

# Data-fusion of high resolution X-ray CT, SEM and EDS for 3D and pseudo-3D chemical and structural characterization of sandstone.

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## Abstract

When dealing with the characterization of the structure and composition of natural stones, problems of representativeness and choice of analysis technique almost always occur. Since feature-sizes are typically spread over the nanometer to centimeter range, there is never one single technique that allows a rapid and complete characterization. Over the last few decades, high resolution X-ray CT ( $\mu$ -CT) has become an invaluable tool for the 3D characterization of many materials, including natural stones. This technique has many important advantages, but there are also some limitations, including a tradeoff between resolution and sample size and a lack of chemical information. For geologists, this chemical information is of importance for the determination of minerals inside samples. We suggest a workflow for the complete chemical and structural characterization of a representative volume of a heterogeneous geological material. This

1 workflow consists of combining information derived from CT scans at different spatial  
2 resolutions with information from scanning electron microscopy and energy-dispersive  
3 X-ray spectroscopy.

4

5 **Keywords:** X-ray tomography;  $\mu$ -CT; scanning electron microscopy; energy-dispersive  
6 spectroscopy; data fusion

## 7 **1. Introduction**

8 Investigation of natural stones, and the behavior of fluids in their pore networks, is  
9 important in many scientific fields. This includes the study of reservoir rocks, which is  
10 important for the oil and gas industry (Antrett, 2013; Jiao et al., 2014), but also for  
11 hydrogeology and water management studies (Cirpka et al., 2014; Lagrou et al., 2004).  
12 Pore-scale processes, such as ice (De Kock et al., 2013; Matsuoka and Murton, 2008) or  
13 salt precipitation (Derluyn et al., 2013) and carbonatation, also have a big influence on  
14 the weathering of building stones (Dewanckele et al., 2014), the behavior of pavement  
15 materials, or the cracking of concrete. A full characterization of the complex  
16 microstructure of porous media is thus of vital importance to understand and predict  
17 the processes regulating the phenomena mentioned above. For many of the materials  
18 that are common in these research fields, one main difficulty to overcome is the multi-  
19 scale character of their structure. Geological materials are composed of a mixture of  
20 macro- to micro- and even nano-scale features. Furthermore, their composition is very  
21 heterogeneous and can vary drastically over a distance of just centimeters. These  
22 inhomogeneous characteristics make it very difficult to find and combine techniques  
23 that can analyze samples over different length scales, since it is often of interest to  
24 analyze centimeter-scale samples at nanometer-precision.

1 Another layer of complexity in the understanding of pore-scale processes in natural  
2 stones is that these processes are not only influenced by the pore structure, but also by  
3 the chemical composition of the material. In this work, a combination of different  
4 techniques is proposed, using a well-defined workflow, to acquire as much information  
5 as possible on the structure and composition of a sandstone sample. This work focuses  
6 on high resolution X-ray CT ( $\mu$ -CT) and scanning electron microscopy (SEM) combined  
7 with energy dispersive X-ray spectroscopy (EDS) (Reed, 2005) and focused ion beam  
8 nanotomography (FIB-nt) (Holzer et al., 2004; Keller et al., 2011).

9 Over the last few decades,  $\mu$ -CT has become an invaluable tool for the 3D  
10 characterization of many materials, including natural stones (Cnudde and Boone, 2013;  
11 Ketcham and Carlson, 2001; Wildenschild and Sheppard, 2013). Although there are  
12 many important advantages for this technique, there are also some limitations. A first  
13 shortfall of  $\mu$ -CT is that high-resolution analysis always requires small samples, even in  
14  $\mu$ -CT systems using magnification optics; although nanofocus X-ray tubes can have a  
15 focal spot size down to about 400 nm, this spatial resolution can only be achieved by  
16 imaging extremely small (less than 1 mm) samples (Al-Raoush and Papadopoulos,  
17 2010). This leads to questions about representative elementary volume (REV),  
18 particularly for heterogeneous geological samples. Second, the lack of chemical  
19 information in  $\mu$ -CT datasets is a big shortcoming for geologists, as this chemical  
20 information is of importance for the determination of minerals present in samples.  
21 Although a rough estimate or relative relationship between the densities and atomic  
22 number of the elements in the sample can be determined, quantitative values remain a  
23 challenge (Jussiani and Appoloni, 2014). This mineralogical information is of importance  
24 for the understanding of various processes that happen in the pore space of geological  
25 materials, including migration of (saline or acid) fluids (Derluyn et al., 2013) and

1 dissolution and precipitation reactions (Dewanckele et al., 2012). In an effort to try to  
2 enhance the information acquired by  $\mu$ -CT, additional experiments using different  
3 methods can be executed. More exactly, we propose a workflow combining  $\mu$ -CT with 2D  
4 imaging techniques (SEM and EDS), and using image registration for a one-to-one fusion  
5 of the data. Data fusion of CT and 2D microscopy techniques was already done by  
6 Huddleston-Holmes and Ketcham (2005), proving the complementarity of both  
7 methods. Over the past decade, image quality and registration techniques have  
8 improved, and 3D registration algorithms have become accessible to researchers.  
9 SEM can provide 2D images of large surfaces, at much higher resolutions than  $\mu$ -CT (less  
10 than 10 nm for secondary electron images; Reed, 2005). Since images can be tiled into a  
11 mosaic image, there is no real limit to the surface that can be analyzed. Combining  $\mu$ -CT  
12 with SEM images can improve information on microporosity and microstructure of  
13 analyzed samples (Sok et al., 2010). When combined with EDS, the chemical  
14 composition can be analyzed as well, to allow the visualization of spatial distribution of  
15 different mineral phases.

## 16 **2. Materials and methods**

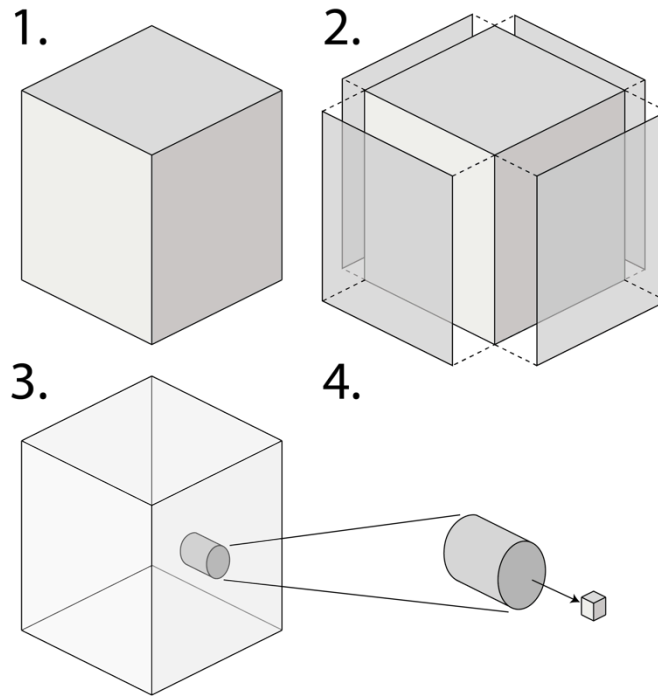
### 17 **2.1. The Vosges sandstone**

18 For this research, a material had to be used that was heterogeneous in terms of  
19 structure and mineralogy. Simple sandstones (e.g. Bentheimer sandstone; Nijland et al.,  
20 2004) are not suited because they require far less analysis to understand their structure.  
21 For this methodological study, a stone was required with a structure possessing a good  
22 variation between macropores ( $> 20 \mu\text{m}$ ) and micropores ( $< 1 \mu\text{m}$ ), and a significant  
23 amount of clays and feldspars to ensure enough chemical variation.

1 Therefore, a variety of Vosges sandstone, the *grès à Meule*, was selected. This fine-  
2 grained pink sandstone from the east of France consists of about 65 % quartz, around 25  
3 % feldspars and about 10 % clay minerals such as micas, smectites and kaolinite  
4 (Schmitt et al., 1994); its average porosity is around 22 % (Bésuelle et al., 2000). The  
5 stone is found in the lower part of the *grès à Voltzia* formation, a local term for a part of  
6 the Buntsandstein formation, deposited in the lower Triassic (Gall and Grauvogel-  
7 Stamm, 2005; Shear et al., 2009) and has been mainly used as a local building stone and  
8 milling stone.

## 9 **2.2. Methods**

10 A workflow was developed to combine information from  $\mu$ -CT, SEM, EDS and FIB-nt. The  
11 base sample of the entire study was a 14 mm, cuboid Vosges sandstone sample (Figure  
12 1). This sample was embedded in resin to ensure mechanical strength, and polished  
13 down to 1/4  $\mu\text{m}$  accuracy. The first step was a 'medium resolution' CT scan of the entire  
14 sample, at the best spatial resolution (9  $\mu\text{m}$ ) possible for that sample size (Figure 1, 1),  
15 using the HECTOR setup (Masschaele et al., 2013a) of the Ghent University Centre for X-  
16 ray Tomography (UGCT, [www.ugct.ugent.be](http://www.ugct.ugent.be)). Next, SEM images and EDS mappings for  
17 three of the main occurring elements (silicon, potassium and iron; determined by point  
18 measurements of the ray spectrum) were performed on all 5 exposed surfaces of the  
19 sample (Figure 1, 2). Thereafter, a small, 1 mm diameter cylindrical subsample was  
20 drilled out of the main sample, and scanned at high resolution (0.86  $\mu\text{m}$ ) using UGCT's  
21 MEDUSA setup (Figure 1, 3). We leave the option open to perform other, higher-  
22 resolution techniques on this small micro-plug (at resolutions <100 nm), to study  
23 regions of the sample in 3D which were too fine-grained for  $\mu$ -CT (Figure 1, 4).



1  
2 **Figure 1: Schematic representation of the workflow for this study. 1) 14 mm sample for medium resolution**  
3 **CT. 2) SEM and EDS mappings of the 5 exposed sides of the sample. 3) Extraction of 1 mm diameter sub-**  
4 **sample for high-resolution CT. 4) Possibility of further investigation of sub-sample using other, high-**  
5 **resolution techniques.**

6 **2.2.1. X-ray computed tomography**

7 *Image acquisition*

8  $\mu$ -CT scans were performed at the UGCT. Two different setups were used, one for the  
9 medium resolution, and another one for the high resolution scans. For the medium  
10 resolution scans, HECTOR (High Energy CT Optimized for Research; (Masschaele et al.,  
11 2013b)) was used. This system uses a microfocus directional target X-ray source up to  
12 240 kV and 280 W (XWT 240-SE, from X-RAY WorX) and a large flat-panel detector (40 x  
13 40 cm<sup>2</sup> PerkinElmer 1620 CN3 CS). For this experiment, the source was operated at a  
14 voltage of 150 kV and a power of 10 W. 2000 projections were taken at an exposure time  
15 of 2 s. The voxel size of this dataset was 8.9  $\mu$ m.

16 High-resolution CT scans were performed on the new MEDUSA setup of the UGCT, which  
17 is equipped with a Varian PaxScan 2520 a-Si flat-panel detector. A transmission type X-  
18 ray tube (X-RAY WorX THC) was used for this experiment. 2800 projections were taken,

1 with an exposure time of 1.5 s. The tube voltage for this experiment was 90 kV. Using  
2 these parameters, a spatial resolution of 0.86  $\mu\text{m}$  was reached. Both systems were  
3 developed in-house at the UGCT, in collaboration with the spin-off company X-Ray  
4 Engineering bvba ([www.XRE.be](http://www.XRE.be)).

#### 5 *Reconstruction and Image Analysis*

6 For the reconstruction of the raw projections, the in-house developed Octopus  
7 reconstruction software was used (Vlassenbroeck et al., 2007)  
8 ([www.octopusreconstruction.eu](http://www.octopusreconstruction.eu)). Image analysis was done using UGCT's Morpho+  
9 software (Brabant et al., 2011), which has recently been commercialized by  
10 InsideMatters ([www.octopusanalysis.eu](http://www.octopusanalysis.eu)), and with Avizo® ([www.vsg3d.com](http://www.vsg3d.com)).

#### 11 **2.2.2. Scanning electron microscopy and Energy Dispersive X-ray spectroscopy**

12 For each of the exposed surfaces of the sample (top surface and 4 lateral surfaces), a grid  
13 of SEM images was acquired to compose a photomosaic of the entire surface. The  
14 analysis was performed on a JEOL 5310-LV system, equipped with a secondary and  
15 backscattered electron detector and an Oxford Instruments silicon-drift detector for EDS  
16 analysis. A total of 300 backscattered SEM images were acquired at each plane, resulting  
17 in a total of 1500 images, which had to be taken manually since the JEOL 5310 is not  
18 equipped with automated stage control. The SEM was operated at 20 kV with a spot size  
19 of 13 nm and a working distance of 20 mm. Images were stitched together using the  
20 Microsoft Image Composite Editor (ICE; <http://research.microsoft.com>), which has  
21 algorithms for automatic grid-based registration of images. Next, EDS mappings were  
22 obtained, at 24 to 28 images per side. Due to time restrictions, only half of the surface of  
23 the SEM photomosaics was covered, at half the magnification. This resulted in 125  
24 separate acquisitions, taking over 40 hours to acquire the data.

### 2.2.3. Data fusion

Image registration was used to link the information from the different techniques. For various reasons it was very difficult to do this registration using automated algorithms: images were acquired using different techniques and different detectors, meaning that the form of corresponding features in images captured with different methods was not completely the same. This effect occurs since the angles of the 14 mm sample were not completely straight, and SEM imaging was therefore not always performed with a beam perpendicular to the sample surface. Furthermore, the direction of the CT slices in the vertical plains was never completely parallel or perpendicular to the samples surfaces. Furthermore, SEM images provide a depth of view beyond the sample surface, which is absent in 2D CT slices. This is usually an advantage of SEM imaging, but proves to be a disadvantage for registration. To overcome these issues, landmark-based registration was employed. Corresponding points in the reference and transformed image were manually selected. The transformation to match those two point groups was then calculated and images were transformed or warped using the same transformation equation. If there was only rigid-body movement, and no deformation, a simple rigid transformation algorithm could be used to register the two images. However, due to the use of different detectors and techniques, distortion was present in the images, and an algorithm called Bookstein Image Warp (Bookstein, 1989) was used to register the images. Using this algorithm, points from the transformed group were forced to fit the reference points, and all image points in between were shifted by using correlation of the surrounding points (Bookstein, 1989). In this research, three types of registration were necessary: 2D to 2D registration to match EDS data to the SEM images, 2D to 3D registration to match these data to the 3D CT volume, and 3D to 3D registration to identify the location of the sub-sample in the base sample. In the first case, only X/Y



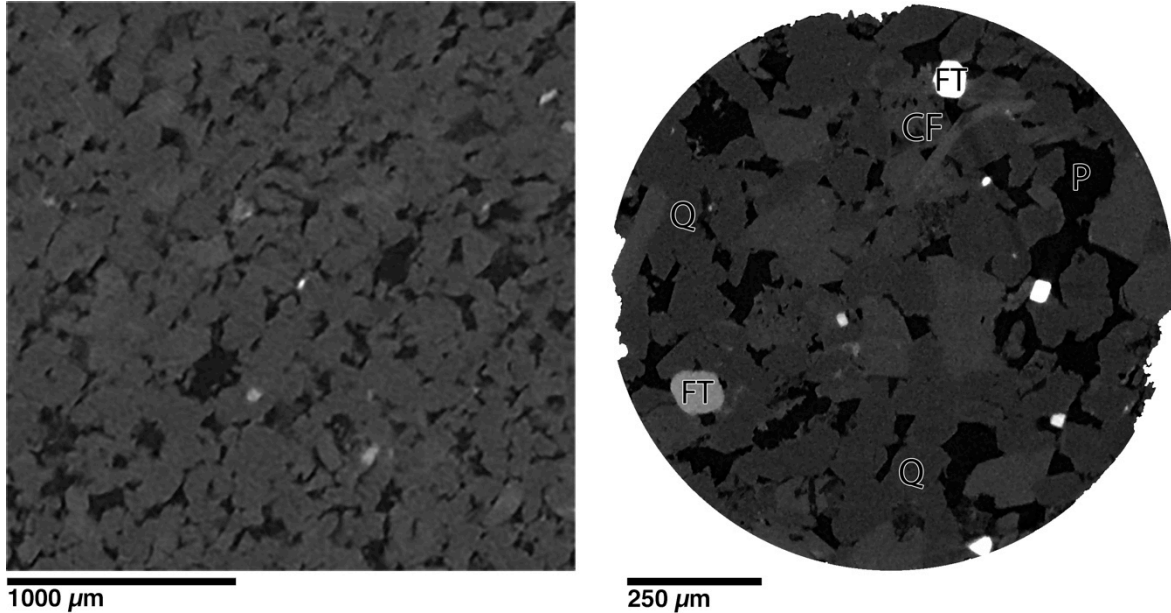
1 transformation was necessary, whereas in the other cases, rotation and translation in  
2 three dimensions had to be performed.

### 3 **3. Results**

#### 4 **3.1. X-ray computed tomography**

5 The first CT scan at medium resolution (voxel size: 8.9  $\mu\text{m}$ ) provides a good insight  
6 regarding the distribution of dense inclusions (mostly titanium oxides), and gives an  
7 idea about the internal structure of the stone. Most of the rest of the solid matrix had  
8 about the same grey value, so there is hardly any contrast between the quartz grains and  
9 other mineral phases present (Figure 2, left). Further, the size of the grains and of the  
10 pore system makes it impossible to perform a decent analysis of grain/pore shape and  
11 size. Image analysis with Morpho+ (Brabant et al., 2011) resulted in a porosity of 9.04  
12 %, consisting of 5.4 % open porosity (connected to the outside of the sample) and 3.6 %  
13 closed porosity. This high amount of closed porosity is not normal for a sandstone like  
14 the Vosges sandstone (Bésuelle et al., 2000), and can only be explained by the lack of  
15 spatial resolution in the dataset. Attention has to be drawn to the high dependency of  
16 results on choices made during image segmentation. A small difference in threshold  
17 value causes high changes in resulting porosity, meaning that the resolution of this CT  
18 scan is just not high enough for decent analysis of the sample. High resolution scanning  
19 of the 1 mm microplug (voxel size: 0.86  $\mu\text{m}$ ) resulted in a much better scan where  
20 porosity was clearly resolved; boundaries between porosity and grains were very sharp  
21 (Figure 2, right), meaning that results from image segmentation can be considered  
22 reliable. Grains and pores could be segmented, and the total porosity by image analysis  
23 was found to be 16.50 %, of which only 0.2 % was characterized as closed porosity. In

1 the high-resolution images, there was contrast between quartz grains, feldspars and clay  
2 minerals, and dense minerals. However, contrast between quartz and feldspars and  
3 clays was not enough to analyze each individual phase.



4 1000  $\mu\text{m}$  250  $\mu\text{m}$   
5 **Figure 2: Left: detail of 2D slice from medium resolution CT scan (voxel size: 9  $\mu\text{m}$ ); right: 2D slice from high**  
6 **resolution CT scan (voxel size: 0.86  $\mu\text{m}$ ). Image quality is much better in the high-resolution scan, so sample**  
7 **analysis is much more accurate than for the low resolution scan. Annotations: Q = quartz, CF = clays and**  
8 **feldspars, FT = iron and titanium oxides, P = pore.**

### 9 3.2. Scanning electron microscopy and Energy Dispersive X-ray spectroscopy

10 As a result of the SEM analysis, panoramic images of each surface of the sample were  
11 obtained. At the magnification that was used, the pixel size in the images was 1.08  $\mu\text{m}$ .  
12 This is about one order of magnitude better than the resolution of the medium CT scan,  
13 i.e. compared to the 8.9  $\mu\text{m}$  voxel size of the CT scan of the entire sample. Contrast in the  
14 BSE images is visible between quartz and feldspars, albeit not very clear at the energy  
15 and spot size used. The dense titanium oxides show a very high contrast with respect to  
16 the other minerals, but are rather scarce in the 2D planes. EDS mappings revealed the  
17 distribution for silicon, potassium and iron. Grains which are very rich in silicon can be  
18 determined as quartz ( $\text{SiO}_2$ ) grains, zones rich in potassium are feldspars ( $\text{KAlSi}_3\text{O}_8$ ) or  
19 clay minerals (e.g. muscovite:  $\text{KAl}_2(\text{AlSi}_3\text{O}_{10})(\text{F,OH})_2$ ) and iron-rich zones are areas

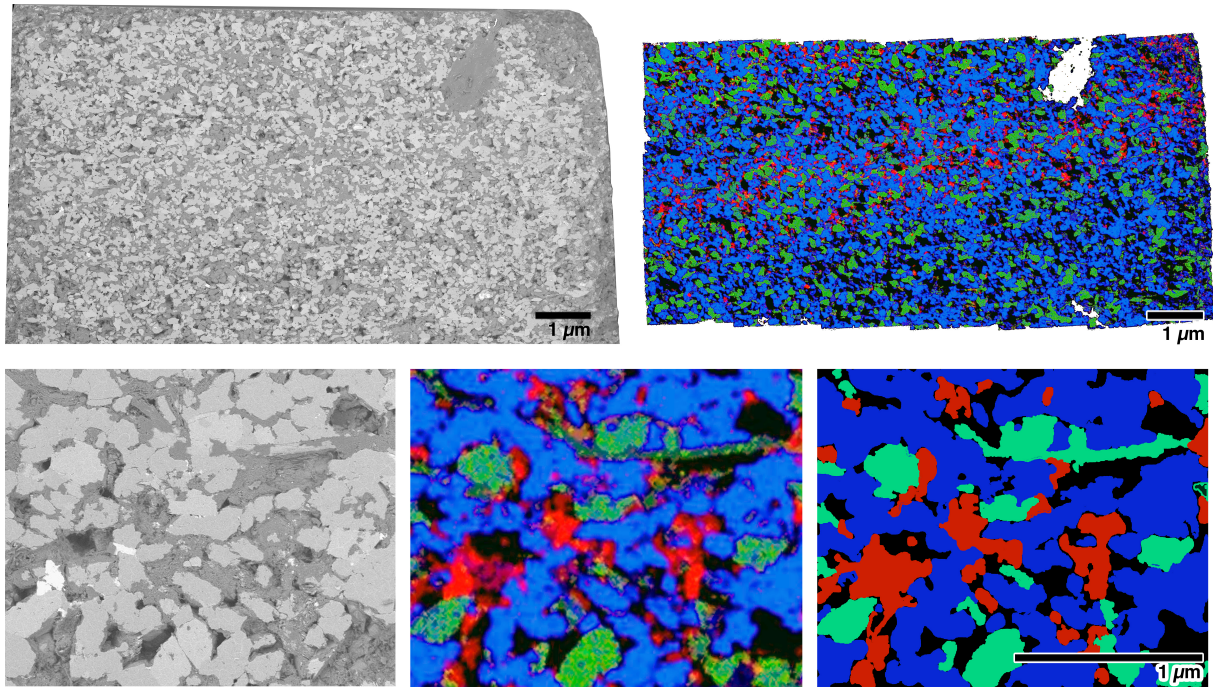
1 where iron oxides are present. The images indicate a rather uniform distribution of  
2 quartz, feldspars and clay minerals throughout the entire sample; however, the iron is  
3 mainly present in a distinct layer, mainly on two of the five surfaces. It is also apparent  
4 that most of the iron is not present in the form of grains, but is often located inside the  
5 cement that binds these grains together (Figure 3).

6 These results prove that SEM combined with EDS gives a good idea about the presence  
7 of different mineral phases inside the investigated sample, and that EDS is still a useful  
8 tool that can reveal the presence of different mineral phases, even if there is no  
9 compositional variance visible in SEM images. Furthermore, distribution of these phases  
10 in the 2D surfaces of the material is easily observed and analyzed.

### 11 **3.3. Image registration**

#### 12 *2D-to-2D*

13 2D-to-2D image registration between SEM and EDS images provides grain-to-grain  
14 information about mineralogy, even when contrast in BSE images is too low (Figure 3).  
15 Image registration was done using landmarks, which are very easy to detect, since both  
16 images have almost the same feature shape and the orientation of the images is the  
17 same. Due to the depth of view in SEM images (images show information beyond the  
18 absolute surface of the sample) it is very difficult to use automated algorithms. We  
19 remark that this step is not necessary on most modern SEM machines, as they can link  
20 SEM and EDS data automatically however it is necessary if the SEM machine is not  
21 equipped with an automated stage control, as was the case in this research. This method  
22 allows combination of SEM images with chemical information derived from other  
23 techniques as well, such as micro-X-ray fluorescence ( $\mu$ -XRF) (Boone et al., 2011).



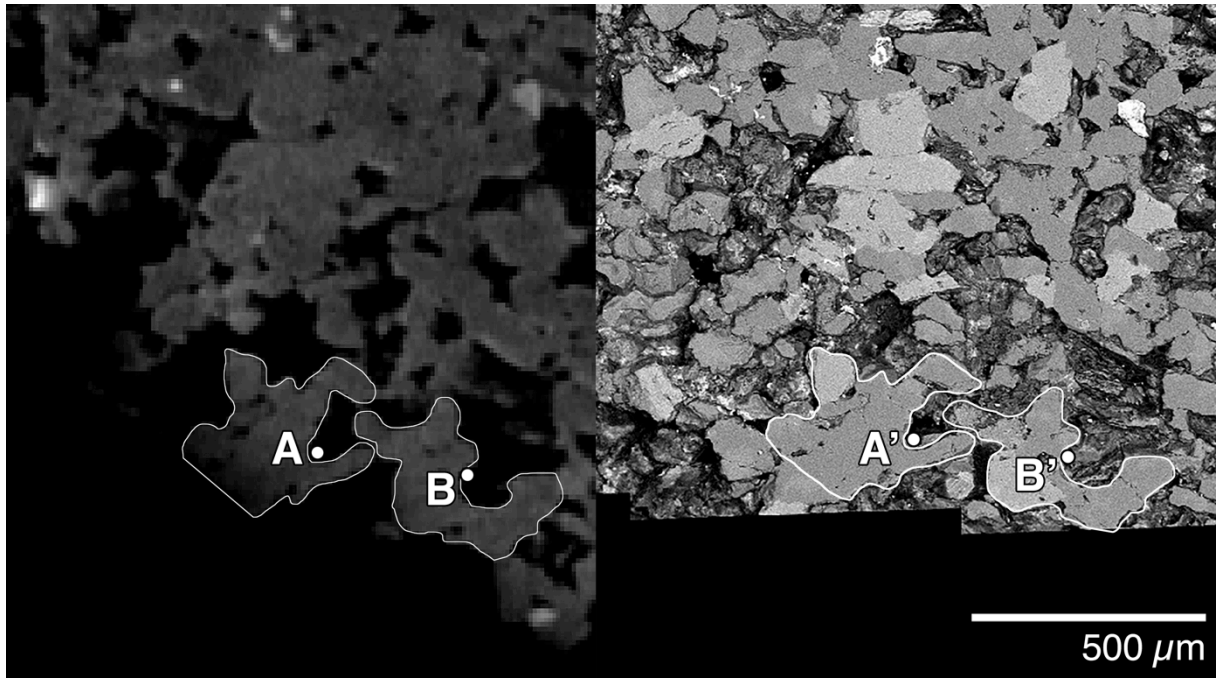
1  
 2 **Figure 3: Top: SEM image and corresponding EDS map. Blue = Quartz, Green = Clays and Feldspars, Red = Iron**  
 3 **oxides.**  
 4 **Bottom: Left: detail of SEM image; middle: detail of EDS map; right: simplified representation of EDS map.**  
 5 **Grains can be identified individually as different minerals.**

6 *2D-to-3D*

7 Although both CT and SEM/EDS images reveal a lot of information on the structure and  
 8 mineralogy of the sample, this information becomes much more useful if the orientation  
 9 and distribution of the different mineral phases can be linked directly to the 'low-  
 10 resolution' CT volume. This way, 3D orientation of certain layers can be observed and  
 11 measured. After registration of the 2D SEM and EDS images to the 3D volume, we can  
 12 see the orientation of the iron-rich layer that was described in the previous paragraph  
 13 (Figure 3 and Figure 7). Although the SEM and EDS images are in 2D, this method yields  
 14 a pseudo-3D look at the distribution of the iron inside the sample.

15 2D to 3D image registration of SEM (and EDS) images to the CT volume is not  
 16 straightforward, and besides the depth of view difference, corresponding features in  
 17 both image sets do not have the exact same form. This effect occurs since the angles of  
 18 the 14 mm sample were not completely straight, and SEM imaging was therefore not

1 always performed with a beam perpendicular to the sample surface. Furthermore, the  
2 direction of the CT slices in the vertical plains was never completely parallel or  
3 perpendicular to the samples surfaces Figure 4 shows that corresponding features can  
4 be recognized, but shapes and sizes in both images are not completely the same.  
5 Additionally, the visual information beyond the sample surface makes it difficult to  
6 directly link the SEM images to CT slices, as grain boundaries are far more difficult to  
7 detect in the SEM images. Landmarks were selected manually after visually looking for  
8 corresponding zones in the CT volume and SEM images (Figure 4). Landmarks in the CT  
9 volume were spread across 50 to 100 slices, since sample surfaces were not perfectly  
10 parallel to the orientation of the slices. Since manual placement of landmarks creates an  
11 inevitable error, about 100 locations per surface were selected as registration landmark,  
12 to average out this error. This proved to be enough for a qualitative grain-to-grain match  
13 between different datasets. Registration was done in two steps: first a rigid rotation was  
14 performed to rotate the SEM images in the same planes of the surfaces of the volume,  
15 using only four landmarks in the corners of both the SEM image and the corresponding  
16 surface in the SEM volume. Afterwards, an X/Y translation step was performed,  
17 combined with image warping to compensate for the difference in feature form in both  
18 imaging methods.



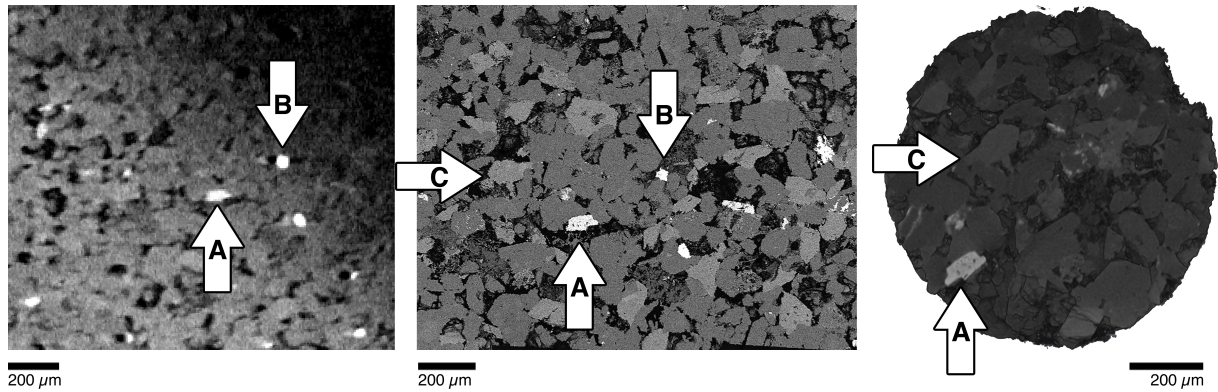
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2 **Figure 4: 2D to 3D image registration. White lines indicate corresponding areas in CT slice (left) and SEM**  
 3 **image (right). White dots indicate possible points for landmark placement (A, A' and B, B').**

4 *3D-to-3D*

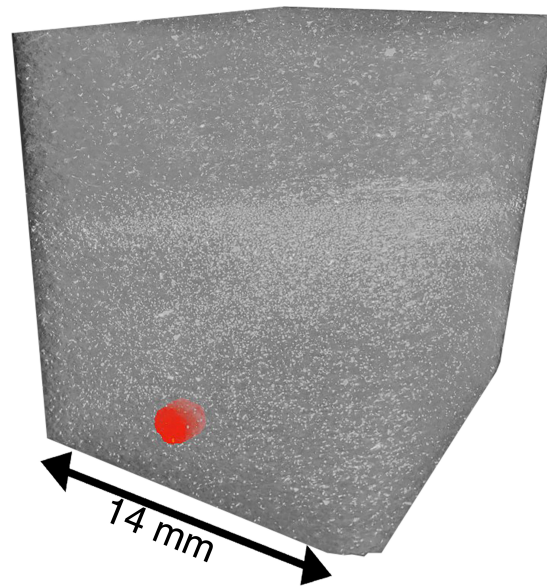
5 The final type of image registration links the reconstructed microplug CT scan to the  
 6 larger CT scan of the base sample (3D-to-3D registration). Since geological materials like  
 7 the Vosges sandstone and other construction materials such as concrete and bricks, are  
 8 heterogeneous in nature, it is important to pinpoint the location of a sub-sample. If a  
 9 sub-sample is very small (in this case, < 1 mm), the properties derived from this one  
 10 microplug might not be representative for the entire sample. As in the 2D-to-3D  
 11 registration step, semi-automatic landmark registration was performed to create the  
 12 alignment between the small and large 3D volume and this time, rigid transformation  
 13 was used to align the images. It should be stressed that the resolution in the lowest  
 14 resolution CT scan was insufficient for direct selection of these landmarks. Even more,  
 15 the 2D-to-3D registration of the SEM images to the low-resolution scan is an absolute  
 16 must if one wants to know the exact location of the microplug in the large sample. Figure  
 17 5 shows that first, landmarks A and B were selected in the low-resolution tomogram and

1 the SEM image. Finding landmark A in the high-resolution scan was much more easy, as  
2 spatial resolution and contrast are high enough in that volume. A combination of these  
3 points, together with some other points, made it possible to register the microplug into  
4 the cuboid.



5  
6 **Figure 5: 3D-to-3D registration using landmarks. Left: Low-resolution CT; center: Registered SEM image;**  
7 **Right: High-resolution CT. The SEM image in the middle is used as an intermediary image to be able to find**  
8 **corresponding points.**

9 Figure 6 shows the importance of this last image registration step. The small microplug  
10 is indicated in red inside the larger sample, which has been partially made transparent.  
11 High-density minerals are not distributed uniformly throughout the sample, and  
12 porosity is not the same throughout the entire sample. Indication of the location of the  
13 first subsample can help to set a strategy for further subsampling, or to determine which  
14 zones in the sample will have the same (flow) parameters as the microplug.



1

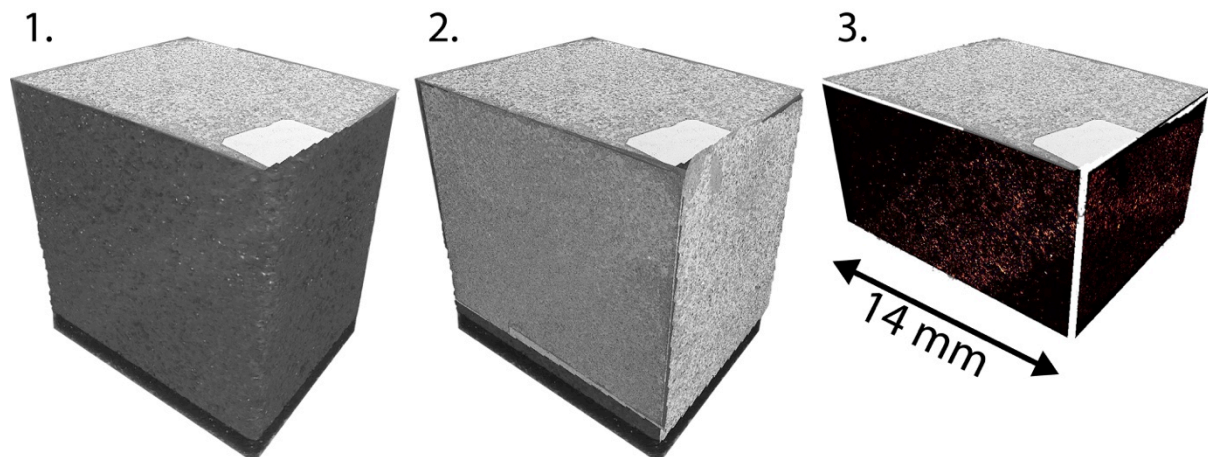
2 **Figure 6: Location of the microplug ( $\varnothing = 1$  mm) in the full sample (width = 14 mm). The partially transparent**  
3 **full sample reveals that dense minerals, i.e. represented in lighter grey, are unevenly distributed throughout**  
4 **the material.**

## 5 **4. Discussion**

6 *Location and orientation of different mineral zones inside a sample*

7 SEM combined with EDS gives useful information about the presence of different  
8 mineral phases inside the investigated sample. Furthermore, the distribution of these  
9 phases in the 2D surfaces of the material is easily observed and analyzed. When 2D  
10 mineral distribution maps are registered to the surfaces of a 3D volume, a pseudo-3D-  
11 look at the orientation and distribution of these minerals is obtained. This means that  
12 the iron-rich laminations can not only be observed, but also spatially oriented in the  
13 sample (Figure 7, 3). This orientation can be compared to that of other features in the  
14 sample, and extrapolated to the sample's interior.





1

2 **Figure 7: 1) Medium resolution CT volume with 1 SEM image registered to the top surface. 2) All SEM images**  
 3 **registered to the surfaces of the CT volume. 3) EDS map for iron registered to two sides of the sample,**  
 4 **showing the orientation of the observed iron oxide laminations.**

5 *Location of the sub-sample inside the base sample*

6 Since geological materials like the Vosges sandstone are heterogeneous in nature, it is  
 7 important to pinpoint the location of sub-samples in their parent sample. Zones with  
 8 different structures or composition cannot always be identified in a macroscopic way, so  
 9 verification of the sample position has to be done through image registration. The  
 10 medium resolution CT scan showed big regional differences in porosity and dense  
 11 mineral content (Figure 6), which means that it is important to know the location of the  
 12 small subsample inside the big dataset. The scans at different resolutions reveal that  
 13 multi-scale imaging is necessary to avoid inaccurate conclusions. Results illustrated that  
 14 when the resolution gap between parent sample and sub-sample is too large, finding the  
 15 location of the sub-sample in its parent is very hard to impossible, only based on the  
 16 images. In this case additional information coming from SEM or other complementary  
 17 techniques is a major aid, maybe even a necessity to make this 3D-to-3D image  
 18 registration work.

19 After localization of the subsample, porosity of that area in the parent sample could be  
 20 analyzed and was found to be 11.3 %, of which about 2 % was closed porosity.  
 21 Compared to the 9 % porosity of the overall sample, it is clear that the subsample was

1 taken in a region with more than average porosity. However, 16 % can be interpreted as  
2 the correct porosity value, as the analysis error on the low-resolution data is just to high.

### 3 **5. Conclusions**

4 At present, no image-based analysis technique can be considered as a multi-scale, all-  
5 area characterization technique for heterogeneous porous media. A combination of  
6 different imaging techniques, both 2D and 3D, followed by intensive post-processing of  
7 the acquired images can be used to provide more complete information on the analyzed  
8 material.

9 In the proposed methodology, medium-resolution  $\mu$ -CT results in three-dimensional,  
10 structural information of a large, centimeter-scale sample. Although the resolution is  
11 limited, large features can be identified, and interesting areas for further research are  
12 easily pinpointed. Data fusion of this medium-resolution 3D information with 2D  
13 imaging techniques such as SEM provide additional information in terms of structure.  
14 More importantly, the implementation of chemical analysis by means of EDS provides a  
15 pseudo-3D distribution of the different mineral phases present in the sample.  
16 Furthermore, SEM and EDS are mature and low-cost methods, accessible for most  
17 researchers in the field of materials characterization. Sub-sampling the material allows  
18 for imaging at higher resolutions using lab-based CT, but the limit of this technique still  
19 lays at a spatial resolution of about 0.5  $\mu\text{m}$ . Since this is not enough to resolve the finest  
20 features inside the Vosges stone, other methods should be addressed. Future research  
21 will include data analysis on the Vosges sandstone from FIB-nt, high resolution  
22 synchrotron X-ray  $\mu$ -CT and ptychographic tomography, focusing on the submicron  
23 porosity. FIB-nt, using secondary electron imaging, can achieve a spatial resolution of  
24 below 10 nm. High resolution  $\mu$ -CT at synchrotron facilities can provide images with

1 resolutions below 100 nm (Kastner et al., 2010) and with ptychographic tomography  
2 (Dierolf et al., 2010), resolutions of up to 16 nm in 3D have been reached (Holler et al.,  
3 2014). In combination with the study presented in this paper, this will provide a  
4 complete workflow for the analysis of the multiscale pore space of sandstones. In all  
5 cases, coupling between different methods is a necessity, since the heterogeneity of  
6 natural materials causes the need for the exact localization of sub-samples in their  
7 parent sample. Image registration techniques in two or three dimensions have therefore  
8 become very important in image processing.

9

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