

1 Guest Editorial

2 **Special Issue: Proximal Soil Sensing – Sensing Soil Condition and**
3 **Functions**

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5 Abdul M. Mouazen^{1,2*}, Zhou Shi³, Marc Van Meirvenne²

6 ¹Cranfield Soil and AgriFood Institute, Cranfield University, Cranfield MK43 0AL,
7 Bedfordshire, United Kingdom

8 ²Department of Soil Management, Ghent University, Coupure 653, 9000 Gent,
9 Belgium

10 ³College of Environmental and Resource Sciences, Zhejiang University, Hangzhou
11 310058, China

12 E-mail of corresponding author: Abdul.Mouazen@UGent.be

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14 Proximal soil sensing (PSS) is defined by McBratney et al. (2011) as the use of
15 ground-based sensors for: 1) *in situ* and mobile field measurements, which is a narrow
16 sense definition, and 2) *in situ* and *ex situ*, mobile and stationary measurements,
17 which is a wide sense definition. Proximal soil sensing has gained broad attention
18 from scientists across different disciplines in the last decade. This led to the
19 establishment of the International Union of Soil Science (IUSS) Working Group on
20 PSS in 2008. Since then four Global Workshops on PSS have taken place,
21 successively in Sydney, Australia (2008), Montreal, Canada (2011), Potsdam,
22 Germany (2013) and Hangzhou, China (2015).

23 The 4th Global Workshop (GW PSS2015) was organised by Zhejiang University and
24 held in 12-15 May, 2015 in Hangzhou, China. The theme of this workshop was
25 “*Sensing soil condition and functions*”, which was chosen to reflect advances and
26 new applications of PSS in soil science, environment, archaeology, land use, and
27 precision agriculture. It provided a forum for researchers, professionals and engineers
28 from all over the world to present their latest research and development results, and
29 exchange and share information and experiences in the fields of PSS, such as *in situ*
30 measurement of soil properties, proximal sensing of soil carbon and biota, sensor data
31 processing and fusion, sensor-based digital soil mapping, combining proximal and
32 remote sensing, development of multi-sensor platforms, advances in soil
33 electromagnetic technologies, and new applications of PSS. The authors working on
34 different topics within PSS, including more than 140 members participated in the 4th
35 Global Workshop, were invited to submit high quality papers to this Special Issue. It
36 is worth noting that another Special Issue on PSS was published in *Geoderma*
37 (Adamchuk and Viscarra Rossel, 2013), based on a collection of papers presented at
38 the 3rd Global Workshop on PSS in Potsdam, Germany (2013).

39 We received 30 submissions to this Special Issue; out of which 13 papers (43%) were
40 finally accepted (and one paper was transferred to a regular issue in *Biosystems*
41 *Engineering* due to an extended revision process). These papers are a mix of
42 laboratory and field PSS methods, with the majority concerning visible and near
43 infrared (vis-NIR) spectroscopy applications (seven contributions). Four papers
44 addressed electromagnetic induction (Jiang et al., 2017), mid infrared spectroscopy (Ji

45 et al., 2016), ground penetration radar (Cavallo et al., 2016) and microscope-based
46 computer vision to characterise soil texture and organic matter (Sudarsan et al., 2016)
47 and the remaining two papers (Rosero-Vlasova et al., 2016; Cho et al., 2016) can be
48 classified as multi-sensor and data fusion approaches.

49 The seven vis-NIR papers covered measurement of cation exchange capacity (Ulusoy
50 et al., 2016), prediction of total dissolved salts and soluble ion concentrations (Peng et
51 al., 2016), prediction of selected soil properties with machine learning methods
52 (Morellos et al., 2016), classification of soil classes (Zeng et al., 2016), improving
53 spatial estimation of soil organic matter (Xie and Li, 2016), evaluation of possible
54 prediction of subsurface salinity from surface spectra (Liu et al., 2016) and estimation
55 of wet aggregate indices (Waruru et al., 2016). This demonstrates the importance and
56 superiority of vis-NIR spectroscopy over other PSS techniques for both field and
57 laboratory measurement conditions. However, another interesting point to note is the
58 increasing number of applications adopting multi-sensor and data fusion approach in
59 PSS, with two papers published in this Special Issue on combining vis-NIR with short
60 wave infrared (SWIR) for characterisation of soils from wildfire burns
61 (Rosero-Vlasova et al., 2016) and soil strength with apparent electrical conductivity to
62 estimate physical soil properties (Cho et al., 2016). This can be attributed to the
63 complex nature of soil, which necessitates accounting for more than a single sensor to
64 measure a key soil property (Kuang et al., 2012). This is in-line with conclusions
65 made in a previous Special Issue published on reflectance and fluorescence
66 spectroscopy in soil science (Mouazen et al., 2016), suggesting the need for future

67 work to focus on adoption of the multi-sensor and data fusion approach, including in
68 particular vis-NIR spectroscopy.

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76 Sincerely Yours,

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78 *Abdul Mouazen, Managing Guest Editor*

79 *Zhou Shi, Guest Editor*

80 *Marc Van Meirvenne, Guest Editor*

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