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Introduction to the Issue on Advanced Signal Processing in Microscopy and Cell Imaging

Microscopy imaging, including fluorescence microscopy and electron microscopy, has a prominent role in life science and medical research. During the past two decades, biological imaging has undergone a revolution by way of the development of new microscopy techniques that allow the visualization of tissues, cells, proteins and macromolecular structures at all levels of resolution, physiological states, chemical composition and dynamic analysis. Thanks to recent advances in optics, digital sensing technologies and labeling probes (*i.e.*, XFP - Colored Fluorescence Protein), we can now visualize sub-cellular components and organelles at the scale of a few nanometers to several hundreds of nanometers. As a result, fluorescent microscopy has become the workhorse of modern biology. Further technological advances include structured and coherent light sources, faster and more sensitive detectors, smaller and more specific molecular probes and automation processes for image acquisition. Additionally, there is a push towards multimodal imaging in order to gather complementary information such as the concentration of fluorophores at various wavelengths and the refractive index of a given sample (phase imaging).

With all the advances in technology for microscopy, many of the existing roadblocks in biological and medical research at the microscopic level are problems of *signal processing*. The need to analyze multidimensional images acquired with high spatial and temporal resolution microscopy poses new challenges for researchers in image processing and image analysis. A salient aspect of microscopy data is the highly variable nature of biological objects (cells, organelles, single molecules, etc.) as well as the range of scales. A second aspect is the dynamic nature of the data and the complexity of biological processes involving many entities (chromosomes, vesicles, membrane fusion and budding). Therefore, dedicated efforts are necessary to develop and integrate cutting-edge approaches in image processing and optical technologies to push the limits of the instrumentation and to analyze and quantify the large amount of data being produced. Once the numerical processing is part of the imaging loop, such processing may actually drive the imaging. A dramatic example of this shift in paradigm is super-resolution localization microscopy (PALM and STED), which was rewarded with the 2014 Nobel Prize in Chemistry.

This special issue is devoted to advances in microscopy image processing with a special focus on mathematical imaging and algorithms design. It covers a wide scope of applications and challenges that are specific to the field, from image reconstruction, segmentation, and classification to object tracking.

Methods that are used for reconstructing microscopy images share commonalities with those deployed in medical imaging—the recent push is also towards “compressed sensing” and the use of sparsity to guide the reconstruction process. This issue includes several papers that apply sparsity-driven models and modern optimization techniques for improving image reconstruction. Arildsen, Oxvig, Pedersen, Østergaard, and Larsen present an approach based on sparse approximation and interpolation to reconstruct images in atomic force microscopy. They propose to reduce the critical scanning time and probe-specimen interaction through the use of a sparse (raster and spiral) sampling pattern. Song and Horowitz revisit the classical subtomogram averaging problem in cryo-electron tomography by reformulating it as a low-rank matrix recovery problem subject to constraints on the misfit of to the radon measurements. The proposed approach does not require prior information, such as the number of classes or references of the target structures. It also addresses the missing wedge issue by formulating tomographic sensing operators at each projection angle. The idea of Fortun, Guichard, Chu, and Unser is to combine multiple views of identical specimens obtained at different orientations in order to overcome the intrinsic axial resolution limit of fluorescence microscopy. To that end, they develop a joint deconvolution-and-multiview reconstruction framework that relies proximal splitting. They present preliminary results of super-resolution 3D reconstruction, as proof of concept.

Images in light microscopy are typically blurred and noisy because of the diffraction-limited and photon-limited nature of the data. Specific algorithms are required to restore information and the special issue includes two such contributions. Ponti, Helou, Ferreira, and Mascarenhas introduce a new approach for image deconvolution that relies on constraint sets and sub-gradient projections. The Richardson-Lucy iterations within the POCS framework are interpreted as a gradient iteration allowing constraints to be enforced during restoration of wide-field and confocal images. Tofighi, Yorulmaz, Köse, Yildirim, Çetin-Atalay, and Çetin propose of novel approach for blind deconvolution where the constraints are specified in terms of two closed convex sets convex sets. First, the set of images with a prescribed Fourier Transform phase is used as constraint set; then, a second set called the Epigraph Set of Total Variation (TV) is used to automatically estimate an upper-bound on the TV value of a given image. Both are used as components of iterative microscopic image deblurring algorithms.

A standard paradigm for solving image reconstruction problems is to impose a statistical or regularization prior such as “total variation”, which is now used routinely for deconvolution. It is expected that the use of better models should improve reconstruction quality. Lenz investigates the generalized extreme Pareto distribution (GPD). He relies on this probabilistic modeling to construct a sharpness function, which is then used to drive an autofocus algorithm for microscopy. Gong and Sbalzarini exploit the statistical distribution of the gradient of the image to specify a data-driven regularizer that can be used for solving a variety of inverse problems in light microscopy. The prior, which is learned from a collection of natural-scene images, then takes the familiar form of a variational energy.

The more advanced methodologies for cell classification are based on machine learning techniques and/or statistical modeling. Xu, Megjhani, Trett, Shain, Roysam, and Han present a general non-parametric Bayesian framework for the classification of arbor morphologies. Their scheme automatically identifies the number of classes even if they are not clearly separated, as is the often case with complex biological datasets. To address the problem of cell classification in histopathology, Kang, Yoo, and Na propose a deep probabilistic architecture referred to the tree-structure sum-product network (t-SPN). It is trained using the maximum margin criterion and a l2 regularization to enhance generalization. The method exploits high-pass and Laplacian-of-Gaussian (LoG) filtering and compares favorably with conventional convolution neural networks on benchmark datasets.

Image segmentation remains a challenging problem in 3D microscopy due to the difficulty to separate closely packed cells, to detect nuclei or delineate cell membranes. The standard methodologies include level sets, active contours and watershed methods. In order to segment nuclei in 3D tissue images, Nandy, Chellappa, Kumar, and Lockett propose a novel seed-detection method followed by a graphcut. Their seed detection utilizes a robust model-based 2D slice-by-slice segmentation followed by a suppression of points that are not local maxima. Toutain, Elmoataz, Desquesnes, and Pruvot address the problem of 3D cell segmentation in confocal microscopy within a unifying graphical framework that combines filtering, image segmentation and classification. Their formulation involves partial derivative equations (PDE) on weighted graphs (graph p-Laplacian). Jonic and Sorzano propose a novel algorithm to convert a 3D transmission electron microscopy density volume into a granulated model by controlling the volume approximation error. This granularization method is appropriately developed to study macromolecular dynamics.

The use of time-lapse video-microscopy to capture dynamics of various biological objects has significantly increased in recent years. This special issue includes four papers on this topic. Sadanandan, Baltekin, Magnusson, Boucharin, Ranefall, Jaldén, Elf, Wälby present a fast and robust method for segmenting and tracking *E. coli* cells over time. It includes a quality control and refinement step to correct errors, which is crucial to perform well on phase contrast microscopy images. The algorithm of Chen, Zhao and Yan for tracking cells in 4D data during *C. Elegans* embryogenesis is based on probabilistic relaxation labeling. The tracking is formulated as a non-rigid point matching problem. Their method has a fast parallel implementation, which can provide a significant advantage for large-scale image analysis. Schlangen, Franco, Houssineau, Pitkeathly, Clark, Smal, and Rickman adopt a Bayesian formulation to track moving object with different motion behaviors and to estimate sensor drift. A probability hypothesis density (PHD) filter with classification is combined with a particle filter to provide a joint estimation of the sensor movement with respect to the monitored sample and the intracellular motion of molecular structures in photo-activated localization microscopy (PALM). Pécot, Kervrann, Salamero, and Boulanger address the problem of traffic analysis at the sub-cellular level in 2D- 3D fluorescence microscopy. They obtain an estimate of particle flux based on the global minimization of an energy functional with sparse constraints. The interest of their approach is that it lies in-between individual object tracking and dense motion estimation.

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