

# Using machine learning to magnetically fingerprint particulate matter

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## Background

High-levels of urban particulate pollution are known to affect human health. European legislature states that the levels of particulate matter, so called PM<sub>10</sub> (particles  $< 10 \mu\text{m}$  in size), should not exceed 50  $\mu\text{g}/\text{m}^3$  on more than 35 days per year in a given city – unless their origin can be shown to be natural, e.g., volcanic ash. To try to meet these European requirements and to minimize high-levels of PM<sub>10</sub>, it is important to understand the origin of PM<sub>10</sub>, its spatial distribution and transport mechanisms. However, current provenance analysis of PM<sub>10</sub> data, that is, determining the origin or source of the PM<sub>10</sub> is limited.

This PhD addresses this challenge by taking advantage of two recent break-throughs, the first in knowledge, the second in method development. First, Muxwworthy et al. (2022) showed that much of the grain-size distribution in urban PM<sub>10</sub> from London is  $< 50 \text{ nm}$ ; most previous studies did not apply methods capable of quantifying this fraction of the grain-size distribution, meaning that its contribution was (and still is) poorly understood (Fig. 1). Why does this fraction of the grain-size distribution matter? The finest particles are both suspended in the air the longest and have the highest surface to volume ratio, which increases their toxicity. There is an urgent need to better quantify this fraction. Second, Pei et al. (2025) developed a machine-learning based method – FORCINN – of inverting for inverting a type of magnetic-hysteresis measurement called first-order reversal curve (FORC) measurement for grain-size distribution (Fig. 2). FORCINN is the first inversion method for this type of magnetic hysteresis data, and is particularly sensitive to grains  $< 300 \text{ nm}$  in size, as magnetic particles display strong variations in magnetic behaviour over this grain-size fraction. Pei et al. (2025) tested this method of volcanic rocks (Fig. 2), however, there is no reason why this method cannot be applied to other magnetic systems natural or not.

The PhD project will extend FORCINN to include a greater range of naturally occurring magnetic minerals and their morphologies, and apply the updated version of FORCINN to particulate matter samples from London; the samples are collected by the School of Public Health at Imperial ([www.londonair.org.uk](http://www.londonair.org.uk)). The aim of the study is to develop a full provenance model for these samples.

**Student Profile:** This project is a combined machine-learning and laboratory project and would suit a candidate with an interest in environmental science. Candidates should have a good degree in any area of science, plus strong numerical skills and an interest in machine-learning approaches.

## References

Muxwworthy, A.R., Lam, C., Green, D., Cowan, A., Maher, B.A. & Gonet, T., 2022. Magnetic characterisation of London's airborne nanoparticulate matter, *Atmos. Environ.*, 287, 119292, doi:[j.atmosenv.2022.119292](https://doi.org/10.1016/j.atmosenv.2022.119292).

Pei, Z., Williams, W., Nagy, L., Paterson, G.A., Moreno, R., Muxwworthy, A.R. & Chang, L., 2025. FORCINN: First-Order Reversal Curve Inversion of Magnetite Using Neural Networks, *Geophys. Res. Lett.*, 52, e2024GL112769, doi:[10.1029/2024GL112769](https://doi.org/10.1029/2024GL112769).

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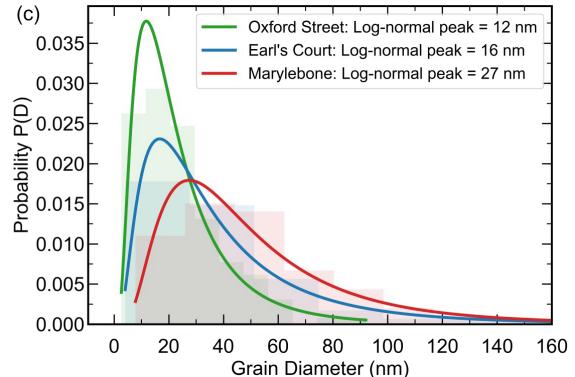


Figure 1. Characterization of grain-size distributions of particulate matter (mostly iron oxides) from various localities in London. From Muxwworthy et al. (2022).

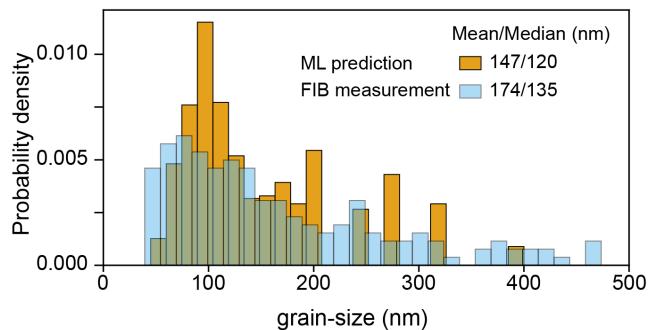


Figure 4. Predicted (ML) and measured (FIB) grain-size distribution for a basalt sample from Vesuvius (Pei et al., 2025). The ML prediction was determined using bulk-sample FORC data.