

### Zineb SAGHI (CEA-Leti)

# 3D analysis of nanoscale materials using electron and FIB-SEM tomography techniques







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## **3D characterization techniques at the nanoscale**





## 3D characterization techniques at the nanoscale



Field ion microscope combined with a mass spectrometer

Transmission electron microscope (TEM) Focussed ion beam combined with a scanning electron microscope (FIB-SEM)

**Dimensions / Morphology:** serial sectioning by FIB-SEM, BF-TEM tomography, HAADF-STEM tomography **Composition / Doping:** APT, STEM-EDX/STEM-EELS tomography, FIB-EDX tomography

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## Key steps in electon tomography (ET)

Data acquisition



Pre-processing, alignment, reconstruction





Alignment and reconstruction algorithms

Tools for segmentation and quantification





(dedicated) TEMs

Dedicated specimen holders



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## **Projection requirement**

### **Projection requirement for ET:**

"The signal used for tomographic reconstruction should be a monotonic function of a projected physical quantity."

### Standard imaging modes in ET:

 Biology: BF-TEM (bright field transmission electron microscopy)

### **3D** electron tomography of brain tissue



M.R Fernandez-Fernandez et al. Journal of Cell Science 2017, 130:83.





## **Projection requirement**

### **Projection requirement for ET:**

"The signal used for tomographic reconstruction should be a monotonic function of a projected physical quantity."

### Standard imaging modes in ET:

- Biology: BF-TEM (bright field transmission electron microscopy)
- Materials science: HAADF-STEM (high angle annular dark field scanning TEM)
  - Incoherent (no diffraction contrast)
  - Z contrast (I ~  $Z^2$ )

### 3D electron tomography of Er-doped porous Silicon



G. Mula et al. Scientific Reports 2017, 7:5957



## **Fourier slice theorem**



The 2D FT of the projections fill the 3D Fourier space the sample

We need to acquire as many projections as possible



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## **Dedicated specimen holders**





## **Dedicated specimen holders**

### **On-axis tomography holder**





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## **Reconstruction approaches**



- Fourier approaches
- Analytical approaches
- Algebraic approaches





## **Fourier-based approaches**



	Cartesian grid						Pseudo- polar grid										Polar grid				
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Polar to cartesian grid interpolation induces many artefacts

• Equally slopped tomography (EST)

Equally slopped projections => pseudo-polar FFT (M. C. Scott et al., Nature, 2012, 483:444)

• Min-max non-uniform FFT (NUFFT)

Convolution of the non-uniformed samples with a kernel + resampling on an oversampled cartesian grid (Fessler, *IEEE T-SP*, 2003, 51(2):560) Widely used in MRI





## **Analytical methods**



A simple backprojection  $b \Rightarrow$  a blurred version of the original object f:  $b(x, y) = f(x, y) \otimes \frac{1}{\sqrt{x^2 + y^2}}$ Solution: apply a ramp filter to compensate for this blurring :  $f(x, y) = FT^{-1}(B(u_1, u_2) \times |u|)$  $u| = \sqrt{u_1^2 + u_2^2}$ <u>https://web.eecs.umich.edu/~fessler/course/516/l/c-tomo.pdf</u>

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### **Algebraic methods**



Various algorithms exist for solving this problem: Algebraic reconstruction technique (ART), Simultaneuous ART (SART), Simultaneous iterative reconstruction technique (SIRT), etc

C.OS. Sorzano et al., Biomed Res Int. 2017, 6482567.



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SIRT

SIRT solves the following least squares problem:

$$argmin_{x} \left\{ \frac{1}{2} \|Px - y\|_{2}^{2} \right\}$$

$$40 \text{ iterations}$$

$$100 \text{ iterations}$$

$$200 \text{ iterations}$$

$$100 \text{$$

G. Haberfehlner, *3D nanoimaging of semiconductor devices and materials by electron tomography*, 2013.





## **Spectroscopic ET**



#### Challenges (compared to STEM-HAADF):

- Long acquisition times
- High electron doses
- Poor-quality reconstructions using standard algorithms



M. Jacob et al., Semicond. Sci. Technol. 2021, 36: 035006.









## Total variation minimization (TV):

TV promotes sparsity in the gradient domain and is very well suited for objects with *piecewise* constant regions and sharp edges



Z. Saghi et al., Nano Letters 2011, 11(11), 4666. R. Leary et al., Ultramicroscopy 2013,131, 70.





### Limitations of TV:

- Staircase artefacts when the object is not truely piecewise constant.
- Complex structures require more projections than reported for simple objects (Y. Jiang et al. Ultramicrosc. 2017, 186)
- The quality of the reconstructions degrade rapidly in the presence of Poisson noise.

### **Higher order TV** (Incorporation of higher order derivatives):

- HOTV (ref: R. Archibald et al. J. Sci. Comput. 2015, 67) and TGV (total generalized variation, see e.g.: M. Benning et al. J. Sci. Comput. 2013, 54:269) TGV NUFFT
- Promote piecewise smooth regions while preserving sharp edges.



T. Sanders et al., Ultramicrosc. 2017, 174: 97.



TV

F. Knoll et al., Magn. Reson. Med 2011, 65(2):480.







### **Sparsity in the wavelet domain:**

- Multiscale approach used in microscopy mainly for denoising purposes.
- Widely used in MRI and recently applied to CT and ET.
- Applied to a wider range of objects, compared to gradient sparsity.
- Knowledge about the sample => choice of the appropriate wavelet function.



Three-dimensional imaging of localized surface plasmon resonances of metal nanoparticles. Nicoletti *et al.* Nature 2013, 502:80







Reference



splineWaveletTransformATrousAlgorithm ssim: 0.89

MeyerWaveletsCompactinFourierSpace ssim: 0.9





FastCurveletTransform

ssim: 0.92

UndecimatedBiOrthogonalTransform ssim: 0.92





H. Cherkaoui et al. EUSIPCO 2018, 36 20 NANOELEC. Commercial software: Inspect3D (FEI), etc. Commercial software with advanced algorithms: NONE.

Name	General features	Reconstruction algorithms
Astra toolbox	Initially developed for X-ray tomography GPU-based implementations. Incorporated in Inspect3D (FEI)	Standard reconstruction algorithms + TV plugin
Тотору	Initially developed for X-ray tomography CPU-based implementations. Recently integrated with Astra toolbox.	Standard reconstruction algorithms
Tomotools	Developed for electron tomography Tools for alignment + reconstruction.	Standard reconstruction algorithms (Astra + Tomopy)
Tomviz	Open source plateform for alignment, reconstruction and visualization.	Standard reconstruction algorithms + TV
Matlab package (T. Sanders et al., Ultramicroscopy 2017, 174: 97)	Alignment, inpainting and denoising.	TV, HOTV, and multiscale HOTV.
<b>Graptor</b> (R. Huber et al., Nanoscale 2019, 11:5617)	Developed for EELS/EDS tomography GPU-based implementations.	TGV algorithm and multi-modal reconstruction approach
<b>PySAP-etomo</b> (M. Jacob et al., Ultramicroscopy 2021, 255: 113289)	Adaptation of PySAP + Modop libraries Includes Astra and Pywavelet.	Implementation of gradient-based (TV, TGV, HOTV) and wavelet-based methods.



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### Application 1: HAADF-STEM tomography of Er-doped porous Silicon

HAADF-STEM mode Tilt angles: -90°:1°:+90°

> **TV/Haar:** staircasing artifacts due to the piecewise constancy assumption.

HOTV\_3 : no staircasing artifacts. Noise-like oscillations appear with large tilt increments.

**Bior4.4:** best results. Induces smoothness with large tilt increments



increment (in degrees).



## Application 2: STEM-EELS tomography of a Ge-rich GeSbTe (GST) thin film for phase-change memory (PCM) applications



From: Phase change materials: science and applications, S. Raoux and M. Wuttig, Springer Verlag, New York, 2009.

Current challenge: PCM memory devices stability is challenged at high temperature (automotive applications).



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## Application 2: STEM-EELS tomography of a Ge-rich GeSbTe (GST) thin film for phase-change memory (PCM) applications





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## **3D characterization techniques at the nanoscale**





## Atom probe tomography (APT)



APT on heterogenous structures with different fields of evaporation:

- Variations in the local magnification
- Distortions in the reconstructed volume
- Abnormal atom density variations



APT of GaN/InGaN/AIGaN multilayers obtained with :(a) the standard method(b) the two-step, interfaceflatness driven, process.

F. Vurpillot et al., Ultramicroscopy 132 (2013), 19.





## **Correlative ET/APT**

Improvement of APT reconstruction using the combination of electron tomography reconstruction with density corrected APT algorithm.



3D APT reconstruction of the 45 nm device (Density corrected algorithm) 3D electron tomography reconstruction of the 45 nm device after segmentation.

A.Grenier et al., APL, 106, 213102 (2015)

3D boron distribution inside the gate with minimal distortions.





## **Correlative spectroscopic ET/APT**







#### Phase Separation in an Alnico 8 Alloy. Guo et al., Microsc. Microanal. 22, 1251 (2016)

50 nm

Fe

Ni

### **EELS/EDX-STEM** tomography:

Non-destructive, large fields of view, high spatial fidelity, limited spatial resolution (few nms), LLD  $\sim 1 \times 10^{20}$  at.cm<sup>-3</sup>, quantification methods not fully developped.

### APT:

Destructive, small fields of view, limited spatial fidelity, high spatial resolution, LLD  $\sim 5 \times 10^{18}$  at.cm<sup>-3</sup>, quantitative method.



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## **FIB-SEM** tomography













 AMST (Alignment to Median Smoothed Template) - J. Hennies et al., Sci. Rep. 10, 2004 (2020)



IEEE Trans. Image Process. 21(10),

4420 (2012).





### Segmentation

- Basic thresholding
- Watershed algorithm
- Machine learning approaches
- (e.g.: random forest classification)
- Deep learning approaches (e.g.: U-net, mask R-CNN, Stardist)







### Phytoplankton cells



cell (SOFC) electrodes LSCF CGO YSZ 22 µm

Solid oxide fuel

Magnetic metalpolymer nanocomposites





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## **3D** characterization techniques at the nanoscale

Multi-scale correlative analysis:

- CT/FIB-3D
  - ET/APT
  - CT/ET

...







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## Thank you for your attention!



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