

# A MODEL-BASED APPROACH TO EXTENDED DEPTH OF FIELD FOR BRIGHTFIELD MICROSCOPY

François Aguet, Dimitri Van De Ville, and Michael Unser

*Biomedical Imaging Group, Ecole Polytechnique Fédérale de Lausanne, Switzerland*

**INTRODUCTION:** Brightfield microscopy suffers from a limited depth of field, often preventing thick specimens from being imaged entirely in-focus. By optically sectioning the specimen, the in-focus regions can be acquired over multiple images. Extended depth of field methods aim at combining the information from these images into a single in-focus image of the texture on the specimen's surface. For the analysis of the in-focus image, knowledge of the specimen's topography can be mandatory. The topography provided by current methods is usually limited to a map of selected pixel positions and is inherently discretized, which limits its use for quantitative evaluation<sup>1-2</sup>. In this work, we present a joint estimation of the in-focus image (texture) and the topography, based on a model of the specimen and the microscope's point spread function (PSF). The method, which can be likened to an object model-based deconvolution, produces a continuous topography.

**METHODS:** Under the hypothesis that the in-focus information is recorded at the specimen's surface (valid when imaging thick specimens), we propose to model the sample  $o(x,y,z)$  as a thin 3-D surface, described by its topography  $p(x,y)$ , onto which the texture  $f(x,y)$  is mapped:

$$o(x, y, z) = f(x, y)\delta(z - p(x, y)). \quad (1)$$

The image formation and acquisition process is modeled as the convolution between the object and the microscope's 3-D point spread function  $h(x,y,z)$ , followed by sampling to obtain a stack of images. In order to reduce the computational load,  $h(x,y,z)$  is modeled as a Gaussian approximation of a state-of-the-art microscope PSF model<sup>3</sup>.

Estimates for the texture and topography are obtained by minimizing the discrete cost function

$$J = \sum_{i,j,k \in S_3} (s(i,j,k) - (o * h)(i,j,k))^2, \quad (2)$$

where  $S_3$  is the 3-D support of the measured image stack  $s$ . This is achieved via a two-step optimization approach that alternates between updating the texture and the topography<sup>4</sup>. During each step, the texture (respectively topography) estimation is achieved by performing a steepest descent on  $J$ . The estimation is initialized with the topography at a constant level, followed by the

first texture estimation step. In general, the method converges rapidly and requires only a few iterations.

**RESULTS:** We tested our algorithm on phantom data and z-stack acquisitions of biological specimens. In the example shown in Fig. 1, the estimated topography clearly reveals the heap-like structure of the sample.

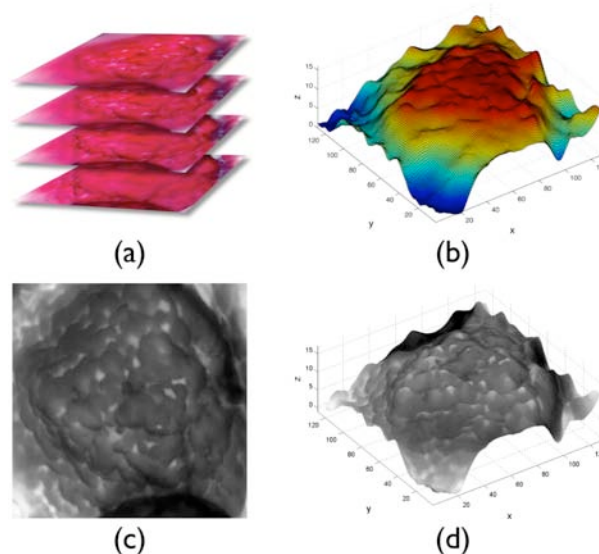


Fig. 1: (a) Partial z-stack acquisition of a murine intestine sample. (b) Estimated topography. (c) Estimated texture. (d) Texture mapped topography.

**DISCUSSION & CONCLUSIONS:** The joint estimation of the texture and topography of a specimen leads to accurate estimates, devoid of the discretization artifacts that commonly result from existing methods. Promising results from simulated and experimental image stacks demonstrate the efficiency of the method.

**REFERENCES:** <sup>1</sup>A. G. Valdecasas, D. Marshall, J. M. Becerra, and J. J. Terrero (2001) *Micron* **32**:559–569. <sup>2</sup>B. Forster, D. Van De Ville, J. Berent, D. Sage, and M. Unser (2004) *Microsc Res Tech* **65**:33–42. <sup>3</sup>S. F. Gibson and F. Lanni (1991) *J Opt Soc Am A* **8**:1601–1613. <sup>4</sup>F. Aguet, D. Van De Ville, and M. Unser (2006) *Proc 3<sup>rd</sup> IEEE Int Symp Biomed Imaging* pp. 778–781.

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