3DSEM: A 3D microscopy dataset

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Abstract

The Scanning Electron Microscope (SEM) as a 2D imaging instrument has been widely used in many scientific disciplines including biological, mechanical, and materials sciences to determine the surface attributes of microscopic objects. However the SEM micrographs still remain 2D images. To effectively measure and visualize the surface properties, we need to truly restore the 3D shape model from 2D SEM images. Having 3D surfaces would provide anatomic shape of micro-samples which allows for quantitative measurements and informative visualization of the specimens being investigated. The 3DSEM is a dataset for 3D microscopy vision which is freely available at [1] for any academic, educational, and research purposes. The dataset includes both 2D images and 3D reconstructed surfaces of several real microscopic samples.

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Specifications table

| Subject area                  | 3D microscopy vision, biology, materials science, mechanical engineering. |
|-------------------------------|--------------------------------------------------------------------------|
| More specific subject area    | 3D surface structure, 3D structural analysis.                             |
| Type of data                  | 2D SEM images (.JPEG,.TIFF), 3D surface models (.OFF,.PLY).              |

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2D SEM images are captured by a Hitachi S-4800 field emission Scanning Electron Microscope (SEM). The 3D Shape models are created using the 3D reconstruction algorithm illustrated in [2].

Digital images, 3D shape models.

Experimental setup along with its parameters described in [2].

Several qualitative and quantitative experiments showed in [2]. The results were promising.

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The dataset is freely available at [1] for any academic, educational, and research purposes. More 2D SEM images and 3D surface models will be added into the dataset continuously.

Value of the data

- Discovering 3D surface structure from SEM images would provide anatomic surfaces and allows informative visualization of the objects being investigated.
- To provide the current dataset, an optimized multi-view 3D SEM surface reconstruction algorithm is designed [2].
- Several experimental validations are performed on real microscopic samples as well as synthetic data. The quantitative and qualitative results are promising [2].
- Many research and educational questions truly require knowledge and information about 3D microscopic structures. The present dataset along with the algorithm would be helpful in this way.
- The current dataset which includes 2D SEM images and 3D surface models, and the underlying methodology may serve as a guide for 3D SEM surface reconstruction.
- The present work is expected to highlight the important roles and applications of 3D microscopy vision, particularly 3D surface reconstruction from SEM images, and open the doors for several interesting directions to advance the level of the research area.

1. Data

Dataset names and attributes are briefly presented in Table 1. Fig. 1 shows two samples of the entire dataset including 2D SEM micrographs and 3D reconstructed surfaces.

A Hitachi S-4800 Field Emission Scanning Electron Microscope is used to capture the micrographs. This microscope is equipped with a computer controlled 5 axis motorized stage capable of 360° of rotation with a tilt range of −5°−70°. Sample manipulation, such as Z-position, tilt, and rotation of the stage, as well as image processing and capture functions are operated through the Hitachi PC-SEM software. The working distance that would give the required depth of focus is specified at the maximum tilt for every specimen at the magnification chosen for image capture. As the sample is tilted in successive 1° increments through the software, the image is centered manually by moving the stage in the x- and y-axes with the stage positioning trackball. The working distance and magnification are kept consistent in every captured image of the tilt series by changing the Z-axis position as required. Brightness and contrast are manually adjusted for consistency between micrographs, using the same structure in every image. The micrographs are acquired with an accelerating voltage of 3 kV, employing the signals from both the upper and lower secondary electron (SE) detectors. Readers interested in SEM imaging are referred to [4,5] for further information.

The 3D surface models and their construction strategy are fully detailed in the paper [2]. At present, the 3DSEM dataset includes three different samples illustrated in Table 1. This dataset is an ongoing project in which further samples will be added to the dataset by near future. As we
mentioned earlier, the 3DSEM dataset is freely available at [1] for any educational, research, and academic purposes.

### 2. Experimental design, materials and methods

3D surface reconstruction from a set of 2D images employs several computational technologies, including multi-view geometry, computer vision, machine learning, and optimization strategies to tackle the inverse problem going from 2D images to 3D surface models [6,7]. The complete pipeline of our proposed optimized multi-view framework for 3D SEM surface reconstruction has six stages. At the first stage, a set of 2D SEM micrographs are taken by tilting the specimen across variant angles. The step requires SEM imaging styles, such as changing magnifications, tilting the specimen, employing SE or/and BSE detectors. We then detect the feature points in each 2D image in the set and estimate image motion based on a set of corresponding points. The step requires SEM imaging styles, such as changing magnifications, tilting the specimen, employing SE or/and BSE detectors. We then detect the feature points in each 2D image in the set and estimate image motion based on a set of corresponding points. Once we are done with estimating the relative position of the images, the 3D position of all corresponding points will be reconstructed by linear triangulation [6–8]. The final step is doing a refinement process by defining a cost function for any set of parameters (e.g., SEM extrinsic parameters and 3D positions) as to whether this is a good or bad set and find the best fitness model in the set.

Further information about the proposed 3D SEM surface reconstruction framework along with several experimental validations are explained in [2].

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### Table 1

| Dataset Name               | Image Attributes | SEM Detector | File Size          |
|----------------------------|------------------|--------------|--------------------|
| Tapetal cell of *Arabidopsis thaliana* | 2560*1920 grayscale, 512dpi SE (mix) | (.zip, 11.3 MB) |
| **Description:** | The dataset contains five 2D images from a biological sample called “Tapetal cell of *Arabidopsis thaliana*” and its 3D surface model (.OFF format). The set of 2D images were obtained by tilting the SEM specimen stage 9 degrees from one to the next in the image sequence. |

| Dataset Name               | Image Attributes | SEM Detector | File Size          |
|----------------------------|------------------|--------------|--------------------|
| Pollen grain from *Brassica rapa* | 854*640 grayscale, 512dpi SE (mix) | (.zip, 620 KB) |
| **Description:** | This dataset includes four 2D images from a biological sample called “pollen grain from *Brassica rapa*” and its 3D point cloud (.PLY format) which could be easily converted to a surface model by MeshLab [3]. The set of 2D images were obtained by tilting the specimen stage 3 degrees from one to the next in the image sequence. |

| Dataset Name               | Image Attributes | SEM Detector | File Size          |
|----------------------------|------------------|--------------|--------------------|
| TEM copper grid (I)        | 2560*1920 grayscale, 512dpi SE (mix) | (.zip, 18.9 MB) |
| **Description:** | The dataset contains five 2D images from a material object called “TEM copper grid” and its 3D surface model (.OFF format). The set of 2D images were obtained by tilting the SEM specimen stage 7 degrees from one to the next in the image sequence. |

| Dataset Name               | Image Attributes | SEM Detector | File Size          |
|----------------------------|------------------|--------------|--------------------|
| Diatom frustule            | 2560*1920 grayscale, 512dpi SE (mix) | (.zip, 10.4 MB) |
| **Description:** | This dataset contains three 2D images from a “diatom frustule” and its 3D point cloud (.off format) which could be easily converted to a surface model by MeshLab [3]. The set of 2D images were obtained by tilting the specimen stage 15 degrees from one to the next in the image sequence. |
Based on the history of downloads (Fig. 2), the 3DSEM appears to be a popular dataset in the research community. In the near future, we would very much like to increase the number of users for the project and create more datasets using different algorithms.
Appendix A. Supplementary material

Supplementary data associated with this article can be found in the online version at http://dx.doi.org/10.1016/j.dib.2015.11.018.

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