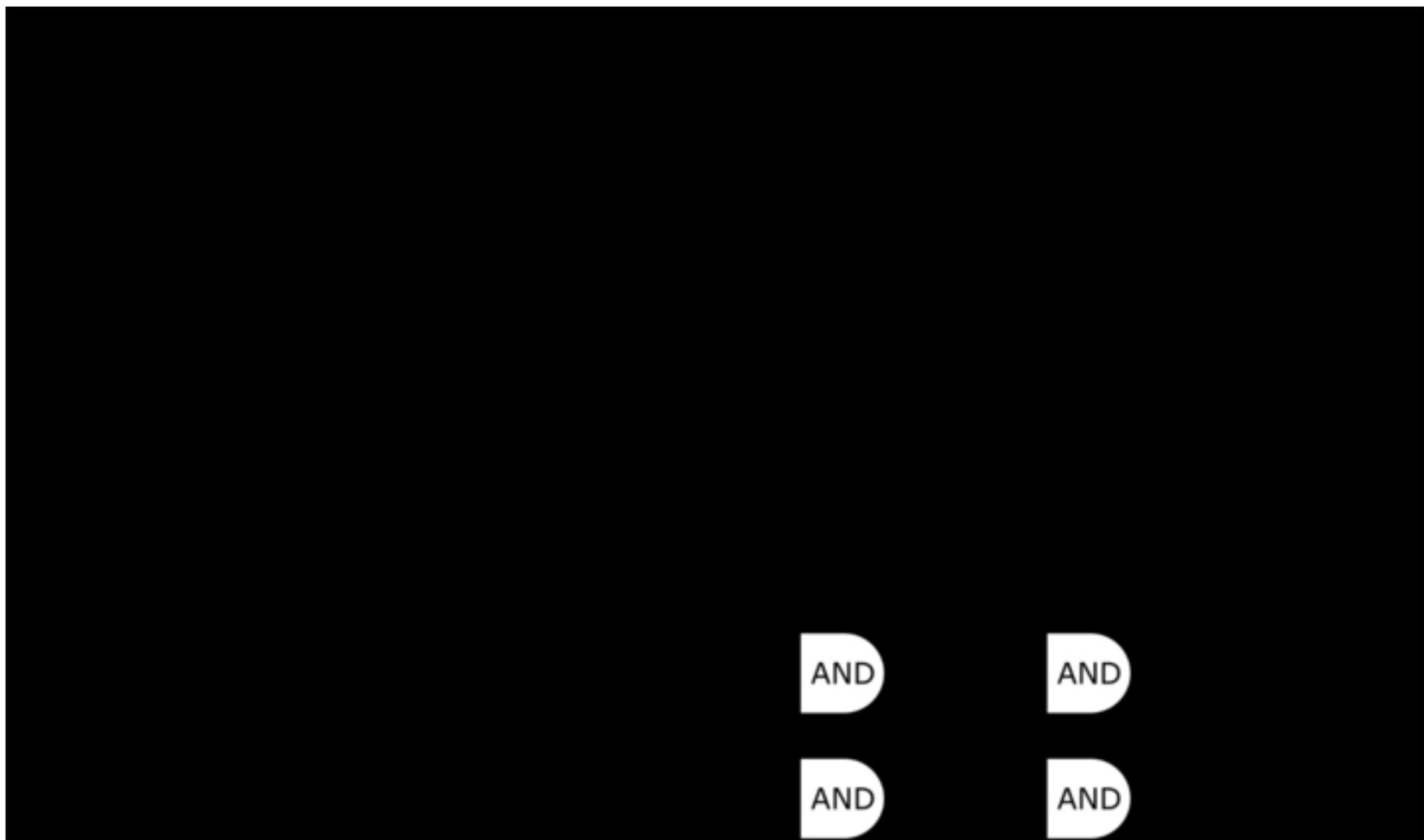


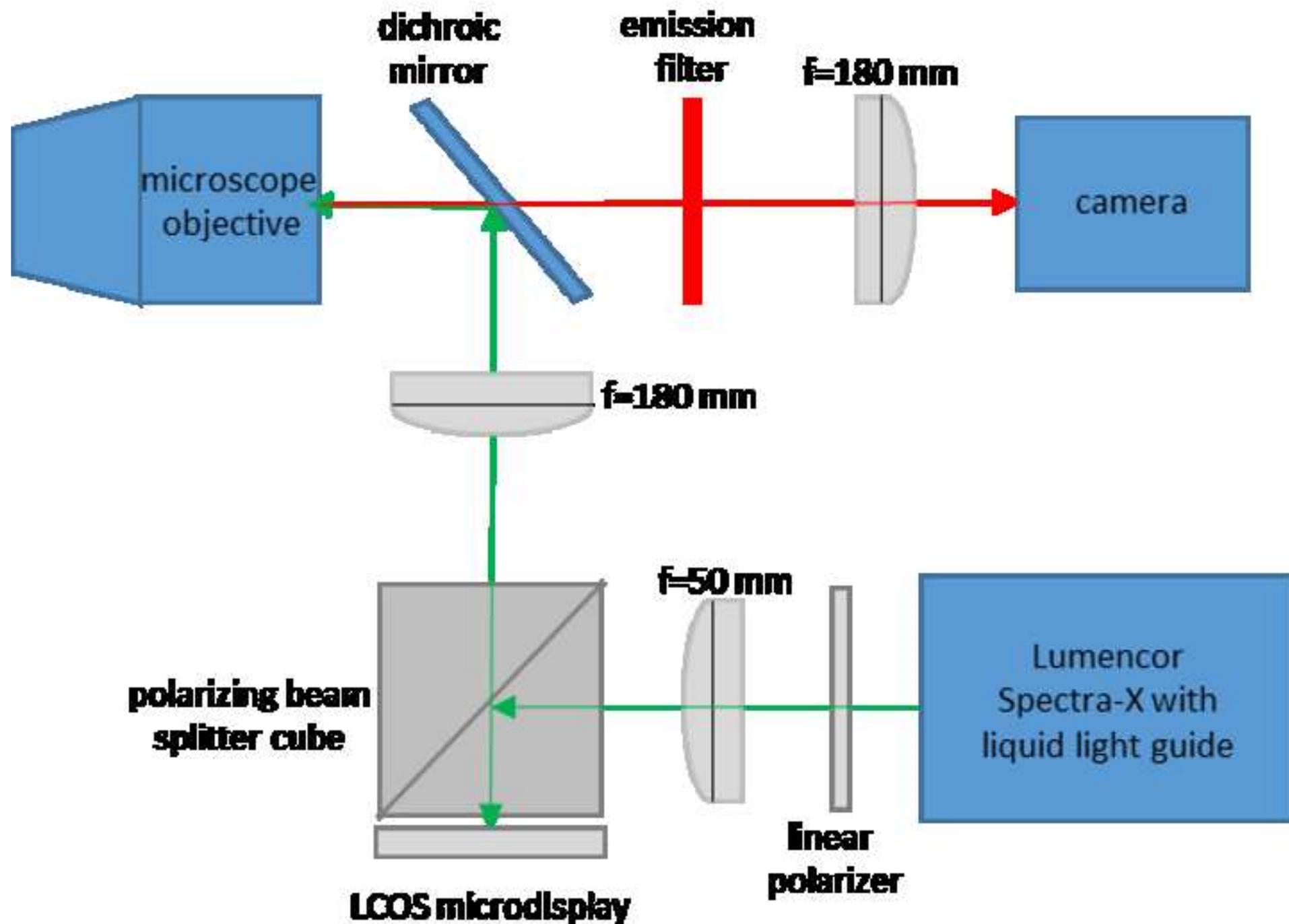














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Dear Editor:

We would like to re-submit our manuscript entitled “Artifact-free whole-slide imaging with structured illumination microscopy and Bayesian image reconstruction” for consideration in *GigaScience* as a data note. We made the following revisions.

1. The reviewer noticed that two of the data files was incomplete or corrupted. We reloaded these two files.
2. We included a table showing the organization of the data files along with their sizes.
3. We included a universal public domain dedication text file with the dataset.
4. We added a statement about ethics approval and consent to participate.
5. We moved all of the information from the supplement into the main paper. We removed the example of the cellular-scale imaging.

We hope that our paper will now be acceptable for publication in GigaScience.

Sincerely,



Guy M. Hagen