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Development of muon scattering tomography for a detection of reinforcement in concrete

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Abstract
Inspection of ageing, reinforced concrete structures is a world-wide challenge. Existing non-destructive evaluation techniques in civil and structural engineering have limited penetration depth and don’t allow to precisely ascertain the configuration of reinforcement within large concrete objects. The big challenge for critical infrastructure (bridges, dams, dry docks, nuclear bioshields etc.) is understanding the internal condition of the concrete and steel, not just the location of the reinforcement. In most new constructions the location should be known and recorded in the as-built drawings, where these might not exist due to poor record keeping for older structures. Muon scattering tomography is a non-destructive and non-invasive technique which shows great promise for high-depth 3D concrete imaging. Previously, we have demonstrated that individual bars with a diameter of $33.7 \pm 7.3$ mm can be located using muon scattering tomography. Here we present an improved method that exploits the periodicity of bar structures. With this new method, reinforcement with bars down to 6 mm thickness can be detected and imaged.

1. Introduction
Old reinforced concrete structures may need to be inspected or replaced. Knowing the location of the steel is the first step towards determining the condition of the reinforcement. The key for assessing and substantiating the structure for life extensions is being able to state the design was built as planned (location and size of reinforcement) to a high quality (no voids from construction) and that the internal condition is satisfying (not degraded—reinforcement corrosion, cracking—beyond a critical value). Current non-destructive evaluation (NDE) scanning technology is based on the detection of reflected or transmitted electromagnetic, thermal or acoustic waves generated by a local source. The most used NDE techniques are magnetic imaging and ground penetrating radar (GPR) [1–4], which can image bars with the diameters of 10–20 mm at depths of 100–500 mm. Low depth imaging, for depths at < 20 cm, can be performed with infrared thermographics [5, 6] and ultrasonics [7, 8]. These techniques are suitable for assessing element thickness and bar location, but precise estimation of bar size is still an unsolved problem [2]. Furthermore, a detailed testing below the first or second reinforcement layer is often beyond the bounds of possibility because errors greatly increase with penetration depth and number of bars [4]. Other NDE methods currently used are x-ray and neutron radiography [9, 10], which enable a high resolution and a high depth scanning. However, the use of active sources of radiation is a serious threat to human health.

Muon tomography is being investigated for many different challenges, relying on both Monte Carlo simulation studies and experiments. A traditional application of muon tomography is the characterization of
nuclear waste drums and related security applications, where contents of concrete or bitumen filled waste drums are studied. Key issues here include the potential presence of gas bubbles in the matrix of the waste drum [11] and identification of the material inside the drums [12–16]. Security applications have been mainly focused on detection of high-Z material in cargo containers [17, 18], but work on the detection of explosives is ongoing as well [19]. Examples include experimental studies of imaging of concrete blocks [20–22]. Muon scattering tomography has been also used for imaging Fukushima Daiichi reactors [23] or Fukushima reactors [24]. Muons can be also be used in radiographic imaging [25]. In 2017 observation of cosmic-ray muons led to the discovery of an unknown big chamber in Khufu’s Pyramid [26].

Previously, we published a novel approach exploiting muon scattering tomography (MST) to detect the presence and location of reinforcement bars [27]. This work has shown that a 100 cm long, singular bar with a diameter of 33.7 ± 7.3 mm can be detected using three weeks of data taking at sea level. It was also shown that the signal has a monotonic dependence on the volume of the bar contained in the concrete drum. Moreover, the volume of the inclusion can be measured with a resolution of 5.4 ± 0.3%, and relative uncertainty below 10%, for bar volumes above 2 500 cm³. This outcome does not depend on the location of the bar. Bars as close as 6 cm apart can be detected as two individual items. However, the separation starts to be observable at a 4 cm distance. The approach also demonstrated to be suitable for imaging purposes, such as depicting bar structures in reinforced concrete. Differentiation between single and double layers of grids with bars diameters of 30 mm was also possible.

Here we present a new method that exploits the periodicity of bar structures, and is able to detect much smaller bar sizes within shorter time of data collection.

For most reinforced concrete structures, bars with diameters between 8 mm and 40 mm are used [28]. The thinnest bars in use are 6 mm in diameter, whereas for walls and bridges much thicker bars (≥10 mm) are used. The yield strength of the concrete depends strongly on the regular placement of the bars. Precise measurements are important for structural re-assessment to define a structural capacity or longevity of a concrete element or building structure. The spacing on most bar products is 10 or 20 cm [29]. Therefore, we have performed our studies with bars of a minimum diameter of 6 mm in a perfect grid of 7.5, 10, 15 and 20 cm.

### 2. Muon scattering tomography (MST)

Muon scattering tomography is a non-invasive method which shows great potential to generate high-depth 3D concrete images. MST uses cosmic rays as probes. Cosmic rays are high-energy charged particles which come to the Earth’s atmosphere from outer space. In the atmosphere, cascades of new particles are produced. The main type of particles that reach sea level are muons. Muons are identical to electrons, but 200 times heavier. Muons can go through large amounts of material as they do not scatter very much due to their high mass. When traversing matter, Coulomb interactions take place between the muons and the nuclei of the material. As a result, muons undergo a series of scattering events and exit the material under an angle. The angular distribution of scattered muons can be described by a Gaussian distribution with a mean of zero and a standard deviation $\sigma_{\theta}$ described by [30]:

\[
\sigma_{\theta} \approx \frac{13.6 \text{ MeV}}{p c \beta} \sqrt{\frac{T}{X_0}} \left[1 + 0.038 \ln \left( \frac{T}{X_0} \right) \right]
\]

\[
X_0 \approx \frac{716.44}{Z (Z + 1) \ln \left( \frac{281}{Z} \right)} \ [\text{g} \cdot \text{cm}^{-2}]
\]

where $p$ is muon’s momentum; $\beta$ is muon’s speed divided by the speed of light $c$; $T$ is the thickness of the material and $X_0$ its radiation length; $A$ is the atomic weight of the medium in g·mol⁻¹. $\sigma_{\theta}$ depends on the atomic number $Z$ of the traversed material. Under the assumption that scattering occurs in a single location, and by reconstructing the incoming and outgoing trajectories of the muons, the scattering angle distribution can be reconstructed and thus information about the traversed material can be inferred.

### 3. MST simulation

In this work, we use Monte Carlo simulations of a MST system. The simulated MST system consists of detectors and a reinforced concrete block. The muons were generated using the CRY library [31], specifically developed for cosmic rays. GEANT4 [32] was used to simulate the propagation of the muons through detectors and scanned objects.

The simulated detector system consists of 3 sets of $2 \times 2$ m² resistive plate chambers (RPCs) on one side of the volume under study and 3 sets of $2 \times 2$ m² of RPCs on the other one. Three layers of detector sets on each
side of the scanned object were chosen to provide a precise reconstruction of the muon direction and thus of the scattering angle. In section 5.2, the effect of adding more layers is discussed.

RPCs consist of a gas volume enclosed between two glass panels over which a high voltage is applied. Muons, which are electrically charged, ionize the gas when traversing the detector. Due to the high voltage, a small avalanche is created locally. The measurable signal induced by the avalanche can be used to reconstruct the location where the muon traversed the RPC. The simulated RPCs performance was modeled using the performance of RPCs that were built for a container scanner prototype [33, 34]. The RPCs had a pitch of 1.5 mm, which resulted in a position resolution of approximately 450 μm, which is the resolution used in the study presented here. The effect of the position resolution on the results is discussed in more detail in section 5.1. One pair of RPCs consists of both X and Y planes, orthogonal to each other, so that both x and y coordinates of the muon paths can be detected, and the muon paths calculated accurately. The thickness of one RPC is 6 mm. The X and Y planes are 19 mm apart and the distance between the RPC pairs is between 56 and 59 mm. The space between top and bottom detector layers, where an object can be scanned is 548 mm. The concrete block was placed in this volume. Reinforced concrete was modeled as a rectangular, concrete-filled object with dimensions of 200 cm × 200 cm × 34 cm. Inside the block, reinforcement bars were arranged in two orthogonal layers to form a single, reinforcement grid. Concrete was modeled as a material with a density of 2.3 g cm⁻³, and the steel reinforcement bars were simulated as iron bars with density of 7.87 g cm⁻³. The density of concrete ranges from 2 to 2.5 g cm⁻³. A schematic drawing of the simulated setup is shown in figure 1.

Multiple scenarios were simulated to perform a detailed analysis. For all simulations the RPCs remained the same, whereas the contents of the concrete were changed. The diameter of the bars varied from 6 mm to 20 mm. Spacings of 7.5, 10, 15 or 20 cm were used. An example with a single layer of reinforcement grid is illustrated in figure 2. Figure 2(a) shows the ZX (front) and ZY (side) projection of the concrete block. The top (YX) projection is shown in figure 2(b).

3.1. Bristol discriminator algorithm

A number of MST imaging approaches have been developed. All of them use different ways to retrieve information from the incoming and outgoing muon tracks [35, 36]. We developed an algorithm which is described in detail in [17]. It was successfully used to detect objects with a high atomic number (high-Z objects) in cargo containers [17, 37], imaging of nuclear waste [12], as well as for discrimination of high-Z materials in concrete-filled containers [13] and detection of voids in concrete filled drums [11]. It was also demonstrated that bars with a diameter of 33.7 ± 7.3 mm can be located using that approach [27].

In our method incoming tracks are reconstructed using the three detector planes above and outgoing tracks using the three detector planes below the block. Subsequently, the hits are refitted under the assumption that the incoming and outgoing track meet in a vertex, where the scattering is assumed to have taken place in a single point. In reality this is not strictly true as the muon actually performs a random walk through the concrete block. However, the vertex assumption turns out to be a very useful one in practice. Our method relies on the
The clusteredness of high angle scatters in higher-Z materials: in high-Z materials the scattering angles tend to be larger, and larger scattering angles result in a well-defined vertex. In addition, a higher number of large-angle scatters occur in higher-Z materials. This makes the method very sensitive to the detection of materials with a higher-Z inside an object of a lower-Z, or the other way around.

An example of incoming, outgoing tracks and a vertex is shown in figure 3. The scanned object is subdivided in voxels. A voxel size of 10 mm × 10 mm × 10 mm was used in this study. Each track-vertex-track combination is assigned to the voxel where the vertex is reconstructed. Since the method exploits the clusteredness of high angle scatters, only the $N$ most scattered tracks assigned to each voxel are considered in further analysis. $N$ of 30 was used for this analysis. For each pair of remaining vertices in that voxel, the weighted metric, $\tilde{m}_{ij}$, is calculated:

$$\tilde{m}_{ij} = \frac{\|V_i - V_j\|}{\theta_i \cdot \theta_j}$$

where $V_i$ is the position of the vertex of muon $i$; $\theta_i$ is the corresponding scattering angle. The median of the weighted metric distribution is calculated for every voxel. The median of that distribution is then used as a discriminator [17]. An example of the median discriminator distribution is shown in figure 4. In low atomic number (low-Z) materials, vertices are located at larger distances as high-angle scattering occurs less often than

![Figure 2](image-url) Figure 2. A schematic drawing of the concrete block used in the simulations. Inside, two layers of reinforcement bars, forming a single grid were placed. Reproduced from [27] © IOP Publishing Ltd. CC BY 4.0.

![Figure 3](image-url) Figure 3. Incoming and outgoing muon tracks are reconstructed with RPCs. On their basis, the vertex is determined. Reproduced from [17] © 2013 IOP Publishing Ltd and Sissa Medialab srl.
hence, lower-Z materials correspond to higher discriminator values. In Figure 4 the discriminator distributions for a case with concrete only and a concrete block with reinforcement placed inside are shown. The reinforced block results in more low discriminator values.

4. Detection of reinforcement bars

In this section a new detection algorithm will be presented which removes more background. An example of its performance will be shown using a concrete with a single grid made of bars with diameter of 1.5 cm, and a spacing of 15 cm. A background scenario was defined as pure concrete block. This method is an improved version of the algorithm published in [27]. In the previous algorithm, for both background and reinforcement scenarios, for each voxel the discriminator is calculated. Next, a number of sub-volumes is created for each scenario and for each sub-volume, a discriminator distribution prepared. Then, for each bin of a discriminator distribution, the absolute difference between the discriminator values of the block containing concrete and the scenario containing bars are taken. The resulting values are summed along the x, y and z-axis. The same study is repeated for every sub-volume, resulting in three projections. A detailed description of the approach is in [27]. An example of the front projection image (ZX) is shown in Figure 5(a).

For the new method, all the above mentioned steps are done and on top of them the background subtraction was improved by generating two more concrete samples (the same size as bar-scenario ones) and performing the same analysis using the two concrete samples and thus generating final projection images for background only sample. Then, the background projections were subtracted, bin-by-bin, from the bar-concrete scenario projections. This method of background subtraction was chosen as the background is not linear and thus resulted in clear differences between bar and concrete. An example of the projection image before and after additional background elimination is shown in Figure 5. Areas with higher signal clearly indicate the presence of iron, while areas with lower signal show where the concrete is. The bars are clearly observable.

This technique allows to detect thick rebars but does not work for thin rebars of a few mm diameters. Here a novel algorithm is presented to detect these thin rebar grids.

For simplicity, the method is presented for single layers of rebar only.
4.1. Detection of the reinforcement grid

From a structural engineering perspective, the reinforcement is always fixed in an orthogonal grid pattern. Due to its regular shape, the reinforcement provides a periodic signal. This periodicity can be exploited to lower the detection limit of the minimum diameter of the bars. This is done by calculating the auto-correlation of the 2D signal distributions (like figure 5(b)), and looking for peaks in the Fourier spectrum of the auto-correlation. The occurrence of peaks indicates the presence of a periodic structure and thus of bar. The peak locations yield the spacing, while the peak amplitude depend on the diameter of the bar.

In this method, a variation of the standard auto-correlation is used. Since there is periodicity in both the x and y direction, the auto-correlation is only evaluated in one direction, i.e. the x direction, as:

\[
R_f(x, y) = \int_{x_{\min}}^{x_{\max}} \int_{y_{\min}}^{y_{\max}} f(x', y') f(x' + \tau, y') \, dx' \, dy'
\]

(4)

The calculation was limited to the area within the acceptance of the detector, and the presence of the reinforcement i.e., \(X < 20;175\), \(Y < 10;29\) in figure 5(b).

The result of the auto-correlation for the example case is shown in figure 6. The triangular shape is due to the variation of the overlapping area. It is observed as a triangular background and the triangular dependence of the amplitude of the periodic structure. The periodic structure is due to the reinforcement spacing. Before the Fourier transformation, that triangular background needs to be subtracted. To estimate it, the complete series of auto-correlation, \(R_{b,k}\) is calculated, where:

\[
R_{b,k} = \int_{x_{\min}}^{x_{\max}} \int_{y_{\min}}^{y_{\max}} f(x', y') f(x' + \Delta_k + \tau, y') \, dx' \, dy'
\]

(5)

Here the function is shifted by an additional \(\Delta_k\), where \(k\) indicates the number of pixels the image has been shifted. The shift occurs in a rolling mode, i.e. when a column is shifted further than \(x_{\max}\), it is placed in column 0. This is illustrated in figure 7. This procedure leads to a shift in the peaks, but the underlying triangular
background shape remains the same, as can be seen in figure 8. For each bin in the signal auto-correlation, the minimum value of $R_{b,k}$ is subtracted. The result is shown in figure 9. The graph still displays the triangular pattern in the amplitude, but the triangular background under the function is removed.

Next, the background-subtracted signal is Fourier transformed using an interface class for Fast Fourier Transforms, TVirtualFFT, provided by ROOT, a data analysis package developed at CERN \[38\]. Figure 10 shows the result of the FFT. The spectrum shows a series of clearly visible distinct peaks. Their position is determined by the bar spacing and the peak amplitudes are related to the bar diameter. A background scenario, consisting of a concrete block without reinforcement, does not display any peaks. These results show that the method is capable of detecting bar grids.

4.2. Variation of the bar diameter
The peak locations of the Fourier transform depend on the spacing, while the peak amplitude is related to the bar diameter. Figure 11 shows the Fourier transforms for reinforcement grid made of 20, 15, 10 and 8 mm diameter
bars with a 15 cm spacing, see figure 12 for a zoomed version of the plot. As expected, having the same spacing, the peaks occur always at the same normalized frequency values. With decreasing bar diameter, the amplitude of the peaks also decreases. This is presented more clearly in figure 13, which shows amplitude of the peak at 0.07 of the normalized frequency plot as a function of bar diameter. The amplitude strongly increases with increasing the diameter. Hence, the bar diameter can be measured based on normalized frequency spectrum.
4.3. Variation of the spacing
In the previous results, a spacing of 15 cm was used. Figure 14 shows the Fourier spectrum for the reinforcement using 8 mm diameter bars with spacing of 10, 15 and 20 cm, see figure 15 for a zoomed version of that figure. All of the cases are clearly distinguishable from the background scenario. Moreover, the peaks are located at different frequencies, which makes this method suitable for the estimation of the spacing.

4.4. Limits of the method
The smallest bars in normal use have a 6 mm diameter. In order to estimate whether the method is capable of finding such a small bar, the Fourier transform of grids with 8, 7 and 6 mm diameter bars were calculated using a 10 cm spacing and a one week worth of data taking. The results are shown in figure 16 and a zoomed version in figure 17. Bars with 7 and 8 mm diameter at normalized frequency values of 0.1 and 0.9 are clearly distinguishable from the background. However, the signal for the 6 mm diameter case is less clear. Please note that the peak locations are determined by the spacing and thus only peaks at the right location need to be considered. To strengthen the 6 mm diameter signal, two weeks worth data taking were used. The peaks become clearer after two weeks of data taking, see figure 18 and a zoomed version in figure 19. Figure 20 shows amplitude of the peak at 0.1 of the normalized frequency spectrum. Clearly, the method can detect the smallest size bar in use in practice for a 10 cm spacing. Amplitude of the bar with 6 mm diameter is almost 6 times higher than a background sample.

To evaluate the limits of the method with respect to the spacing, a spacing of 20 cm was also considered. However, bigger spacing between bars results in a smaller amount of steel in the scanning area. This results in a
smaller signal amplitude, as can be seen in figure 21, zoomed version in figure 22. The signal for 6 mm diameter bars is now almost below background level. However, when increasing the data taking time to two weeks (figure 23, zoomed version in figure 24), even the smallest bar is clearly visible again.

Figure 16. Normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 10 cm.

Figure 17. Zoomed, normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 10 cm.

Figure 18. Normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 10 cm. Time of data taking was increased to two weeks.
Figure 19. Zoomed, normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 10 cm. Time of data taking was increased to two weeks.

Figure 20. Amplitude of the peak at 0.1 of the normalized frequency distribution for different size of the bar diameter and spacing fixed at 10 cm. Amplitude of the bar with smallest commonly used diameters is almost 6 times higher than a background sample.

Figure 21. Normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 20 cm.
Figure 22. Zoomed, normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 20 cm.

Figure 23. Normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 20 cm. Time of data taking was increased to two weeks.

Figure 24. Zoomed, normalized frequency spectrum for a reinforcement made of 6, 7 or 8 mm diameter bars and spacing of 20 cm. Time of data taking was increased to two weeks.
5. Detector choices

Our simulation study is undertaken with a 3 layer system with a hit resolution of 450 \( \mu \text{m} \). Here the effects of the hit resolution and the number of planes will be discussed.

5.1. Effect of detector resolution

We have built and tested a large number of RPCs, see [33, 34]. Their spatial resolution was found to be lower than 450 \( \mu \text{m} \) and therefore this was the resolution chosen for our studies. Figure 25, zoomed version in figure 26, shows the normalised frequency spectrum for the 8 mm diameter rebar for position resolutions of 450 \( \mu \text{m} \), 1 mm and 1.5 mm. When the resolution is worse, the tracks are reconstructed with less precision. This results in a more smeared out initial image, an example of which is shown in figure 5. A more smeared out image leads to less pronounced peaks in the Fourier spectrum, as can be seen in figures 25 and 26. The result also shows that the 8 mm diameter rebar can still be observed clearly with 1.5 mm position resolution.

5.2. Effect of the number of layers

Our system consists of three layers per station. This is the minimum to perform a track fit. Tracks fits ensure that the right hits are combined and suppress the combinatorial background, which can come from hits due to environmental radiation. Adding more layers will improve the precision on the reconstructed incoming and outgoing track, but it will also lead to higher cost of the system and reduces the angular acceptance. When traversing the object under study, the muons undergo multiple scattering. This is a series of random scattering
events. Hence, the assumption that the muon enters the detector and object under study in a straight line, scatters and exists again in a straight line is not correct. However, it is a very useful assumption and used in many algorithms as it allows to calculate an effective scattering angle. Hence, the incoming and outgoing tracks give a good indication of where the muon went and how much it was scattered but fundamentally the tracks do not describe exactly what happens in reality and thus the improvement in the tracking precision does not necessarily result in better results. Figure 27, zoomed version in figure 28, shows the normalised frequency spectrum for 8 mm thick rebar with 10 cm spacing for a 3, 4 and 5 layer system using only tracks that resulted in a hit in each plane. The graphs show that the peak in the Fourier spectrum for 4 layers has the same width and lower amplitude. This is because there are many muons that fall in the angular acceptance of a 3 layer system but out of the angular acceptance for a 4 layer system. This is even worse for the 5 layer system. The improved tracking precision does not compensate for the reduction in the number of tracks. Of course, for a 4 layer system, tracks that only go through 3 planes can be included, but give that the width of the 4 layer peak is the same, the improvement is marginal and does not compensate for the additional cost of adding another layer.

6. Summary

Inspection of ageing, reinforced concrete structures is a world-wide challenge and needs novel non-destructive evaluation techniques with large penetration depths to precisely ascertain the configuration of reinforcement and the internal condition of the structure and steelwork, which can possibly contain some impurities like voids. Muon scattering tomography offers a technique that suits those needs. A method was presented to locate reinforcement placed in a large-scale concrete object. The reinforcement was simulated as two layers of 2 m long
bars, forming a grid, placed at a fixed distance from each other inside a large concrete block. The technique exploits the periodicity of the bars in a reinforcement grid by considering the Fourier-transformed signal. The presence of a grid leads to peaks in the normalized Fourier frequency spectrum. Peaks locations are determined by the grid spacing and their amplitude by the bar diameters. It is therefore possible to estimate both bar presence of a grid leads to peaks in the normalized Fourier frequency spectrum. Peaks locations are determined by the grid spacing and their amplitude by the bar diameters. It is therefore possible to estimate both bar

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