Imaging of Defect Rich Heterogeneous Interfaces using Compressive Sensing STEM

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Compressive sensing (CS) scanning transmission electron microscopy (STEM) has been used to both increase the acquisition rate [1] and lower the electron dose [2] required to form high resolution images of both crystalline and non-crystalline structures [3] by the use of probe subsampling and an appropriate dictionary learning algorithm, such Beta Process Factor Analysis (BPFA) [4]. Presented here is an application of compressive sensing to the imaging of a cadmium telluride (CdTe) thin film solar cell material on a silicon (Si) substrate, with a focus on the many defects present in the sample and the interface between the CdTe and Si. The interface of these materials and the number and location of defects [5] affects the efficiency of the solar material and therefore is of high importance to manufacturers. As such, during the application of CS to these materials it is vital that the interface and defects are retained during the inpainting process.

The CdTe lamella was prepared by FIB preparation using a FEI Helios Nanolab 600i dualbeam FIB/SEM and the STEM imaging was carried out using an aberration corrected JEOL 2100F. The images were subsampled artificially and inpainted using the Nuxutura Image Inpainting software, which utilises a Beta Process Factor Analysis with Expectation Maximisation (BPFA-EM) algorithm implementation.

Figure 1 shows an annular bright field STEM image of the interface between CdTe and Si with many defects present, an artificially subsampled image of the CdTe-Si interface, and the BPFA-EM reconstruction of that subsampled image. The reconstruction shows that both the interface and defects are retained through the subsampling and reconstruction process with sufficient quality to be analysed. Figure 2 shows the same schematic as Figure 1 except using a linehop sampling pattern rather than a random sampling pattern. Again, the atomic information along with the interface and defects are retained through the subsampling and inpainting process.

References:

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