

EDITORIAL

## Advanced x-ray tomography: experiment, modeling, and algorithms

To cite this article: K Joost Batenburg *et al* 2018 *Meas. Sci. Technol.* **29** 080101

View the [article online](#) for updates and enhancements.

### Related content

- [A Distributed Compressive Sensing Scheme for Event Capture in Wireless Visual Sensor Networks](#)  
Meng Hou, Sen Xu, Weiling Wu et al.
- [A modified interval subdividing based geometric calibration method for interior tomography](#)  
Zhang Feng, Yan Bin, Li Lei et al.
- [Advanced electron tomography of nanoparticle assemblies](#)  
T. Altantzis, D. Zanaga and S. Bals



**IOP | ebooks™**

Bringing you innovative digital publishing with leading voices to create your essential collection of books in STEM research.

Start exploring the collection - download the first chapter of every title for free.



## Editorial

# Advanced x-ray tomography: experiment, modeling, and algorithms

## Guest Editors

**K Joost Batenburg**

Centrum Wiskunde & Informatica, Amsterdam, Netherlands

E-mail: [joost.batenburg@cwi.nl](mailto:joost.batenburg@cwi.nl)

**Francesco De Carlo**

Advanced Photon Source, Argonne National Laboratory, Lemont, IL, United States of America

**Lucia Mancini**

Elettra - Sincrotrone Trieste S.C.p.A., Basovizza (Trieste), Italy

**Jan Sijbers**

imec-Vision Lab, University of Antwerp, Antwerp, Belgium

Being able to visualize and analyze the 3D interior of an object in a nondestructive way is a fundamental capability for obtaining insights in medicine, materials science, geology, and many other research fields. Also in industry, nondestructive ways of 3D imaging are the key to on-the-fly quality control. X-ray tomography is a powerful technique for carrying out such 3D investigations at resolutions ranging from the millimeter down to the submicron level. It relies on a series of x-ray images acquired from a range of angles around the object. After the acquisition, a 3D image is computed from the measured data by a reconstruction algorithm, possibly combined with various pre- and postprocessing steps. Finally, the resulting data needs to be analyzed and visualized in an application specific manner to obtain answers to a variety of questions about the scanned object. As a data-driven research tool, tomography can provide not just qualitative images but also quantitative results on the internal state of the object.

To push the boundaries of current tomography capabilities, state-of-the-art x-ray imaging devices must be combined with novel, advanced computational methods for processing the resulting data. Traditionally, optimization of the imaging experiments and working on the computational (reconstruction/modelling) part have been separated into different academic communities with little interaction between them, thereby losing opportunities to improve on the state-of-the-art by a combination of advanced experiments and computation. Also, the publication venues and conferences typically focus on either one of the two aspects.

To stimulate the interaction and create a cross-disciplinary research community for advanced x-ray tomography that combines experiments, computational models, and algorithms, the EXTREMA COST Action (MP1207) was initiated in 2013, forming an EU-funded research network with participants from over 20 countries. This special issue of *Measurement Science and Technology* is specifically aimed at collecting research results that combine x-ray imaging experiments with sophisticated modelling and/or computational approaches. It collects results obtained from EXTREMA collaborations as well as research that spans beyond the boundaries of the research network.

The issue collects 18 articles on various aspects of advanced x-ray tomography. In [1], a review is given on the development of low-dose phase contrast tomography in the synchrotron, which can be seen as illustrative for the extensive development effort that goes into new tomographic techniques. The remaining articles contain a broad range of novel research results related to various aspects of the tomographic imaging pipeline.

Understanding the details of the experimental imaging system is crucial to optimizing the results of the complete imaging pipeline. The articles [2–6] deal with *simulation*, *calibration*, and *metrology* in the context of x-ray tomography. Based on a model of the imaging system, the key computational challenge in tomography is to carry out image *reconstruction*. Although the standard reconstruction techniques have been developed decades ago, advances in reconstruction algorithms can yield superior image quality in cases of limited data or data artefacts introduced during acquisition. In [7–9], new algorithmic approaches to tomographic reconstruction are presented.

Both the acquisition and the reconstruction steps may introduce *artifacts* in the reconstructed image: structured image distortions that often have a characteristic visual structure. The articles [10–13] propose different strategies for reducing such artifacts through alternative ways of acquiring and processing the data.

Besides image quality, another crucial performance metric of the tomography pipeline is the time required for a scan and the computational requirements of its data processing. The

articles [14–16] focus on various aspects of *high-throughput* tomography systems, related to running time and data storage requirements.

During the running period of the EXTREMA COST Action (2013–2017), sharing of public tomography data across the community for validation and benchmarking was found to be of primary importance for furthering the advanced tomography community. The papers [17, 18] present the results of two projects aimed at promoting such *open data* research.

## Acknowledgments

The guest editors of this special issue on Advanced X-Ray Tomography acknowledge the networking support of the EXTREMA COST Action MP1207. JS additionally acknowledges the Research Foundation Flanders (FWO) and the Austrian Science Fund (FWF) under the grant numbers G0F9117N and FFG Bridge Early Stage project no. 851249 (ADAM).

## ORCID iDs

Francesco De Carlo  <https://orcid.org/0000-0003-1068-7785>

Lucia Mancini  <https://orcid.org/0000-0003-2416-3464>

Jan Sijbers  <https://orcid.org/0000-0003-4225-2487>

## References

- [1] Mittone A, Bravin A and Coan P 2018 Low-dose quantitative phase contrast medical CT *Meas. Sci. Technol.* **29** 024006
- [2] Lifton J J and Carmignato S 2017 Simulating the influence of scatter and beam hardening in dimensional computed tomography *Meas. Sci. Technol.* **28** 104001
- [3] Zanini F and Carmignato S 2017 Two-spheres method for evaluating the metrological structural resolution in dimensional computed tomography *Meas. Sci. Technol.* **28** 114002
- [4] Wiebicke M, Andó E, Herle I and Viggiani G 2017 On the metrology of interparticle contacts in sand from x-ray tomography images *Meas. Sci. Technol.* **28** 124007
- [5] Jailin C, Buljac A, Bouterf A, Poncelet M, Hild F and Roux S 2018 Self-calibration for lab- $\mu$ CT using space-time regularized projection-based DVC and model reduction *Meas. Sci. Technol.* **29** 024003
- [6] Viganó N R and Solé V A 2018 Physically corrected forward operators for induced emission tomography: a simulation study *Meas. Sci. Technol.* **29** 034005
- [7] Kazantsev D, Guo E, Phillion A B, Withers P J and Lee P D 2017 Model-based iterative reconstruction using higher-order regularization of dynamic synchrotron data *Meas. Sci. Technol.* **28** 094004
- [8] Purisha Z, Rimpeläinen J, Bubba T and Siltanen S 2018 Controlled wavelet domain sparsity for x-ray tomography *Meas. Sci. Technol.* **29** 014002
- [9] Dahl V A, Dahl A B and Hansen P C 2018 Computing segmentations directly from x-ray projection data via parametric deformable curves *Meas. Sci. Technol.* **29** 014003
- [10] Borg L, Jørgensen J S, Friel J and Sporring J 2017 Reduction of variable-truncation artifacts from beam occlusion during in situ x-ray tomography *Meas. Sci. Technol.* **28** 124004
- [11] Brun F, Delogu P, Longo R, Dreossi D and Rigon L 2018 Inpainting approaches to fill in detector gaps in phase contrast computed tomography *Meas. Sci. Technol.* **29** 014001
- [12] Pelt D M and Parkinson D Y 2018 Ring artifact reduction in synchrotron x-ray tomography through helical acquisition *Meas. Sci. Technol.* **29** 034002
- [13] Borges L R, Azzari L, Bakic P R, Maidment A D A, Vieira M A C and Foi A 2018 Restoration of low-dose digital breast tomosynthesis *Meas. Sci. Technol.* **29** 064003
- [14] Janssens E, De Beenhouwer J, Van Dael M, De Schryver T, Van Hoorebeke L, Verboven P, Nicolai B and Sijbers J 2018 Neural network Hilbert transform based filtered backprojection for fast inline x-ray inspection *Meas. Sci. Technol.* **29** 034012
- [15] Mancini L, Kourousias G, Billè F, De Carlo F and Fidler A 2018 About a method for compressing x-ray computed microtomography data *Meas. Sci. Technol.* **29** 044002
- [16] Buurlage J W, Kohr H, Palenstijn W J and Batenburg K J 2018 Real-time quasi-3D tomographic reconstruction *Meas. Sci. Technol.* **29** 064005
- [17] Jørgensen J S, Coban S B, Lionheart W R B, McDonald S A and Withers P J 2017 SparseBeads data: benchmarking sparsity-regularized computed tomography *Meas. Sci. Technol.* **28** 124005
- [18] De Carlo F *et al* 2018 TomoBank: a tomographic data repository for computational x-ray science *Meas. Sci. Technol.* **29** 034004