

# Identification of Optimal Sampling Patterns for Compressed Sensing RARE MRI in Porous Media

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**Introduction:** The aim of this work is to acquire pore-scale MRI images and flow velocity maps within porous rocks to better understand structure-flow relationships and to aid the development of so-called digital rock (DR) flow simulators. Given the long acquisition times of such high-resolution images, compressed sensing (CS) is used for the MRI data acquisitions. Optimal  $\mathbf{k}$ -space sampling pattern design is a critical aspect of compressed sensing MRI (CS-MRI). Although various  $\mathbf{k}$ -space sampling strategies have been developed, they often require parameter optimisation to obtain optimal  $\mathbf{k}$ -space sampling patterns [1]. We developed a new, parameter-free  $\mathbf{k}$ -space sampling approach using input from high-resolution 3D X-ray micro-computed tomography ( $\mu$ CT) data sets to derive optimal  $\mathbf{k}$ -space sampling patterns for acquiring high spatial resolution 3D MRI images of rocks.

**Methods:**  $\mu$ CT images (grain space) of 4 mm diameter rock core plugs, routinely acquired as part of a DR workflow, were inverted to obtain pore space images. These were then Fourier transformed into the expected  $\mathbf{k}$ -space for these rocks. From these  $\mathbf{k}$ -space signatures, optimal CS variable-density sampling patterns [1] were generated for each rock type, spatial resolution, and sampling fraction. CS-RARE [2] in combination with the new  $\mathbf{k}$ -space sampling strategy was used to acquire pore-scale structural CS-MRI images of rocks at (isotropic) resolutions as high as 17.6  $\mu\text{m}$ . Pore size analysis in Avizo was used to benchmark the quality of the MRI images relative to a 5  $\mu\text{m}$   $\mu$ CT image.

**Results and discussion:** The  $\mu$ CT-based sampling strategy delivers a bespoke  $\mathbf{k}$ -space sampling pattern for each rock type (Fig. 1) according to the morphology of the rock, which was shown to give an optimal CS reconstruction quality for a range of different sampling fractions (0.125–0.375). The pore space analysis revealed excellent agreement between the pore size distributions of the acquired MRI data of Ketton

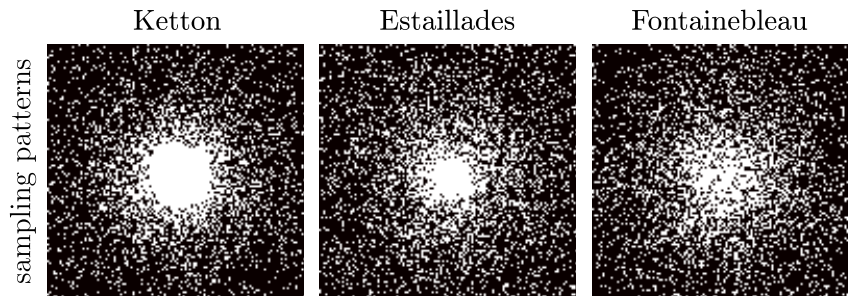


Fig. 1: Sampling patterns of Ketton limestone, Estailades limestone, and Fontainebleau sandstone as generated by the new,  $\mu$ CT-based sampling approach for a  $\mathbf{k}$ -space sampling fraction of 0.25. The white pixels in the sampling patterns represent the points sampled in both phase-encoding directions. For a 3D CS-RARE experiment, a line in the frequency-encoding direction, orthogonal to both phase-encoding directions, is fully sampled for each of these points.

rock using the  $\mu$ CT-based sampling approach and the 5  $\mu\text{m}$  resolution  $\mu$ CT images. These results highlight the advantage of using the  $\mu$ CT-based approach to deliver accurate pore space reconstructions at spatial resolutions for which a fully-sampled acquisition at those resolutions would be prohibitively long to acquire.

**Conclusion:** A novel, robust  $\mathbf{k}$ -space sampling strategy has been demonstrated to generate optimised  $\mathbf{k}$ -space sampling patterns for 3D high spatial resolution microstructural MRI acquisitions from 3D  $\mu$ CT data. This approach can be used to accelerate other MRI acquisitions relevant for DR applications, such as spatially-resolved propagators, fluid velocity maps, and relaxation time maps.

**References:** [1] Lustig, Magn. Reson. Med. (2007). [2] Ramskill, J. Magn. Reson. (2016).