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

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Abstract:	<p>Background</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. Findings</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. Conclusion</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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<p>Experimental design and statistics</p> <p>Full details of the experimental design and statistical methods used should be given in the Methods section, as detailed in our Minimum Standards Reporting Checklist. Information essential to interpreting the data presented should be made available in the figure legends.</p> <p>Have you included all the information requested in your manuscript?</p>	<p>Yes</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.

Data description

24 **Context**

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

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

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

46 Fluorescence microscopy is becoming more important in histopathology. Traditional bright field
47 microscopy diagnostic methods require a time-consuming process, involving chemical fixation and physical
48 sectioning. The use of optical sectioning fluorescence microscopy allows high-quality images to be captured without
49 the need for physical sectioning. Consequently, it has been shown that imaging can be performed on large human
50 tissue samples within 1 hour after excision [36]. Additionally, other studies have shown the results of fluorescence

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

56 **Methods**

57 *Samples*

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

62 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

63 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 × 11	4×/0.16
Adenocarcinoma of Breast	6 × 6	4×/0.16
Lung Tuberculosis	8 × 10	10×/0.4

64 *Microscope setup and data acquisition*

65 We used a home-built SIM setup based on the same design as described previously [6,10,15] (Fig. 1). The
66 SIM system is based on an IX83 microscope (Olympus) equipped with a Zyla 4.2+ sCMOS camera (Andor) under
67 the control of IQ3 software (Andor). We used the following Olympus objectives: UPLSAPO 4×/0.16 NA,
68 UPLSAPO 10×/0.4 NA, LUCPLFLN 20×/0.45 NA, and UPLSAPO 30×/1.05 NA silicone oil immersion. For color
69 images we used an aca1920-40uc color camera (Basler) under control of Pylon software (Basler). We used a MS-
70 2000 motorized microscope stage (Applied Scientific Instrumentation) to acquire tiled SIM images. In all datasets,
71 the stage scanning was configured such that all image edges overlapped by 20%.

72 Briefly, the SIM system uses a ferroelectric liquid crystal on silicon (LCOS) microdisplay (type SXGA-
73 3DM, Forth Dimension Displays). This device has been used previously in SIM and related methods in fluorescence
74 microscopy [5,10,15,33,43–47] and allows one to produce patterns of illumination on the sample which can be
75 reconfigured at will by changing the image displayed on the device. The light source (Lumencor Spectra-X) is
76 toggled off between SIM patterns and during camera readout. Close synchronization between the camera
77 acquisitions, light source, and microdisplay ensures rapid image acquisition, helps reduce artifacts, and reduces light
78 exposure to the sample. The supplementary material contains more information about system integration.

79 **INSERT FIGURE 1**

80 *SIM data processing*

81 SIM reconstructions were performed using SIMToolbox, an open-source and freely available program that
82 our group developed for processing SIM data [33]. We generated optically sectioned, enhanced resolution images
83 using a Bayesian estimation method, maximum *a posteriori* probability SIM (MAP-SIM) [15]. MAP-SIM works
84 using maximum *a posteriori* probability methods, which are well known in microscopy applications [48,49], to
85 enhance high spatial frequency image information. We then combine this information, in the frequency domain,
86 with low spatial frequency image information obtained by OS-SIM methods, then produce the final image by an

87 inverse Fourier transform [15]. We typically measure the final resolution obtained by analyzing the frequency
88 spectrum of the resulting image, as is discussed below.

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

$$92 \quad I_{WF} = \frac{1}{N} \sum_{n=1}^N I_n,$$

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

95 *Vignetting correction*

96 Following SIM reconstruction, vignetting artifacts remain in each tile. If not removed prior to stitching, this
97 vignetting introduces a distracting grid pattern in the final stitched image. We performed vignette removal by
98 dividing each tile of the mosaic by an image representing the vignetting profile common to all tiles. Other studies
99 have used an image of a uniformly fluorescent calibration slide as a reference for vignette removal [36], where
100 information concerning non-uniform illumination is captured. However, we found that SIM processing introduces
101 vignetting artifacts beyond those due to non-uniform illumination. Additionally, these artifacts vary depending on
102 properties of the sample being imaged. As such, performing pre-acquisition calibration on a uniformly fluorescent
103 slide is not sufficient to remove vignetting artifacts from SIM reconstructions. Instead, an estimate of the vignetting
104 profile is found through analysis of the mosaic tiles after SIM reconstruction.

105 A blurred average intensity projection of the tiles is a good approximation of the vignetting profile, as an
106 average intensity projection merges the tiles into a single image with reduced foreground information while
107 preserving vignetting. Subsequent blurring with an appropriate radius and edge-handling method also eliminates the
108 high spatial frequency foreground without impacting the low spatial frequency illumination profile. To eliminate
109 errors during the blurring step due to the blurring area extending outside the original image, we used an edge
110 handling method in which the blurring area is reduced near the edges of the image such that no values outside the
111 image border are sampled. Unlike edge handling methods in which the image is padded with a uniform value (or

112 mirrored and tiled) to accommodate a blurring area which extends beyond the original image limits, this method is
113 free from major artifacts, such as erroneous brightness of the image edges (see supplementary figure S1). This
114 approximation of the illumination profile works especially well for histological samples, as such samples are non-
115 sparse and require many tiles, factors which improve the accuracy this approach. We performed all steps of this
116 devignetting process using built-in functions and the ‘Fast Filters’ plugin in ImageJ [50]. The effect of devignetting
117 is illustrated in Fig. 2.

118 **INSERT FIGURE 2**

119 *Image Stitching*

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

123 The data processing procedure is summarized in Fig. 3. The total time for processing each dataset was
124 about 30 min.

125 **INSERT FIGURE 3**

126 *Color image data processing methods*

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

133 *Resolution measurement*

134 We evaluated our results by measuring image resolution using SR Measure Toolbox. SR Measure
135 Toolbox [51] measures the resolution limit of input images through analysis of the normalized, radially averaged
136 power spectral density (PSD_{ca}) of the images, as previously described [6]. Briefly, the resolution limit in real space

137 is determined by evaluating the cutoff frequency in Fourier space. The cutoff frequency is estimated by calculating
138 the spatial frequency at which the PSD_{ca} (after noise correction) drops to zero.

139 Focusing on the basal cell carcinoma sample, we selected 125 (out of 522 total) image tiles, calculated the
140 PSD and resolution for each tile, and averaged the results. We found that, in the case of this sample, the image
141 resolution was 593 ± 20 nm for WF and 468 ± 2.5 nm for MAP-SIM (average \pm standard deviation). This data was
142 acquired with a UPLSAPO 30 \times /1.05 NA silicone oil immersion objective. Figure 4 shows an example measurement
143 for one image tile. Figure 5 shows a plot of PSD_{ca} for this image tile.

144 **INSERT FIGURE 4**

145 **INSERT FIGURE 5**

146 **Results**

147 Figure 6 shows images of a prepared slide containing a human prostate carcinoma sample stained with
148 H&E. Fig. 6a shows a stitched color overview, and Fig. 6d shows a zoom-in of the region indicated in Fig. 6a,
149 acquired separately using a UPLSAPO 20 \times /0.75NA objective. Fig. 6b shows a stitched widefield fluorescence
150 image, and Fig. 6c shows a stitched SIM image. Figs. 6e and 6f each show zoom-ins of the stitches shown in Figs.
151 6b and 6c, respectively. Using the acquisition and processing methods described, whole-slide images are produced
152 without any visible stitching artifacts. Additionally, the MAP-SIM reconstruction method produces resolution
153 superior to that of the widefield data.

154 Figures 7-10 show similar comparisons for basal cell carcinoma, ovary adenocarcinoma, breast
155 adenocarcinoma, and tuberculosis of the lung, respectively.

156 The data shown in figures 6-10 is freely available through Giga DB [reference to be added]. This dataset
157 includes all color overviews as well as WF and MAP-SIM stitches at full resolution. In addition, all image tiles
158 (prior to deconvolution) used to create the WF and MAP-SIM stitches of the basal cell carcinoma sample are
159 provided.

160 **INSERT FIGURE 6**

161 **INSERT FIGURE 7**

162 **INSERT FIGURE 8**

163 **INSERT FIGURE 9**

164 **INSERT FIGURE 10**

165 **Discussion**

166 Many past studies into stitching of SIM mosaics have suffered from noticeable image artifacts, arising from
167 flaws in the optical setups used as well as imperfections in the SIM reconstruction and image stitching processes.
168 While these artifacts are sometimes minimal enough to remain uncorrected, certain artifacts seriously inhibit the
169 usefulness of the final stitched image. In [23], the authors note that issues in triggering and evenly illuminating the
170 microdisplay being used for illumination resulted in striping and vignetting artifacts; similarly, in [22,24,36,52],
171 stitching artifacts are apparent in the images. Here, optimization of the optical setup, camera-microdisplay
172 synchronization, and image processing methods yielded whole-slide images free from visible SIM or image stitching
173 artifacts. In addition to the elimination of artifacts, our use of SIMToolbox to perform SIM reconstruction on the
174 data allows for a variety of reconstruction algorithms to be used, including super-resolution algorithms such as
175 MAP-SIM. This too presents an improvement over previous works. Our methods also allow for stitching of high-
176 magnification tiles into large-FOV images with subdiffractive detail (see supplementary Fig. S3).

177 Another advantage of the acquisition and processing methods demonstrated here is the minimization of
178 user intervention, and in turn, reductions in acquisition and processing time. Firstly, the use of Andor IQ during
179 acquisition allows for stage movement, sample focusing, image acquisition, and SIM pattern advancement to be
180 controlled automatically. Loading of the sample, definition of the mosaic edges, and manual focus on 3-5 positions
181 of the sample are the only steps needed to be taken by the user before acquisition can begin. Recent developments in
182 autofocus technology for SIM may allow for the manual focus step to be shortened or omitted [52]. These automated
183 steps during acquisition allow for large mosaics to be acquired. The quality of the final stitched images does not
184 degrade for larger mosaics – in fact, the quality of the devignetting process improves with larger datasets, as more
185 data is available to produce an accurate estimation of the illumination profile. SIMToolbox (version 2.0), which is
186 capable of utilizing the processing power of modern consumer graphics cards during MAP-SIM processing, also
187 reduces the time spent during the data processing phase. Finally, unlike other super-resolution reconstruction

188 methods such as SR-SIM, MAP-SIM is able to produce artifact-free results without tuning of reconstruction
189 parameters by the user, a process which is difficult to automate and requires significant user experience.

190 One drawback the method presented here is the inability to image the entire volume of samples thicker than
191 ~0.5 mm. However, this limitation does not prevent large, unsectioned samples from being imaged, as is the case
192 with bright field microscopy, where samples must be thin enough for transmitted light to reach the objective. Rather,
193 as the light which illuminates the sample in fluorescence microscopy emanates from the objective, all surface
194 regions of a large sample may be imaged. Additionally, due to the optical sectioning exhibited by SIM, light from
195 out-of-focus regions of the sample is almost completely attenuated. Consequently, imaging the surfaces of large
196 samples with SIM produces high-contrast images of thin regions without the need for physical sectioning, as
197 previously demonstrated [23,36].

198 Here, we demonstrated our imaging techniques on traditionally prepared histopathological samples in order
199 to provide a comparison between bright field imaging and SIM, but the same techniques can be used to image a
200 wide variety of fluorescently labelled samples, as demonstrated in the supplementary material. The ability to
201 seamlessly image the entire surface region of large samples has multiple potential applications in histopathology.
202 SIM presents unique advantages in analyzing the surgical margins of large tissue excisions, as demonstrated by
203 Wang [36]. Briefly, due to the ability of SIM to image an unsectioned sample, analysis of surgical margins using
204 SIM requires imaging of far less surface area than that needed for bright field imaging. Confocal imaging of core
205 needle biopsy samples has been previously demonstrated to produce images suitable for medical diagnosis [42], a
206 practice easily adapted to SIM. The speed at which sample preparation and image acquisition can be performed in
207 fluorescence microscopy presents opportunities for intra-operative analysis of tissue samples using SIM techniques,
208 as mentioned by multiple other studies [23,36,53,54].

209 **Reuse potential**

210 The data provided here presents various opportunities for reuse. Firstly, the unstitched image tiles provided
211 in the dataset, which still contain vignetting artifacts, may be used to reproduce the results of our devignetting
212 process, as well as to further develop more sophisticated devignetting approaches suited for SIM. These tiles might
213 also be used to create or modify existing stitching software for global minimization of stitching artifacts. For
214 example, the frequency-domain detection of periodic stitching artifacts discussed in the supplementary material

215 could be used to minimize such artifacts in developing new stitching software. With the multiple high-resolution
216 color overviews and stitched SIM images, comparison of structures visible in the brightfield and fluorescent images
217 could be performed to further study the use of fluorescence microscopy in histopathology.

218 **Availability of source code and requirements**

219 Project name: SIMToolbox version 2.12

220 Project home page: <http://mmtg.fel.cvut.cz/SIMToolbox/>

221 Operating system: platform independent

222 Programming language: MATLAB

223 License: GNU General Public License v3.0

224 **Detailed software compatibility notes**

225 The SIMToolbox GUI was compiled with MATLAB 2015a and tested in Windows 7 and 8. The GUI is a stand-alone
226 program and does not require MATLAB to be installed. To use the MATLAB functions within SIMToolbox (i.e.,
227 without the GUI), MATLAB must be installed. The functions were mainly developed with 64 bit MATLAB versions
228 2012b, 2014a, 2015a in Windows 7. When using SIMToolbox functions without the GUI, the MATLAB “Image
229 Processing Toolbox” is required. SIMToolbox also requires the “MATLAB YAML” package to convert MATLAB
230 objects to/from YAML file format. Note that this package is installed automatically when using the GUI.

231 **Availability of data**

232 All raw and analyzed data is available on GigaDB at <http://gigadb.org/site/index>.

233 **Abbreviations**

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

237 **Ethics approval and consent to participate**

238 Not applicable

239 **Consent for publication**

240 Not applicable

241 **Competing interests**

242 The authors declare that they have no competing interests.

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251 **Author Contributions**

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

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

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390 **FIGURE CAPTIONS**

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

392 Figure 2: Vignetting artifacts and their removal. (a) shows the result of stitching images without applying the
393 devignetting process, while (b) shows a stitch of the same data after devignetting has been applied. (c) shows the
394 average intensity projection of the images used to stitch (a), which estimates the vignette profile of each frame. This
395 estimate can be refined by application of an edge-limited blurring filter, as shown in (d). (e) shows the average
396 intensity projection of the data used in (b), after devignetting has been applied. The uniform brightness of (e)
397 indicates that no major vignetting artifacts remain in the devignetted data.

398 Figure 3. Panoramic SIM data processing workflow. Devignetting was performed after SIM reconstruction. Note the
399 vignette profile differs between reconstruction methods, necessitating separate projection, blurring and division
400 steps. Av Int Proj refers to average intensity projection.

401 Figure 4. Evaluating image resolution. (a) and (b) show a tile from the data in Fig. 7 (basal cell carcinoma sample)
402 after widefield and MAP-SIM reconstruction, respectively. (c) and (d) each show a zoom-in of (a) and (b),
403 respectively. (e) and (f) each show the FFT of (a) and (b), respectively. The dotted lines in (e) and (f) indicate the
404 resolution of each image according to the resolution measurement described.

405 Figure 5. Normalized, radially averaged power spectral density (PSD_{ca}) and resolution analysis measured on the tiles
406 shown in Figs. 4a and 4b.

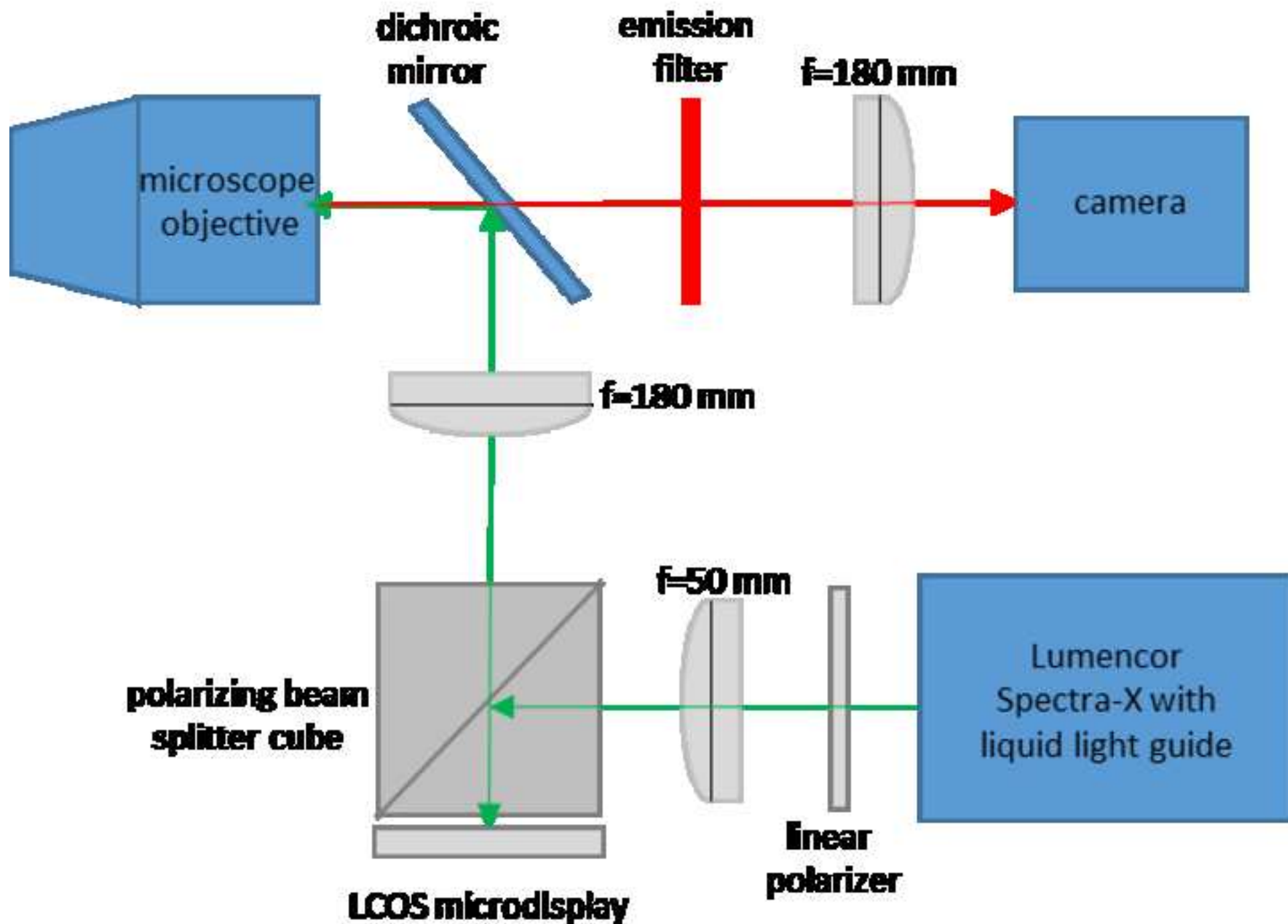
407 Figure 6: Carcinoma of human prostate. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a region
408 of the sample indicated in (a). (e) and (f) each show a zoom-in of (b) and (c), respectively, in the region indicated in
409 (a).

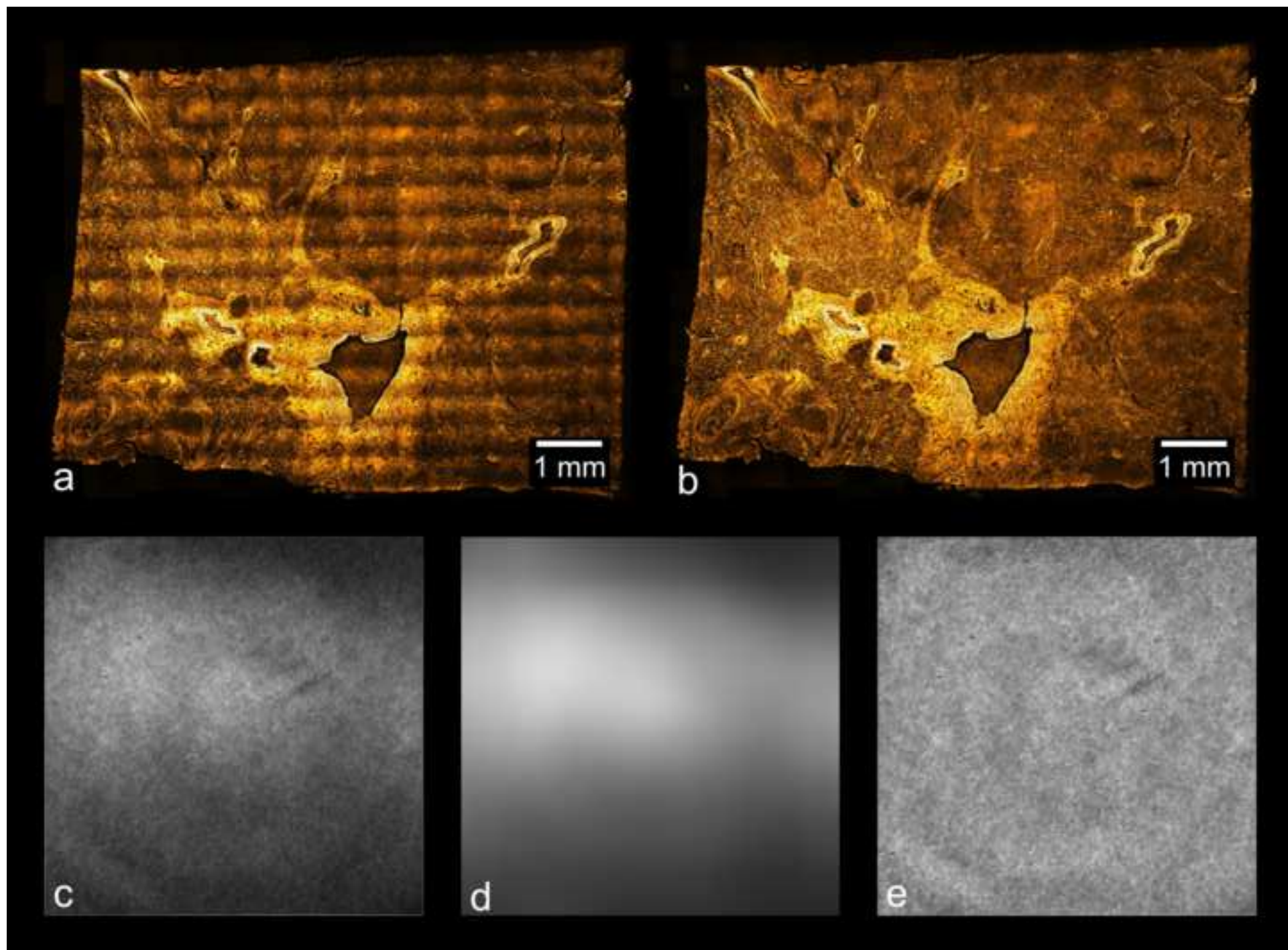
410 Figure 7: Basal Cell Carcinoma. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a region of the
411 sample indicated in (a). (e) and (f) each show a zoom-in of (b) and (c), respectively, in the region indicated in (a).

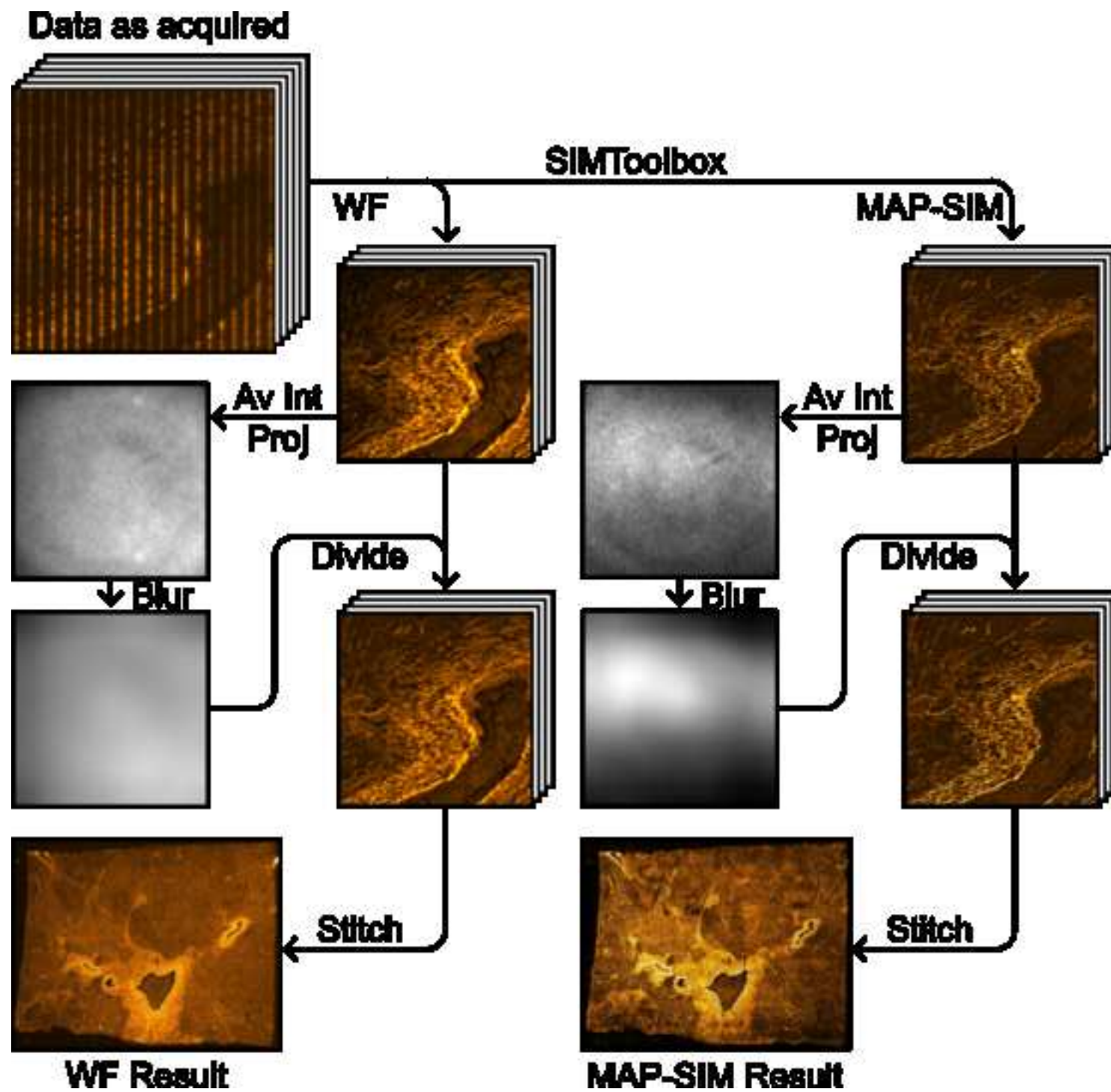
412 Figure 8: Adenocarcinoma of human ovary. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d), (g) show a
413 region of the sample indicated in (a), acquired separately from (a) using a 10× objective. (e) and (h) show a zoom-in
414 of (b), while (f) and (i) show a zoom-in of (c), all in the regions indicated in (a).

415 Figure 9: Adenocarcinoma of human breast. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a
416 region of the sample indicated in (a), acquired separately from (a) using a 10× objective. (e) and (f) each show a
417 zoom-in of (b) and (c), respectively, in the region indicated in (a).

418 Figure 10: Tuberculosis of human lung. (a) Color overview, (b) WF stitch, (c) MAP-SIM stich. (d) shows a region
419 of the sample indicated in (a), acquired separately from (a) using a 20× objective. (e) and (f) each show a zoom-in of
420 (b) and (c), respectively, in the region indicated in (a).







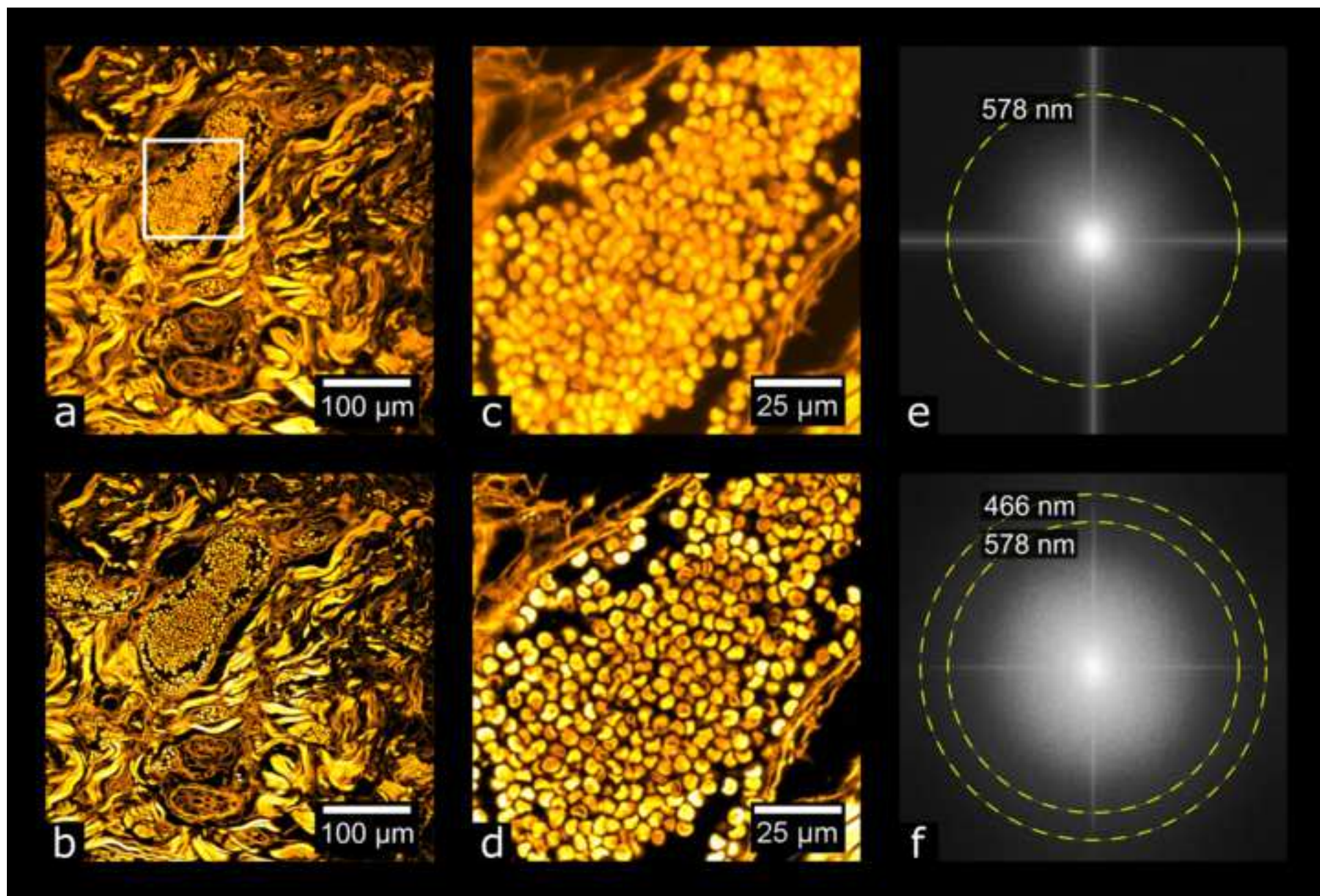
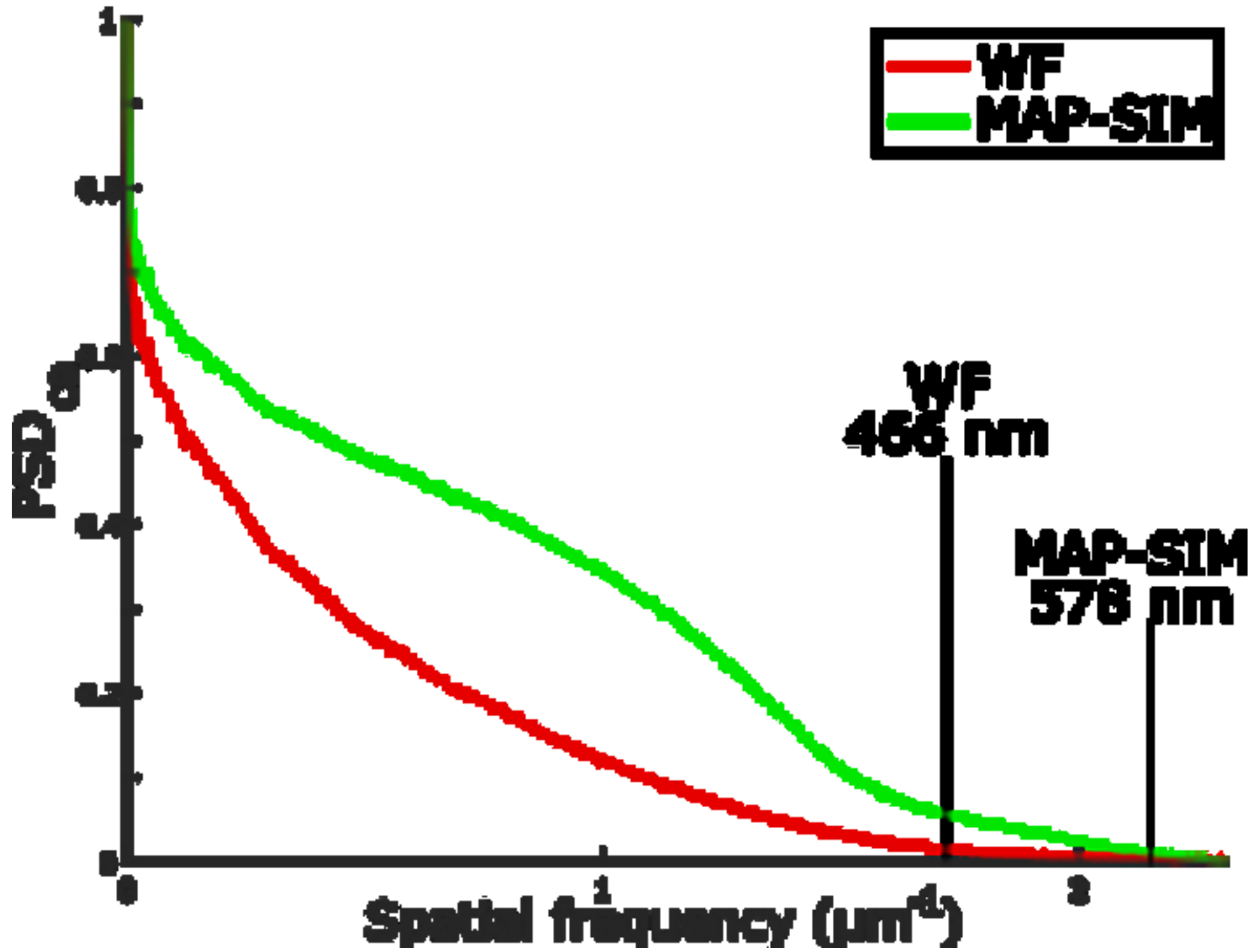
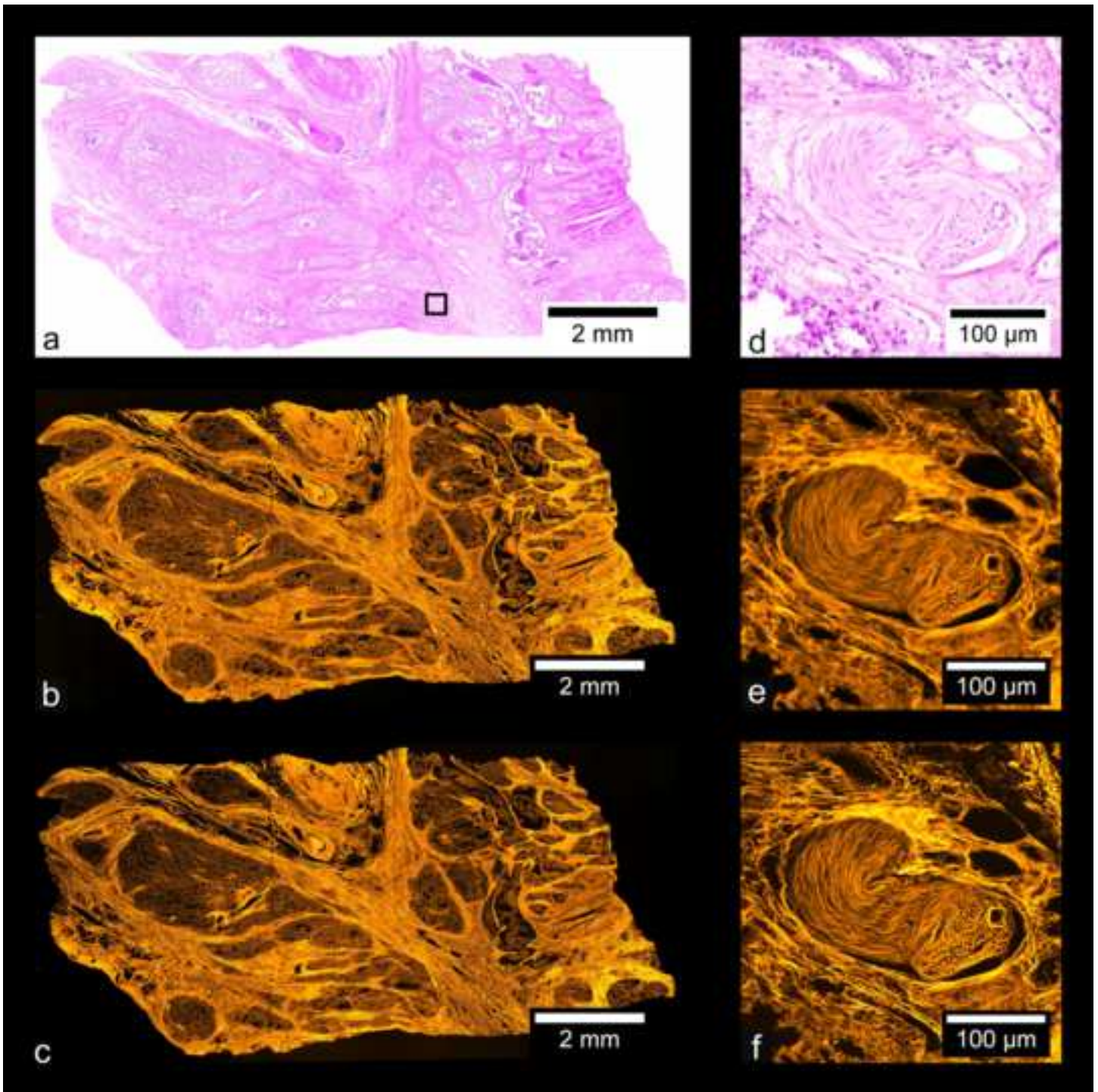
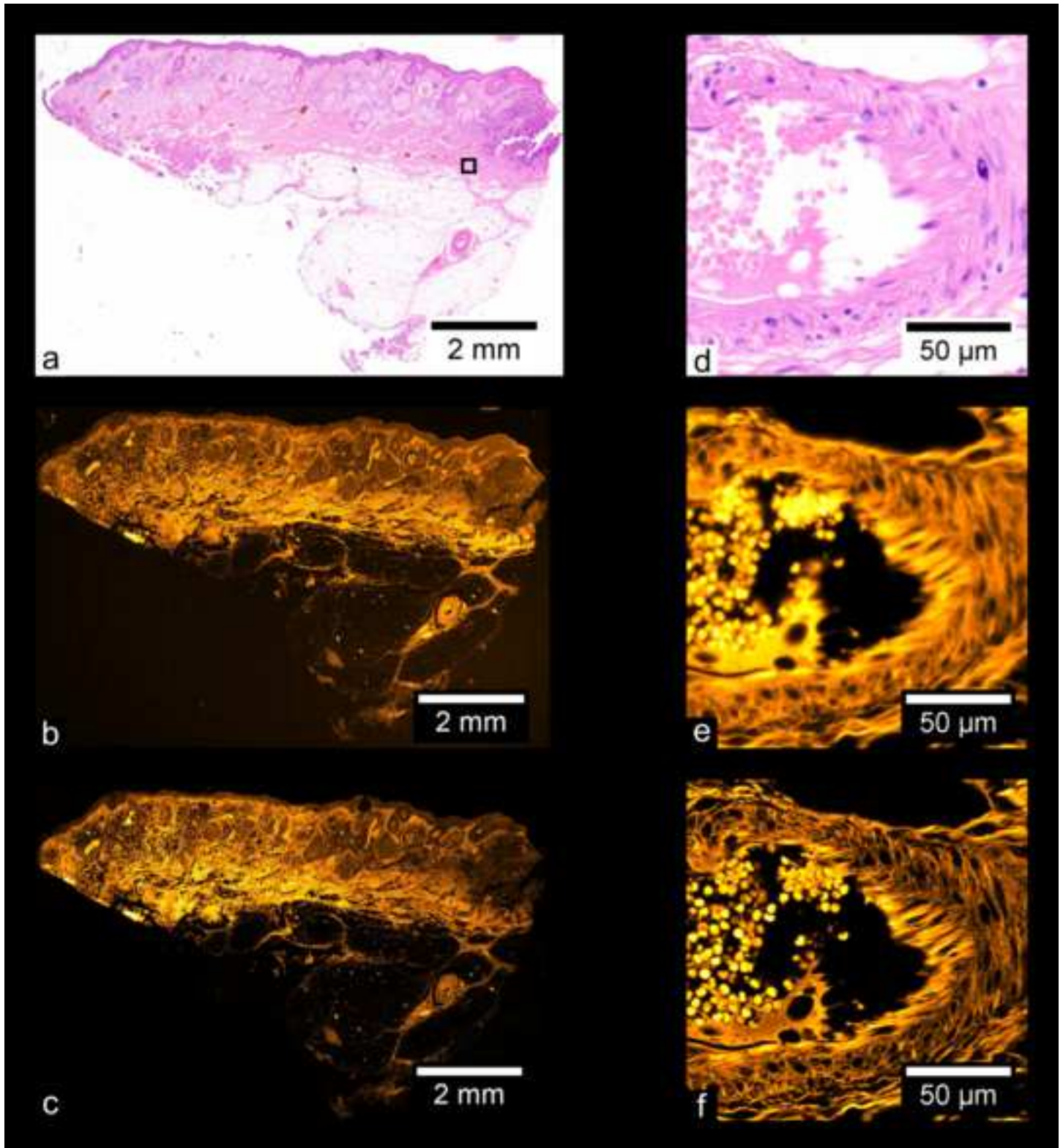
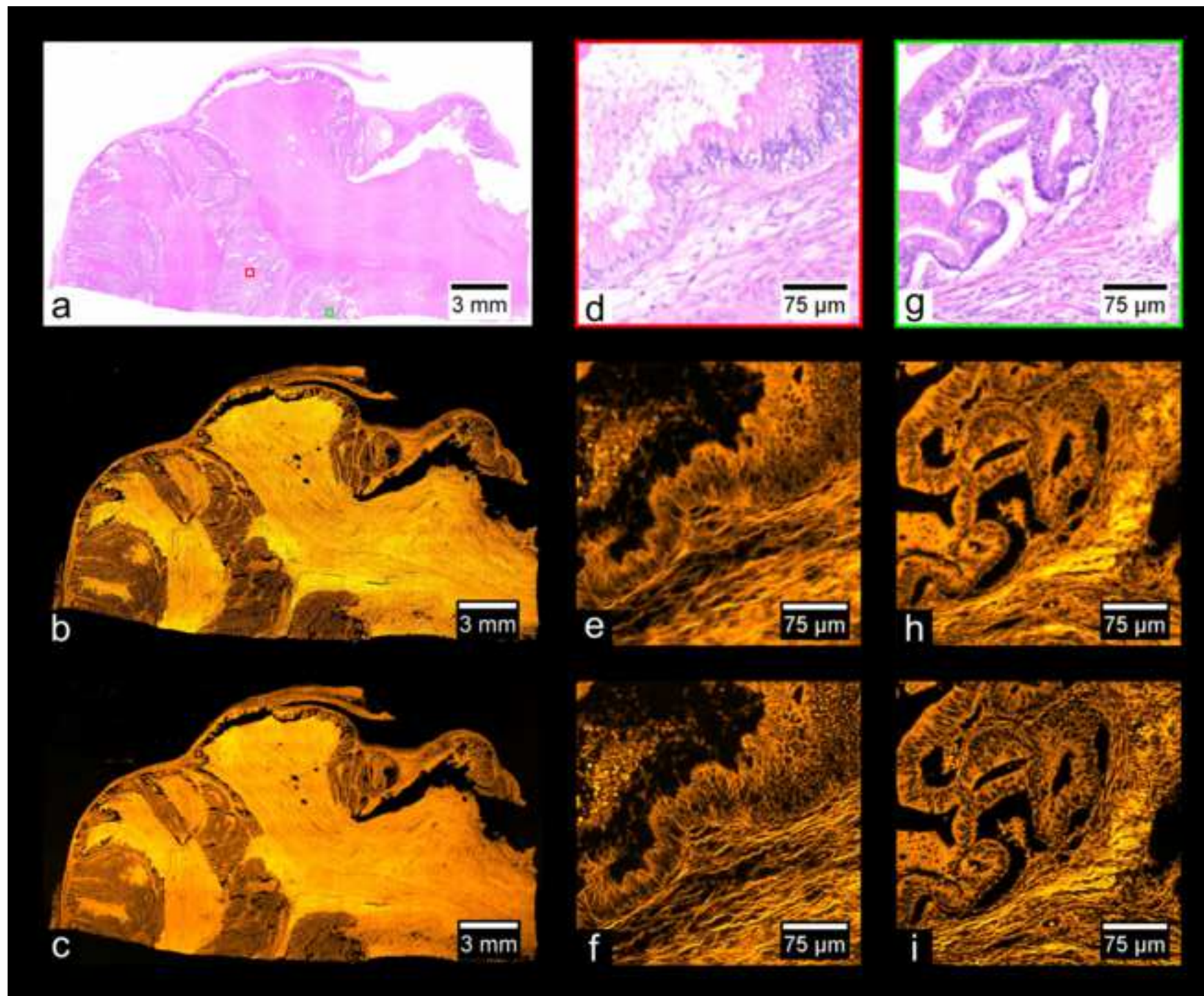


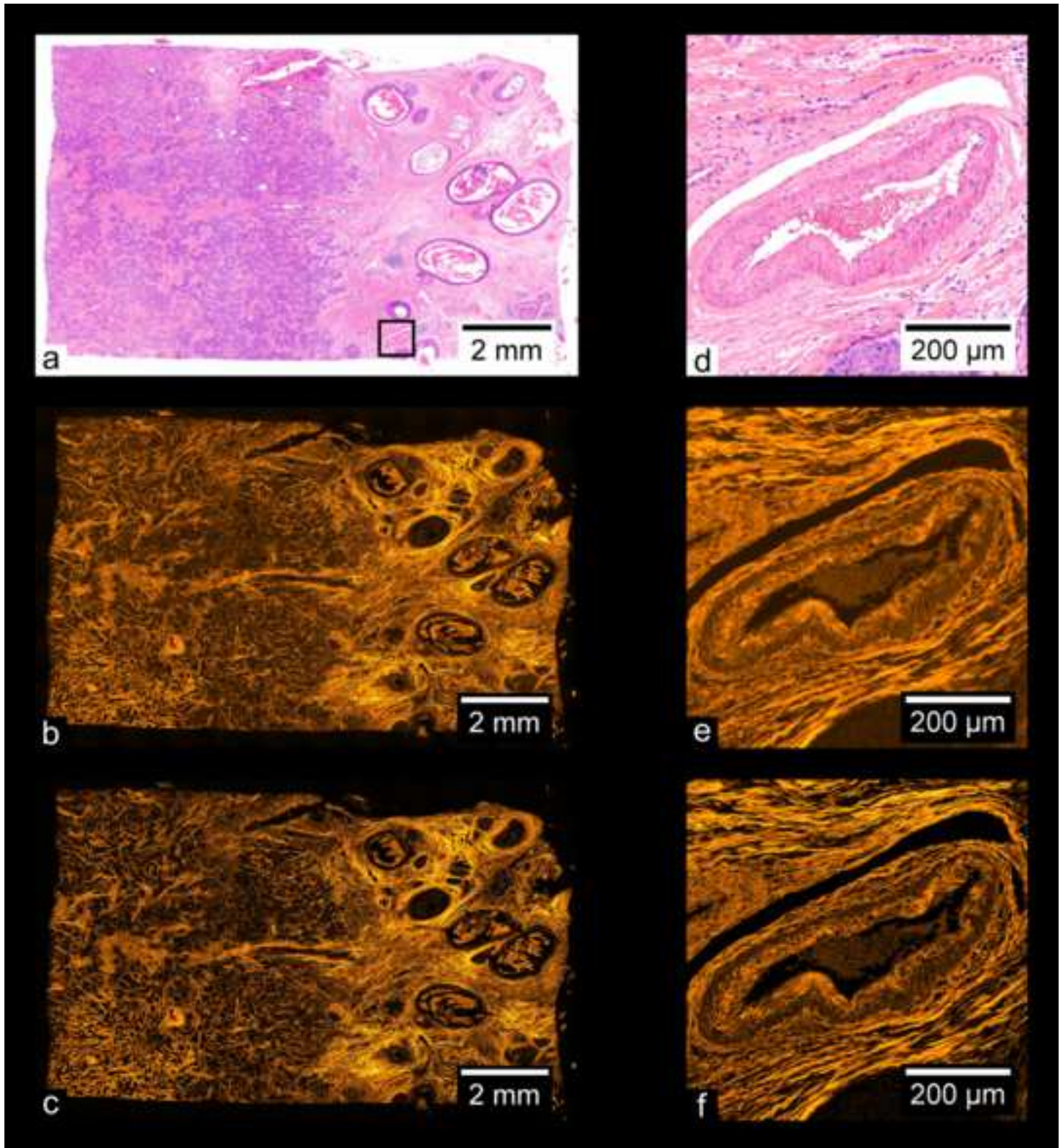
Figure 5

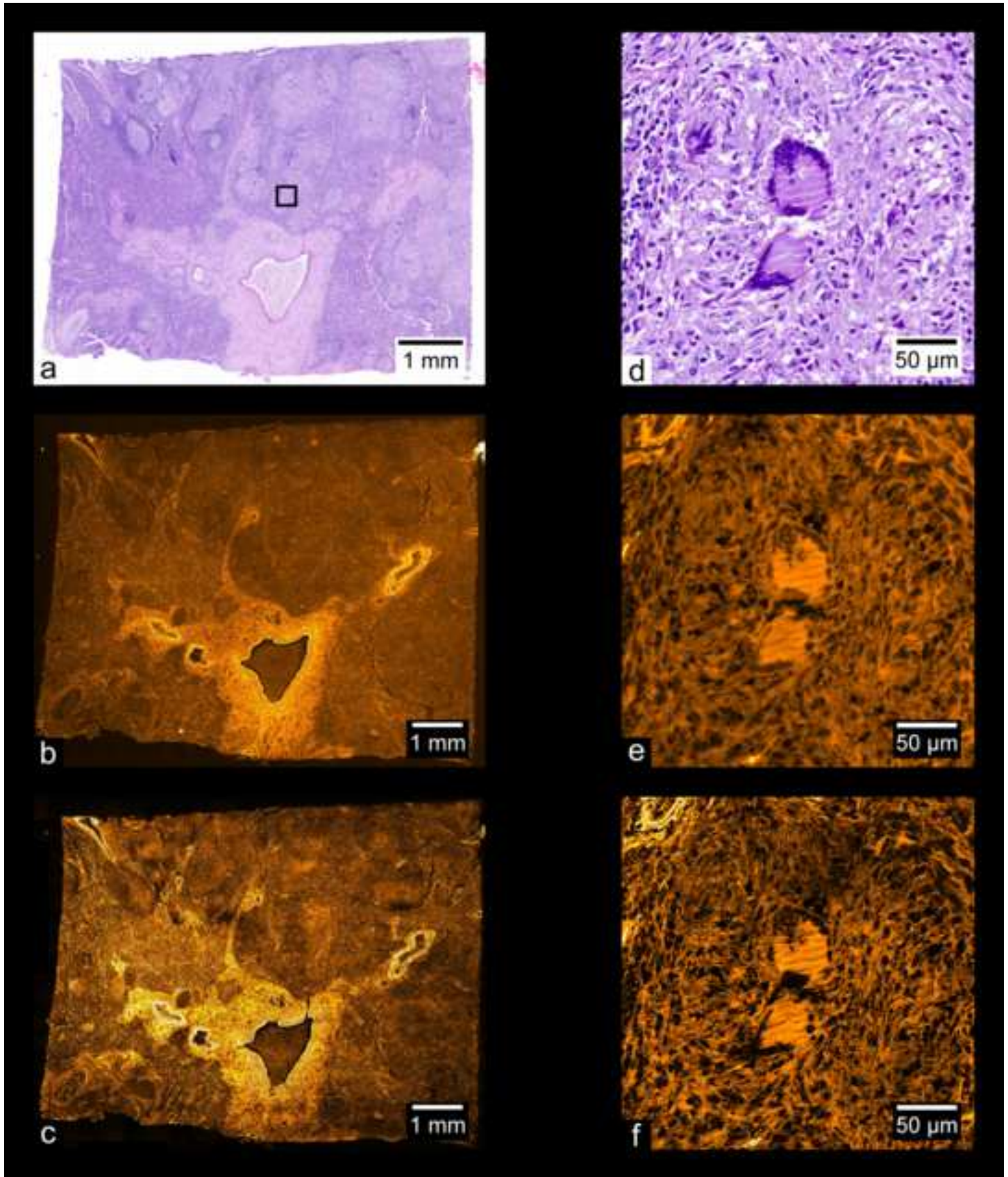














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Dear Editor:

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

Fluorescence microscopy is finding increased use in histopathology. Structured illumination microscopy (SIM) is a method which can produce fluorescence images with optical sectioning (similar to confocal microscopy), and/or with super-resolution. Methods in which a larger sample is imaged with subsequent stitching of image tiles is a popular approach, but previously published data often contains noticeable artifacts due to image stitching. This is often visible as dark bands surrounding each sub-image of a larger mosaic. Other artifacts can arise due to SIM methods as they are typically applied.

Here, we present datasets which do not contain such artifacts and the detailed methods we developed to produce them. We imaged human tissues under pathophysiological conditions, including cancers of the prostate, breast, skin, and ovary. The samples we used were acquired commercially (for example from Carolina Biological) and are standard histological preparations.

We plan to upload our datasets to the Giga-database after an invitation to do so from the Journal. These datasets include the un-stitched image tiles, as well as high quality, high resolution images of the samples after de-vignetting and stitching. These datasets will be useful for other researchers investigating SIM methods or image stitching approaches, and also for clinicians who are interested in fluorescence-based histopathology. These datasets can be rather large, which has prevented their publication up to now. We believe *GigaScience* offers an excellent venue for such publication.

We would like to suggest the following reviewers for the manuscript:

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Sincerely,

A handwritten signature in blue ink that reads "Guy Hagen" with a long, sweeping horizontal line extending to the right.

Guy M. Hagen