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using cadmium fiducial markers**

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Correlative X-ray and neutron tomography of root systems using cadmium fiducial markers

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Abstract

The interactions between plant roots and soil are an area of active research, particularly in terms of water and nutrient uptake. Since non-invasive, *in vivo* studies are required, tomographic imaging appears an obvious method to use, but no one imaging modality is well suited to capture the complete system. X-ray imaging gives clear insight to soil structure and composition, however water is comparatively transparent to X-rays and biological matter also displays poor contrast with respect to the pores between soil particles. Neutron imaging presents a complementary view where water and biological matter are better distinguished but the soil minerals are not imaged as clearly as they would be with X-rays.

This work aims to develop robust methods for complementary X-ray/neutron tomographic imaging of plant root samples which should lead to new insight into water and nutrient transport in soil. The key challenges of this project are to develop experiments that will meet the requirements of both imaging modalities as well as the biological requirements of the plant samples and to develop ways to register a pair of reconstructed volume images of a sample that will typically have been produced with entirely separate facilities. The use of cadmium fiducial markers for registration has been investigated. Simulations were conducted to investigate the expected registration accuracy as the quantity and distribution of the markers varied. The findings of these simulations were then tested experimentally as plant samples were grown and imaged using neutrons with the IMAT instrument at ISIS Neutron and Muon Source at the STFC Rutherford Appleton Laboratory in Harwell, and with X-rays at μ -VIS X-ray Imaging Centre at the University of Southampton.

Lay Description

The interactions between plant roots and soil are an area of active research, particularly in terms of water and nutrient uptake. The samples used in this research are typically imaged so that they can be studied without digging up the roots and destroying the sample in the process. X-ray and neutron imaging techniques have both been used as each can show different materials within the sample. Since neither can show all the components of the system by itself, this work explores methods for combining scans of the same sample to give a more complete image of the system. In particular this

work focusses on the use of fiducial markers as a strategy for preparing the samples in such a way that the resulting images can be aligned. The effectiveness of this method was tested in simulation and then in practice. The samples used within this work were imaged using neutrons on the IMAT instrument at ISIS Neutron and Muon Source at the STFC Rutherford Appleton Laboratory in Harwell, and with X-rays at μ -VIS X-ray Imaging Centre at the University of Southampton.

1 Introduction

2 The human race is dependent upon plants to provide all food, either directly or indirectly. As the
3 earth's population increases, a corresponding growth in crop yield is required. Recent estimates
4 state that crop production will need to double by 2050 in order to keep pace with projected
5 population growth [1], a target that is not predicted to be achieved by current growth projections
6 [2]. Climate change will make this all the more difficult, in particular through reduced water
7 availability and the drive to reduce fertiliser usage [3]. The *green revolution* is the name given to a
8 period in the mid twentieth century when a number of scientific advancements, including the
9 introduction of fertilisers and genetic modification, led to a tremendous gain in crop yields in a
10 relatively short period of time [4]. The green revolution was primarily centred on the manipulation
11 of the portion of the plant that is visible above the ground and roots were largely overlooked. The
12 root system is central to plant functions such as water and nutrient uptake, anchorage and
13 interaction with symbiotic organisms and so it has been recognised that root growth and
14 development could be further exploited to maximise crop yield [5]. It has been suggested that the
15 deployment of crops with more efficient water and nutrient uptake due to improved traits below the
16 ground could lead to a second green revolution and help to address the world wide challenge of
17 food security [6].

18 Non-invasive, *in vivo* studies of plant roots present a challenge that has traditionally been addressed
19 through rhizotrons (containers that force 2D growth conformation with transparent windows for
20 observation) or transparent, artificial growth media. Although widely used, these methods generally
21 result in root systems that vary considerably from those grown in natural soil conditions [7]. X-ray
22 computed tomography is the primary method for 3D imaging of root systems grown in soil but is not
23 without its limitations [8]. While X-rays show the soil structure and composition well, they are not an
24 ideal tool for imaging water distribution in particular since there is very little contrast between
25 water, plant roots and any other biological soil constituents. None of these hydrogen-rich
26 constituents are resolved clearly and as a result it is difficult to differentiate between them in the
27 image data produced.

28 Neutron imaging offers a solution as many of the imaging techniques are similar but the mechanisms
29 by which neutrons interact with matter are very different and hence different elements, in particular
30 light elements such as hydrogen, provide strong contrast in a neutron image [9]. A number of
31 experiments have shown neutron imaging to be well suited to showing water dynamics, where X-ray
32 imaging would have struggled [10-15]. Neutron imaging has its own limitations however: it is slower,
33 less readily available and produces images of lower resolution when compared to equivalent X-ray
34 techniques. Just as X-ray imaging cannot provide a good representation of all the materials in a plant
35 sample, neutron imaging struggles to show the soil minerals and structure clearly.

36 Registration

37 Bi-modal datasets require registration to align the data from each modality. Although some cases
38 have successfully used mutual information in the sample to achieve registration [16], in general it
39 has been shown that, due to the complementary nature of the modalities, there is no guarantee that
40 there will be similar local features in corresponding datasets (particularly with multi-phase images)
41 and it is therefore difficult to find a good registration solution based on common features within a
42 sample [9]. To overcome this difficulty, fiducial markers can be attached to a sample to aid in
43 registration. A fiducial marker is an object placed within an image to be used as a point of reference.

44 To register volumes using fiducial markers, at least three fiducial points must be selected in both the
 45 reference and target images so that the registration parameters can be found and a suitable
 46 transform applied. In the case of CT volume data, this is typically achieved through an affine
 47 transform.

48 Inevitably, there will be error in the registration process. Maurer et al. defined three types of error
 49 that can occur when using fiducial points for registration [17]:

- 50 • Fiducial localisation error (FLE): the average error in locating the position of the fiducial
 51 points. (Figure 1 (A))
- 52 • Fiducial registration error (FRE): the root mean square (RMS) error between corresponding
 53 fiducial points after registration. (Figure 1 (B))
- 54 • Target registration error (TRE): the error between corresponding points other than the
 55 fiducial points after registration. (Figure 1 (C))

56 Fitzpatrick and West built on these definitions by providing expressions for the expected FRE and
 57 TRE errors in terms of the expected FLE and the set of fiducial points used (eq. 1, eq. 2) [18]. From
 58 this expression, Wang and Song derive an equation (eq. 3) to relate TRE to a particular distribution of
 59 markers that are independent of FLE or FRE. This is achieved by using eq. 1 to substitute FRE for FLE
 60 before setting FRE to 1. They propose a deterministic, optimisation method for determining the
 61 quantity and layout of markers to minimise TRE at a point of interest r [19]. The distance from r to
 62 each axis is denoted as d_k , where $k = (1, 2, 3)$ and f_k is the RMS distance of all fiducial points to the k^{th}
 63 coordinate axis.

$$64 \quad E\{FRE^2\} \approx \frac{(N-2)E\{FLE^2\}}{N} \quad (eq.1)$$

$$65 \quad E\{TRE^2(r)\} \approx \frac{E\{FLE^2\}}{N} \left(1 + \frac{1}{3} \sum_{k=1}^3 \frac{d_k^2}{f_k^2} \right) \quad (eq.2)$$

$$66 \quad TRE_M(r) = \frac{1}{N-2} \left(1 + \frac{1}{3} \sum_{k=1}^3 \frac{d_k^2}{f_k^2} \right) \quad (eq.3)$$

67

68 X-ray and neutron imaging equipment

69 This project is a pathfinder application for the new IMAT instrument at ISIS Neutron and Muon
 70 Source. IMAT is a combined cold neutron imaging and diffraction instrument designed to take
 71 advantage of the second ISIS target station to provide neutron radiography, neutron tomography,
 72 energy-selective neutron imaging and spatially resolved diffraction scans [20]. The instrument
 73 construction was completed in 2016 and since then IMAT has been running its imaging configuration
 74 [21].

75 ISIS TS-2 is a short-pulse source which operates at 40 kW and delivers pulses at a rate of 10 Hz. IMAT
 76 uses a cold (20 K), coupled liquid hydrogen moderator to slow the neutrons. A straight, 44 m
 77 supermirror neutron guide transports the neutrons from the target to the experimental area. Three
 78 choppers are placed within the guide to filter the beam. A T0 chopper removes fast neutrons and
 79 gamma radiation, then a pair of double-disk choppers define the wavelength band to ensure there is
 80 no frame overlap between successive neutron pulses. At the end of the guide is a pinhole selector
 81 that allows the aperture diameter (D) to be varied between five values to define different L/D ratios,
 82 where L (the distance from the aperture to the sample) is 10 m [22]. This results in a total flight path
 83 of 56 m to the sample. Between the aperture and the sample, the beam travels through a 9 m

84 evacuated flight tube and is shaped by five sets of jaws [23]. The sample is placed on a combined
85 translation and rotation system that is rated for up to 1.5 tonnes.

86 IMAT has a variety of detectors but this work made use of IMAT's optical detector. This is based on a
87 Zyla sCMOS 4.2 Plus camera in an optical camera box with a field of view varying between 50×50
88 mm^2 and $200 \times 200 \text{mm}^2$ and acts as an integrating detector with a range of scintillators [24].

89 μ -VIS' modified Nikon/X-Tek HMX (225kVp) was used to produce the X-ray data for this work. This is
90 a customised, general purpose X-ray CT and radiographic inspection system. It can take samples up
91 to 300 mm in height although this is reduced to approximately 150 mm if the robotic sample
92 exchanger is used. The 225 kVp X-ray source can be configured for high resolution or high flux by
93 using different anodes and uses a PerkinElmer PE1621 flat panel detector to capture the image. The
94 detector is made up of a 2000 x 2000 matrix of amorphous silicon pixels with discrete gadolinium
95 oxysulphide scintillators.

96 Simulation

97 In order to determine the best configuration of fiducial markers before imaging the samples, a
98 simulation was conducted based on the method presented by Wang and Song [19]. The method was
99 adapted to the case where the sample is cylindrical and the registration is to be optimised over the
100 whole volume of the sample rather than at a single point of interest.

101 $TRE_M(r)$ is a dimensionless quantity that relates the expectation value of TRE to that of FRE at a point
102 r , for a given set of markers M . Potential marker locations were given by a grid of points on the
103 surface of the cylinder. By evaluating $TRE_M(r)$ for a set of points evenly distributed throughout the
104 volume and taking the RMS value, an estimate of the TRE_M value is calculated for the whole volume.
105 This means that TRE_M gives a measure of how well two volumes can be registered with a set of
106 markers M independently of the FLE and FRE.

107 Markers are added to the simulation one by one, minimising the TRE_M each time until the desired
108 number of markers is reached. From this point the marker with the highest contribution is
109 repositioned to reduce TRE_M . This step is repeated until the repositioned marker still gives the
110 largest contribution to TRE_M .

111 The simulation was run for $N = \{4, \dots, 16\}$, where N is the number of fiducial markers, to establish
112 how many fiducial points should be used and how they should be distributed about the sample.
113 Figure 2 shows how the value of TRE_M falls as the number of fiducial markers is increased. The
114 significance of each marker to the accuracy of the registration (C_{fim}) also falls quickly as N is
115 increased as shown in figure 3. This was calculated by taking the RMS average of the change in TRE_M
116 when each marker is removed from M . This trend confirms that for a larger number of markers, each
117 marker is less significant so less error is introduced if a marker cannot be accurately located or used
118 for the registration.

119 Figure 4 shows the optimal distributions found by the simulation for $N = 10, 16$. A number of trends
120 can be observed from the distributions of markers for different values of N and applied to the
121 general distribution of fiducial markers around a cylinder.

122 The first clear pattern is that the markers are distributed evenly between the very top and the very
123 bottom of the sample – maximising the distance between the markers and the centre of the volume.
124 This is a simple principle to apply when attaching markers to the sample and also has practical
125 benefits for the imaging process since it means that the markers can be placed higher and lower
126 than the soil in the sample tube. This means that areas of photon or neutron starvation, that could
127 produce artefacts in the reconstructed data due to the markers, can be located away from the
128 region of interest as the soil sample will not fill the very top and bottom of the sample tube.

129 Another clear pattern is that the markers are placed on a 180° arc at the top and another
130 corresponding arc at the bottom rather than, say, alternating between the top and bottom of the
131 cylinder. The markers are also evenly distributed around the cylinder, ensuring that the centroid of
132 the fiducial points is close to the centre of the sample.

133 Following these results from the optimisation algorithm a number of particular distributions were
134 chosen to further evaluate these patterns. By evaluating distributions with an even distribution of
135 markers between the top and bottom it was found that this generates very similar results. When
136 compared to arcs at the top and bottom, with the same positions in X and Y, the results were equal
137 to 2 significant figures and the small variations at greater precision did not conclusively show either
138 arrangement to be consistently better than the other for all values of N. It seems reasonable that the
139 simulation gives arcs because it begins the optimisation with a small number of markers and adds
140 optimised markers iteratively up to the required total rather than started with N markers and
141 attempting to redistribute them all.

142 The optimal arrangements found by the simulation were compared to random arrangements to see
143 how significant the differences are. 1,500,000 unique, random arrangements were evaluated for
144 each value of N. Figure 5 shows the minimum, mean and maximum values from these tests as well
145 as the optimal values (as seen in Figure 2). Figure 5a shows that a poor arrangement can increase
146 the error by as much as an order of magnitude when very few markers are used. With higher
147 numbers of markers however it becomes clear that the variation between good and bad
148 arrangements becomes insignificant. Figure 5b shows the same data in the range where N varies
149 between 12 and 16. It can be seen that the optimal solutions found by the simulation are better than
150 any of the values found in the 1.5 million random arrangements but not by a significant amount. On
151 average, a random distribution of N+1 markers will give better results than the optimal distribution
152 of N markers.

153 Equation 1 states that the expectation of target registration error squared will be proportional to the
154 FRE, the number of markers and the distribution of the markers. The simulation results show that
155 with only four markers the TRE can be reduced to half of the FRE and that it can be reduced to less
156 than a tenth of the FRE by using more than twelve points. Using a high number of markers makes a
157 clear improvement to the result and also allows for the potential loss of a point without seeing a
158 large drop in accuracy.

159 Materials and Methods

160 A set of samples were imaged using both X-ray and neutron tomography and then registered to test
161 the proposed registration scheme and demonstrate the complementarity of the two modalities for
162 further studies.

163 Cadmium was proposed as a suitable material for fiducial markers since it has a large attenuation
164 coefficient for both neutrons and X-rays, allowing easy segmentation in either modality. An initial X-
165 ray scan, to confirm the suitability of the sample tube and markers, showed that the cadmium
166 produced significant artefacts, primarily due to beam hardening and scattering. It was decided that
167 the artefacts could be reduced to a satisfactory level provided the marker size was minimised and
168 there were sufficient variations in the height of the markers to avoid streaking between two markers
169 in a slice. Smaller cadmium pieces were cut with a maximum thickness of 1 mm and a new sample
170 tube was scanned to ensure these changes were sufficient.

171 Once the sample tube and marker configuration were shown to be suitable for imaging, a set of
172 plants were grown and imaged using both IMAT and the HMX at to produce a set of complementary
173 volumes that could be used to develop and test registration techniques. Twenty lupine seeds were
174 placed in wet paper towels to germinate. After six days, eight seeds that had begun to sprout were
175 selected and transferred to the sample tubes. New sample tubes were introduced for this

176 experiment that were made from boron free quartz with an inner diameter of 14 mm and a wall
177 thickness of 1.5 mm. Each tube had a single fibreglass wick to draw water for the plant. Within the
178 tube the wick was surrounded by 1 tsp of sand with particle sizes between 1.18 mm and 0.6 mm. Soil
179 of the same particle size range was then used to fill the tube, covering the seed around 20 mm from
180 the top. Each sample had a different quantity or arrangement of fiducial markers, as listed in Table 1.
181 These were selected to allow the simulation results to be compared to measured data. Due to beam-
182 time limitations, samples 2 and 6 were not scanned. The samples were left to grow for five days
183 before the scans began. The neutron scans were conducted first over two days and the X-ray scans
184 were collected over the following two days.

185 The neutron scans were conducted using the optical camera with a 135 mm lens, this gave a FOV of
186 60 x 60 mm. A 60 μm ZnS/LiF scintillator with a surface area of 90 x 90 mm was used. The beam was
187 shaped using the 40 mm pinhole and the jaws were set from 40 mm to 70 mm to match the beam
188 profile to the FOV as closely as can be achieved without introducing artefacts. The samples were
189 positioned 15 mm from the detector and 964 projections were taken with an exposure time of 30 s
190 per projection. The projections were reconstructed with a voxel size of 31 μm using the filtered back
191 projection algorithm in Octopus Reconstruction versions 8.9.3.4 and 8.9.4.2 [25]. The X-ray scans
192 were conducted in the HMX at 80 kVp and 87 μA . 1571 projections were taken with four frames and
193 a 500 ms exposure time. The projections were reconstructed using Nikon CT Pro 3D version
194 2.2.5386.22184 giving a voxel size of 19 μm .

195 Once the scans were reconstructed, registration was attempted by segmenting the cadmium pieces
196 and taking their centres as fiducial points, before finding and applying the affine transform to best
197 match these points. The cadmium centres were located using a threshold segmentation in
198 conjunction with FIJI's 3D Objects Counter [26]. An affine transform can be determined to match the
199 two point sets. This is achieved by removing the translation and scaling differences by centring the
200 two points sets and using orthogonal reduction before the rotation component is determined using
201 Horn's algorithm [27]. Once the rotation has been found, the scaling and translation components
202 can be determined easily [28]. After the transform was applied, the volumes were then cropped to
203 matching dimensions. In order not to discard the higher resolution data in the X-ray scan, the
204 neutron data was considered the target volume. As a result it was scaled and hence resampled at a
205 higher resolution than IMAT could have achieved.

206 Results

207 Figures 6 and 7 show example slices and volumes from the reconstructed data collected using
208 neutron and X-ray imaging respectively.

209 Following the registration process, the fiducial markers were then resegmented in each modality and
210 the new positions compared to give the FRE for each sample. This can be taken as a reasonable
211 indicator of registration accuracy in the case of an affine transform. The FRE values are shown in
212 Table 1. The mean FRE of sample 5 was so much higher than that of the other samples due to one
213 marker which was misaligned by 18.5 voxels. The mean FRE without taking that marker into account
214 is only 4.5, which is far closer to the typical FRE values produced by the other samples. This error
215 was introduced as a result of a large FLE for this marker in the X-ray data due to an artefact. Figure 8
216 shows how the threshold segmentation detects an area far greater than that of the cadmium piece
217 due to the artefacts surrounding it. This increase in the volume of the segmented marker pulls the
218 measured centre away from the true value and perhaps more importantly away from the position
219 found in the complementary modality to which it is to be matched.

220 Figure 9 shows a slice from sample 8 after registration. In addition to the combined data, each
221 modality is shown separately. The side by side and overlaid comparison clearly shows the differences
222 in contrast and signal to noise ratio between the two modalities but also the accuracy of the
223 registration – in particular when observing the aluminium tape around the outside of the tube. This

224 image also allows the complementarity to be seen clearly. Not only does the neutron data show the
225 plant root with greater contrast than the X-ray data but it shows some of the soil particles that
226 appear in the X-ray data while omitting others. This means that the combined data can be used to
227 infer information about the different materials making up soil particles which could not be
228 distinguished using X-rays alone.

229 Discussion and Conclusions

230 The data collected in this experiment demonstrates the suitability of X-ray and neutron tomography
231 for multi-modal studies, particularly into plant soil systems. In addition we have shown that fiducial
232 markers and the registration algorithms used allow the data from the two modalities to be
233 registered accurately, overcoming a lack of clear mutual information in the sample, and it can be
234 seen that more information can be taken by combining techniques than could be collected from
235 either technique in isolation.

236 There is no correlation between the FRE of the samples and the number and distribution of fiducial
237 markers used. For example, sample 1 has the second lowest FRE but had the worst set of fiducial
238 markers since it had the fewest markers and their arrangement was close to co-linear. Sample 5 had
239 the greatest number of markers and therefore had the best set of fiducial points, but it showed the
240 worst FRE of all the samples. The product of the FRE and TRE_M values (Table 1) gives an estimate for
241 the TRE for each sample. The variation FRE is far more significant in this calculation than that of TRE_M
242 which shows that being able to accurately image and locate the markers is more important to the
243 overall registration accuracy than having the markers ideally distributed.

244 The higher FRE of sample 5 indicates that artefacts in the scans which affect the segmentation of the
245 fiducial markers are the greatest source of error in registration (Figure 8). Cadmium's high
246 attenuation introduces artefacts and these have been shown to have a negative impact on the
247 registration accuracy such that the effect of marker arrangement could not be tested in detail. This
248 suggests that while it can be used as a fiducial marker material, it is not ideal. A material that would
249 attenuate both X-rays and neutrons less would result in reduced artefacts, leading to not only better
250 scan data but more accurate registration since large FRE values as a result of image artefacts have
251 been shown to be the primary contributor to FRE in the registered data.

252 Further studies are being planned that will test the suitability of other materials, such as borosilicate
253 as fiducial markers.

254 Movement and other changes in the samples can be observed in the registered results, as a result of
255 the time and travel between the two scans of each sample. In particular, the seedlings begin to
256 droop and the top layers of soil can move considerably between the two scans. This will be
257 addressed in future studies by booking time to conduct both scans at the same site with minimal
258 delays between them. In addition, these studies will integrate synchrotron X-ray imaging which will
259 improve the resolution of the data and allow more complex imaging techniques to be introduced.

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Figure 1. *The three types of registration errors: FLE, FRE and TRE. Based on Figure 1 of [29].*

Figure 2. *The change in target registration error as the number of fiducial points is increased.*

Figure 3. *The average contribution of a single marker to the target registration accuracy as the number of markers increases.*

Figure 4. *The optimal fiducial marker distributions suggested by the simulation for $N = 10$ (top) and $N = 16$ (bottom).*

Figure 5. *A comparison of the optimal marker distributions to the mean, maximum and minimum TRE_M values found in 1,500,000 random trials. 5a (top) shows the full range of N values. 5b (bottom) shows $N = 12 - 16$.*

Figure 6. *A slice and partial volume render from the neutron scan of sample 7.*

Figure 7. *A slice and partial volume render from the X-ray scan of sample 8.*

Figure 8. *Sample 5 had the worst registration accuracy despite having the most fiducial markers. This figure shows how an artefact prevents accurate segmentation (segmentation shown in red) of a cadmium piece, creating a large FLE for the corresponding fiducial marker.*

Figure 9. *A slice from sample 8 with the X-ray data on the left and in red and the neutron data on the right and in green. This slice shows the match of a fiducial marker, the accuracy of the registration and the complementarity of the modalities.*

Table 1. *The fiducial marker arrangements, FRE measurements and TRE_M values for each of the sample tubes that was scanned.*

Lay Description

The interactions between plant roots and soil are an area of active research, particularly in terms of water and nutrient uptake. The samples used in this research are typically imaged so that they can be studied without digging up the roots and destroying the sample in the process. X-ray and neutron imaging techniques have both been used as each can show different materials within the sample. Since neither can show all the components of the system by itself, this work explores methods for combining scans of the same sample to give a more complete image of the system. In particular this work focusses on the use of fiducial markers as a strategy for preparing the samples in such a way that the resulting images can be aligned. The effectiveness of this method was tested in simulation and then in practice. The samples used within this work were imaged using neutrons on the IMAT instrument at ISIS Neutron and Muon Source at the STFC Rutherford Appleton Laboratory in Harwell, and with X-rays at μ -VIS X-ray Imaging Centre at the University of Southampton.

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Table 1

SAMPLE	FIDUCIAL ARRANGEMENT	FRE	TRE _M	TRE (FRE × TRE _M)
1	5 markers in a vertical line	2.66	0.33	0.89
3	6 markers in arcs at the top and bottom of the FOV	6.49	0.25	1.62
4	8 markers in arcs at the top and bottom of the FOV	2.03	0.17	0.34
5	10 markers in arcs at the top and bottom of the FOV	11.02	0.13	1.38
7	9 markers in arcs at the top, middle and bottom of the FOV	4.33	0.14	0.62
8	9 markers in arcs at the top, middle and bottom of the FOV	2.85	0.14	0.41

Table 1. The fiducial marker arrangements, FRE measurements and TRE_M values for each of the sample tubes that was scanned.

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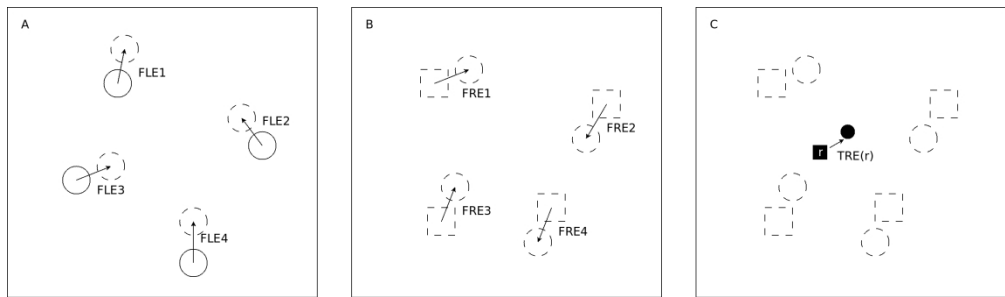


Figure 1 The three types of registration errors: FLE, FRE and TRE. Based on Figure 1 of [29]

768x223mm (96 x 96 DPI)

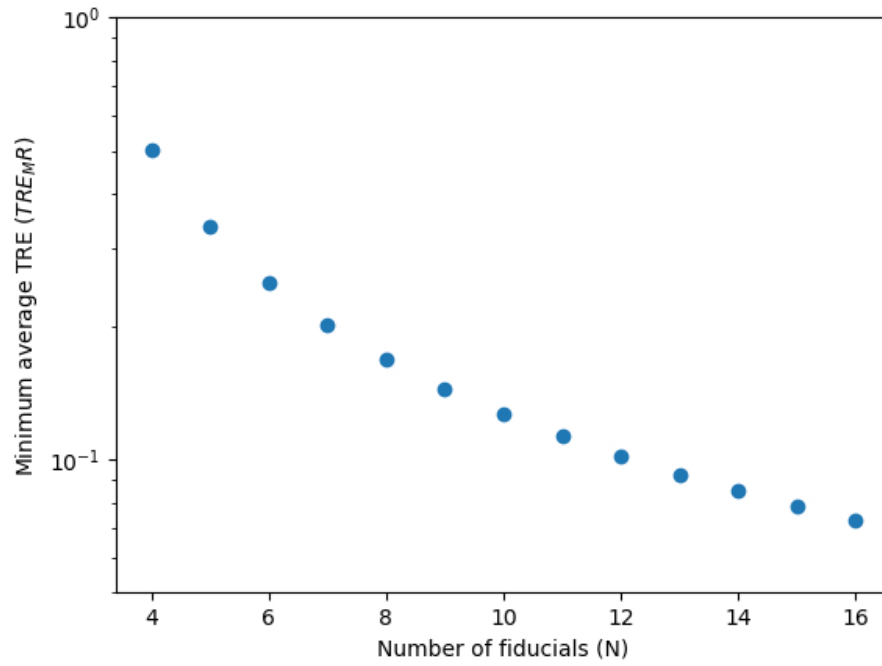


Figure 2 The change in target registration error as the number of fiducial points is increased.

162x121mm (100 x 100 DPI)

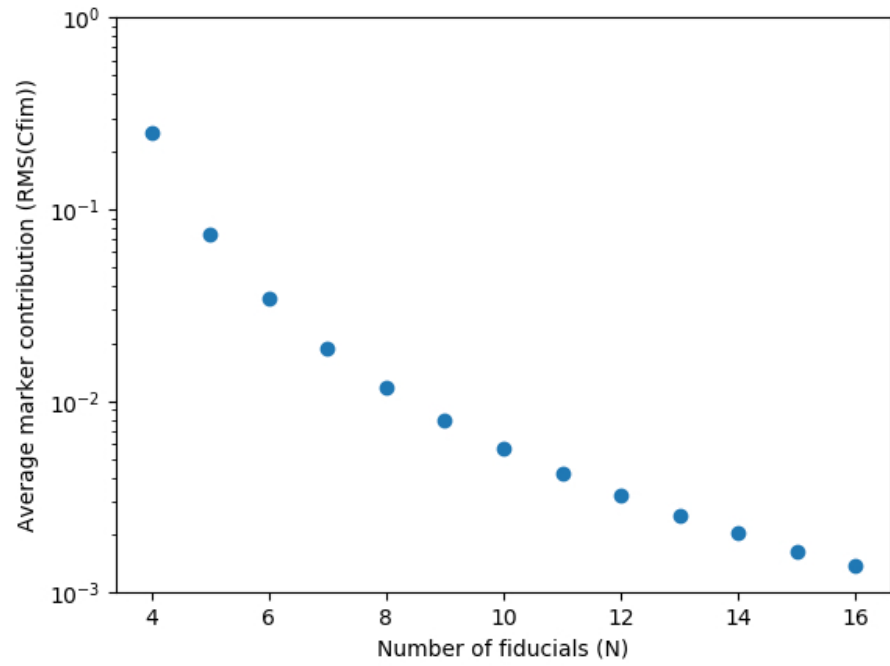


Figure 3 The average contribution of a single marker to the target registration accuracy as the number of markers increases.

162x121mm (100 x 100 DPI)

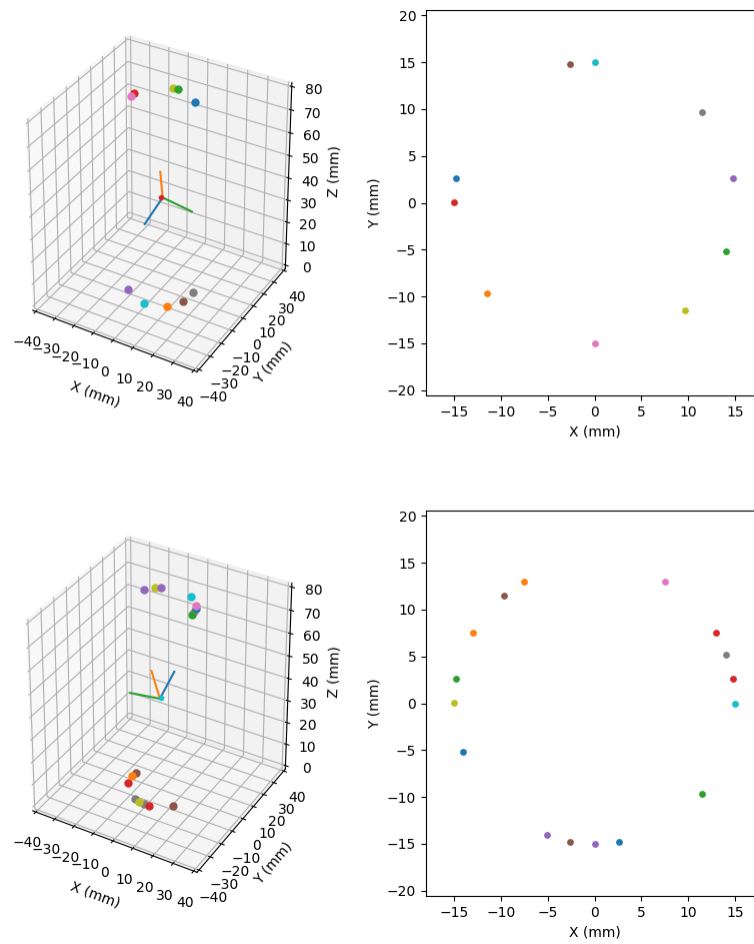


Figure 4 The optimal fiducial marker distributions suggested by the simulation for $N = 10$ (top) and $N = 16$ (bottom).

254x254mm (100 x 100 DPI)

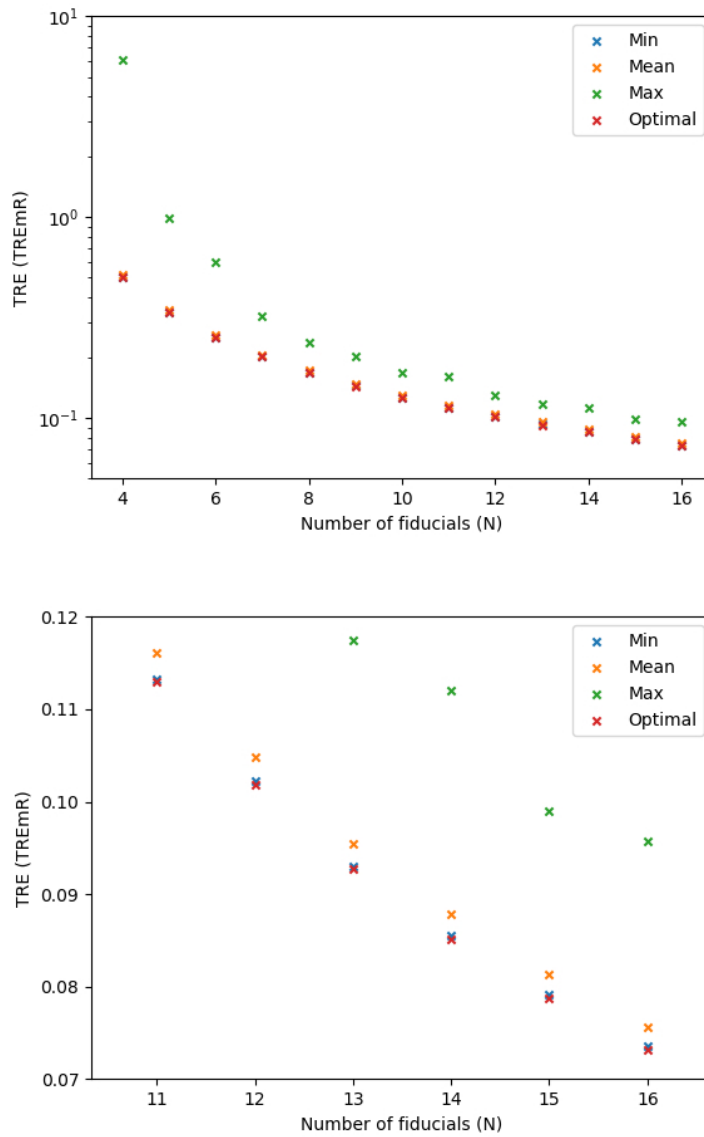


Figure 5 A comparison of the optimal marker distributions to the mean, maximum and minimum TRE_M values found in 1,500,000 random trials. 5a (top) shows the full range of N values. 5b (bottom) shows N = 12 – 16.

162x242mm (100 x 100 DPI)

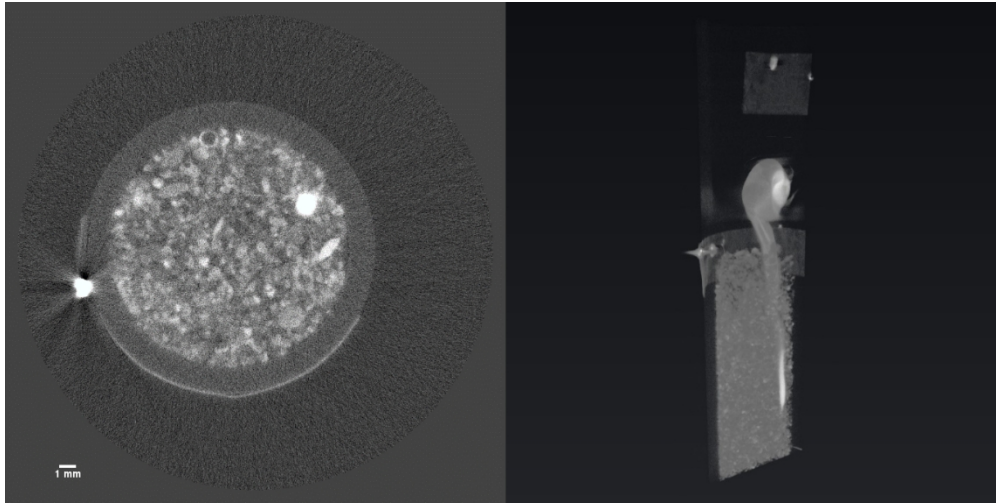


Figure 6 A slice and partial volume render from the neutron scan of sample 7.

1508x755mm (23 x 23 DPI)

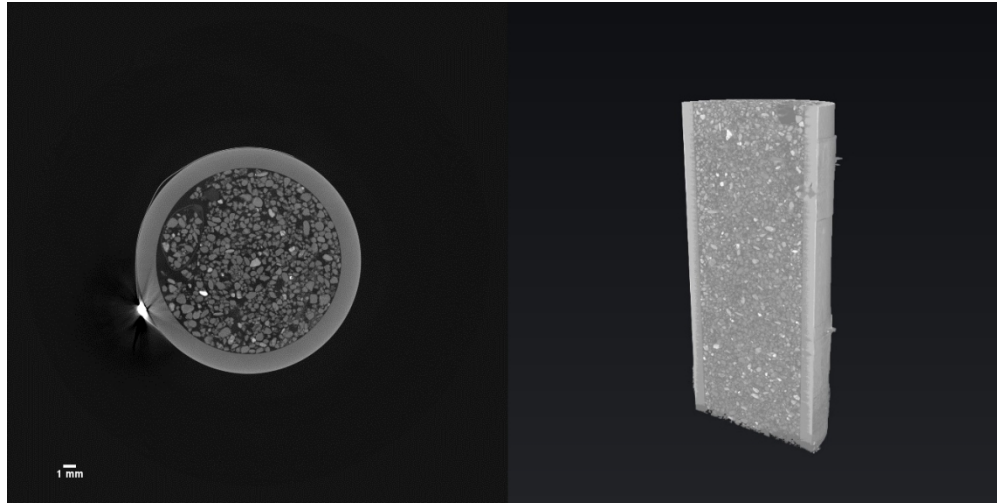


Figure 7 A slice and partial volume render from the X-ray scan of sample 8.

1993x996mm (17 x 17 DPI)

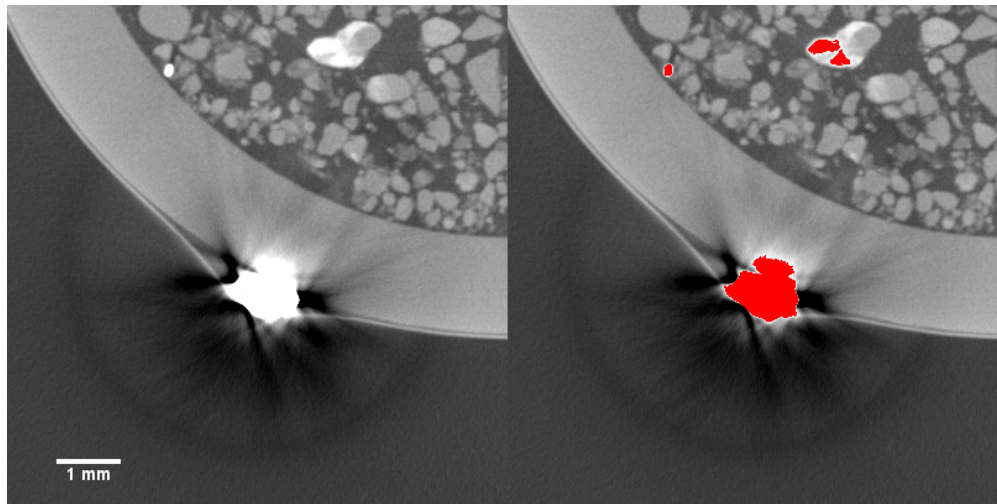


Figure 8 Sample 5 had the worst registration accuracy despite having the most fiducial markers. This figure shows how an artefact prevents accurate segmentation (segmentation shown in red) of a cadmium piece, creating a large FLE for the corresponding fiducial marker.

395x197mm (52 x 52 DPI)

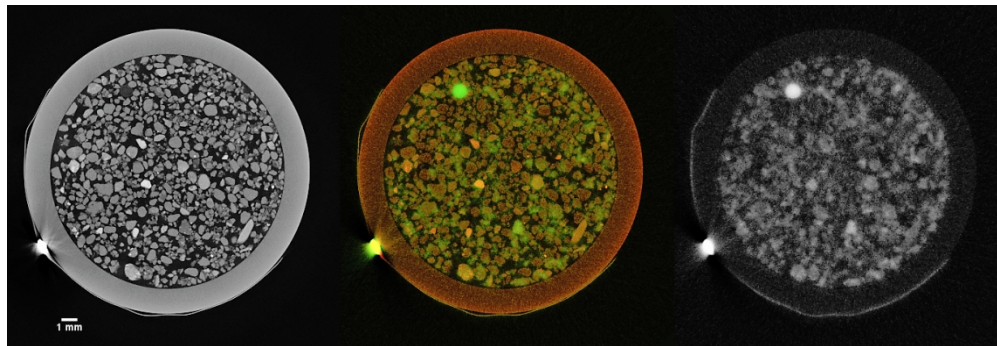


Figure 9 A slice from sample 8 with the X-ray data on the left and in red and the neutron data on the right and in green. This slice shows the match of a fiducial marker, the accuracy of the registration and the complementarity of the modalities.