



- 1 **Title:** Development of a harmonized soil profile analytical database for Europe: A
- 2 resource for supporting regional soil management
- 3 Authors:
- 4 Jeppe Aagaard Kristensen<sup>1,2\*‡</sup>, Thomas Balstrøm<sup>2</sup>, Robert J.A. Jones<sup>3</sup>, Arwyn Jones<sup>4</sup>, Luca
- 5 Montanarella<sup>4</sup>, Panos Panagos<sup>4</sup>, and Henrik Breuning-Madsen<sup>2†‡</sup>.

## 6 Affiliations:

- <sup>7</sup> <sup>1</sup>Department of Physical Geography and Ecosystem Science, Lund University, Sölvegatan 12,
- 8 223 62 Lund, Sweden.
- 9 <sup>2</sup>Department of Geosciences and Natural Resource Management, University of Copenhagen,
- 10 1350 Copenhagen K, Denmark.
- <sup>3</sup>School of Energy, Environment and AgriFood, Cranfield University, College Road, Cranfield,
- 12 MK43 0AL, UK.
- <sup>4</sup>European Commission, DG Joint Research Centre, Via E. Fermi 2749, 21027 Ispra (VA), Italy.
- 14 \**Correspondence to jeppe.aa.kristensen@gmail.com*
- 15  $^{\dagger}Deceased$
- 16 <sup>*‡*</sup>*These authors contributed equally to this work.*
- 17 Running head: A harmonized soil profile analytical database for Europe.





# 18 Abstract

19	Soil mapping is an essential method to obtain a spatial overview of soil resources that are
20	increasingly threatened by environmental change and population pressure. Despite recent
21	advances in digital soil mapping techniques based on inference, such methods are still immature
22	for large-scale soil mapping. During the 1970s, 80s and 90s, soil scientists constructed a
23	harmonised soil map of Europe (1:1M) based on national soil maps. Despite this extraordinary
24	regional overview of the spatial distribution of European soil types, crude assumptions about soil
25	properties were necessary to translate the maps into thematic maps relevant for management. To
26	support modellers with analytical data connected to the soil map, the European Soil Bureau
27	commissioned the development of the Soil Profile Analytical Database for Europe (SPADE) in
28	the late 1980s. This database contains soil analytical data based on a standardised set of soil
29	analytical methods across the European countries. Here, we review the principles adopted for
30	developing the SPADE database during the past five decades, and the work towards fulfilling the
31	milestones of full geographic coverage for dominant soils in all the European countries (SPADE
32	level 1), and the addition of secondary soil types (SPADE level 2). We demonstrate the
33	application of the database by showing the distribution of the root zone capacity, and by
34	estimating the soil organic carbon (SOC) stocks to a depth of 1 m for EU-27 to 76 x $10^{15}$ g. The
35	increased accuracy, potentially obtained by including secondary soil types (level 2), is
36	demonstrated in a case study of estimating SOC stocks in Denmark. In the lack of systematic
37	cross-European soil analysis schemes, integrating national soil maps and locally assessed
38	analytical data into a harmonised database is a powerful resource to support soil resources
39	management at regional and continental scales by providing a platform to guide sustainable soil
40	management and food production.





- 41
- 42 Keywords: EU soil map; SPADE; Soil data harmonisation; Soil organic carbon; Root zone
  43 capacity
- 44

# 45 Introduction

46	In a world subject to constant environmental change and increasing population pressure, soil
47	becomes an increasingly important but threatened resource (FAO 2015; Sustainable Food Trust
48	2015). This challenge must be met at multiple levels and scales; hence, accurate understanding of
49	the available resources at the appropriate scale is required (e.g. Robinson et al. 2017). In spite of
50	advances in digital soil mapping using remote sensing and geographical information systems to
51	infer soil properties (McBratney et al. 2003; Arrouays et al. 2014; Minasny and McBratney
52	2016; Zhang et al. 2017), we still lack adequate standardised methods for large scale soil
53	mapping. Furthermore, the existing methods are particularly challenged in densely vegetated
54	areas and for subsoil properties (Mulder et al. 2011), which are highly relevant for environmental
55	management and food production. A recent assessment of the implications of uncertainty in soil
56	data found that it could potentially offset climate change impact on future crop yields, due to the
57	dependence on soil type (Folberth et al. 2016).
58	During the last century, national soil maps were established in most European countries but they
59	were not harmonised across borders as they were based on specific national soil classification
60	systems (Morvan et al. 2008). Therefore, international soil classification systems were

- 60 systems (Morvan et al., 2008). Therefore, international soil classification systems were
- 61 developed during the 1960s and early 1970s to facilitate the construction of globally standardised





62	soil maps (FAO-Unesco 1974, SMSS/USDA/AID 1983). The FAO maps portrayed the soils
63	resources for each individual country as mapping units with a distinct set of soil types. The soil
64	types comprised three categories: dominant soils, associated soils, and inclusions. The dominant
65	soil type covered the largest proportion of the mapping unit; associated soils occupied 20% to
66	50% of the unit while the inclusions accounted for less than 20%. The maps were published with
67	an explanatory text describing the geology, geomorphology, land use and a map showing the
68	level of knowledge behind the map construction, i.e. the level of confidence (King et al. 1994).
69	In the beginning of the 1980s, the ten European Communities (EC) Member States elaborated
70	the FAO-Unesco approach to make an expanded and a more detailed version of the FAO-Unesco
71	(1974) system for the soil types present in their respective countries. Based on this, the EC
72	published seven soil maps (scale 1:1M, Commission of the European Communities, 1985). The
73	complete soil map of Europe was digitised by the end of the 1980s (Platou et al. 1989) as a part
74	of the EC financed CORINE programme (Briggs & Martin 1988), and quickly, it became an
75	important dataset in the forecasting of national crop yields across Europe by the European
76	Commission's Joint Research Centre's Monitoring Agriculture of Remote Sensing (MARS)
77	project (Vossen 1993). Subsequently, the EC soil map was used widely to underpin soil resource
78	assessments within the European Union (EU) including the mapping of carbon (C) stocks
79	(European Commission, 2005; Jones et al., 2005; Lugato et al., 2014), soil erosion risks (Kirkby
80	et al. 2008, Panagos et al. 2015), vulnerability to compaction (Jones et al. 2003, Schønning et al.
81	2015) and salinity (European Commission, 2005), as well as raising awareness and providing
82	education materials (e.g. European Commission, 2005).
83	Yet, such assessments are based on assumptions about each soil type's characteristics or

84 extrapolations from limited amounts of (often) country specific analytical data. Therefore,





85	incorporating national datasets into one uniform European database would dramatically increase
86	the quality of predictions and evaluations based on the EC Soil Map across Member State
87	borders. A global attempt to meet this challenge has led to the development of the Harmonized
88	World Soil Database (FAO/IIASA/ISRIC/ISSCAS/JRC 2012), but this database extracts its data
89	from Europe from the European Soil Database (v.2.0), which in turn is based on the soil profile
90	analytical database for Europe (SPADE). This paper demonstrates how this cornerstone in the
91	European Soil Data Centre (Panagos et al. 2012) was created based on soil physical and chemical
92	soil data provided by national stakeholders from each member state. Specifically, a database
93	containing estimated analytical data for all dominant soil types within the EU with full
94	geographical coverage (SPADE 14) was compiled. Furthermore, a level 2 database was
95	developed for a subset of countries, and a full coverage level 2 database (SPADE 18), will in the
96	years to come be expanded to cover the entire EU and surrounding countries. Finally, it is
97	demonstrated how the database can be used to obtain estimates of environmentally relevant soil
98	properties (e.g. root zone capacity and SOC-stocks).
99	
100	Establishing the Soil Profile Analytical Database of Europe framework (SPADE 1)

101 A working group of Europe-wide soil specialists was formed to advise the Commission of the

102 European Communities on the establishment of a soil profile analytical database (SPADE)

103 connected to the EC soil map (Figure 1a). By the end of the 1980s, the Working Group proposed

that it should be based on four levels of analytical data (Breuning-Madsen 1989): Level 1 would

105 provide analytical data from a typical soil profile for the dominant soil typological unit (STU) in

106 each soil mapping unit (SMU), preferably on arable land; Level 2 would expand the database to

107 include a typical dataset for all STUs, including associated soils and inclusions; Level 3 would





112

108	be a further expansion to include soil analytical data for all soil types with a differentiation
109	between land uses; Level 4 would allow different soil analytical data for the same soil type
110	(STU) that occurs in different sub-regions (e.g. based on geology or geomorphology). (See
111	Figure 1b for a timeline).

Initially, two soil analytical databases were established; one containing estimated mean values for typical soil profiles according to fixed soil analytical procedures provided by national 113 stakeholders (referred to as Proforma I), while another contains soil profile data measured using 114 established analytical procedures (referred to as Proforma II). The Proforma I database contains 115 116 data comparable across country borders while this is not always the case for the Proforma II database. In order to make the database functional as soon as possible for the entire coverage 117 area, each Member State stakeholder was asked to deliver one full set of Proforma I (estimated) 118 119 analytical data for each dominant soil type (STU) in each of the soil mapping units (SMU) 120 delineated on the Soil Map of Europe (1:1M). Providing data for the Proforma II (measured) 121 database was optional. Where possible, the data should be provided for agricultural land, as the 122 primary aim of the database was to underpin large-scale assessments of agricultural land 123 management. In 1993, Proforma I and II schemes (including guidelines) were sent to the stakeholders in order 124

to collect data for the individual countries; detailed guidelines for compilation of the SPADE 1 125 dataset was published by Breuning-Madsen and Jones (1995). 126

127 Subsequently, the SPADE 1 database was expanded to include data from the new EU Member 128 States but also from non-EU nations such as Albania, Norway and Switzerland. By the end of the

129 1990s, SPADE 1 was subject to a data quality assessment and scrutinised to identify missing data





- and evaluate overall data reliability. Based on the recommendations presented at a European Soil
- 131 Bureau Network (ESBN) meeting in Vienna 1999, the national stakeholders were requested to
- update their individual datasets. Meanwhile, only a few national stakeholders responded, which
- 133 left the SPADE 1 incomplete and not well suited for modelling at the European level.
- 134

#### 135 An attempt to populate SPADE with measured data (SPADE 2)

136 Due to the limitations of SPADE-1, SPADE-2 was developed to derive appropriate soil profile

137 data to support, for example higher tier modelling of pesticide fate at the European level (Hollis

138 et al., 2006). Data were supplied from national data archives, similar to SPADE 1 Proforma II.

139 Despite the analytical methods differing between countries, the raw national data were

harmonised and validated to provide a single data file for use in conjunction with the existing

141 Soil Geographical Data Base of Europe (Platou et al. 1989). The primary soil properties required

142 for each soil were: Horizon nomenclature (e.g. A, E, B, C), upper and lower depth (cm), particle-

size distribution: clay, silt, total sand and content of at least 3 sand fractions, pH in water (1:2.5

soil:water), organic carbon content (%) and dry bulk density ( $g \text{ cm}^{-3}$ ).

145 The acquisition of data happened in two steps; first datasets were obtained from Belgium,

146 Luxembourg, Denmark, England and Wales, Finland, Germany, Italy, the Netherlands, Portugal

147 and Scotland (Hollis et al. 2006), and next the database was expanded with data from Bulgaria,

- 148 Estonia, France, Hungary, Ireland, Romania, Slovakia, Spain, France and Ireland . The final
- 149 database (SPADE2v11) only exists as a beta version of collated datasets from the first and
- second phases of soil profile data acquisition (Hannam et al. 2009). However, it was used to
- estimate bulk densities for missing data in the later SPADE 14.





153	Steps tow	vards full geographical coverage (SPADE 8)
154	In an effo	rt to obtain a functional database with full spatial coverage for the EU, a small
155	specialist	group from Denmark (Prof. Henrik Breuning-Madsen, Assoc.Prof. Thomas Balstrøm
156	and MSc.	Mads Koue from the Institute of Geography, University of Copenhagen) undertook a
157	scrutiny o	f the national datasets in 2008 using error finding equations based on literature values,
158	expert jud	Igements, and pedotransfer functions (Koue et al. 2008).
159	First, a qu	ality check was conducted on all data. This process consisted of:
160	i)	cross-checking of interdependent variables (e.g. pH vs. base saturation or porosity vs.
161		saturated water content); and
162	ii)	checking the plausibility of all values according to published theoretical or empirical
163		values (e.g. for bulk density (BS) or C:N-values).
164	Examples	of common questionable data were occurrences of bulk soil C:N values <5,
165	mismatch	es between BS and pH (e.g. BS>90% at pH<4.5), and volumetric water content at
166	saturation	exceeding the porosity. Moreover, in several cases the sum of clay, silt and sand
167	fractions	differed from 100 %. Based on this examination, implausible values were either
168	adjusted t	o plausible values or marked as unlikely based on predefined criteria. In terms of
169	spatial ex	tent, it was only possible to link a soil analytical dataset for a dominant soil type to
170	approxim	ately70% of the soil mapping units (SMU) in the area covered by the database, due to
171	missing d	ata.





- 172 Following an ESBN meeting in Paris, December 2008, the reviewed SPADE 8 database was
- 173 discussed and the national evaluation reports, together with the country specific databases, were
- sent to the national stakeholders with a request to i) review and change the existing data to
- 175 plausible values based on the expert scrutiny, and ii) estimate new datasets for the dominant soil
- types without data based on their local expertise. The modifications received from the
- stakeholders were incorporated in the SPADE 8 database that was renamed SPADE 11.
- 178 However, once again only few responses were received, which still left the database incomplete,
- so SPADE 11 remained as unpublished work in progress.

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#### 183 Establishing a SPADE for dominant soil types with full coverage of the EU (SPADE 14)

- 184 Without further input from the national stakeholders, implausible data identified in SPADE 8
- 185 were estimated to make the Proforma I (level 1) database more functional for modelling. Thus, in
- 186 2014 and early 2015, the SPADE 8 database was updated by a working group consisting of the
- 187 current paper's authors.
- 188 Specifically, this work package had three key goals:
- i: To implement the suggested improvements of the existing data in the SPADE databasesuggested during the 2008 evaluation,





- ii: To estimate values for the profiles lacking data (approximately 32% of the dominantSTUs) based on matching of similar soil types in neighbouring countries, the data in
- 193 SPADE 2, or other reference data sources.
- iii: To update the existing SPADE database with the complete dataset after revision bythe national stakeholders.
- 196 The final SPADE14 database is publically available through JRC's European Soil Data Centre
- 197 (ESDAC) website (<u>http://esdac.ec.europa.eu/</u>).

Firstly, the questionable values identified in SPADE 8, but not corrected by stakeholders due to 198 199 passivity, were adjusted to fit theoretical or average values according to predefined equations or 200 guidelines (Breuning-Madsen et al. 2015). Secondly, data for profiles lacking stakeholder 201 estimated values were assigned by copying complete datasets from identical soil types in 202 neighbouring countries. If no matching profiles were identified, the search was extended to the entire database. Thirdly, data for the remaining  $\sim 15\%$  of the dominant soil types (STUs for 203 204 which no estimated data existed anywhere in the database) was created by adjusting existing data 205 from similar soil profiles, preferably from the country itself or neighbouring countries to 206 minimise confounding factors. The evaluation guidelines sent to the stakeholders during the SPADE 14 evaluation (Breuning-Madsen et al. 2015) provided a detailed description of the 207 208 methodology. The entire database was quality controlled with the updated versions of equations and guidelines used during the 2008 evaluation thus ensuring consistency across Member States. 209 210 Finally, the quality controlled national data where sent to each stakeholder for final checking and 211 revision before publication.

212 Examples of correction guidelines





- For some parameters, no correction guidelines were specified during the 2008 evaluation, in
- which case they were developed during the 2014/15-evaluation. As an example, the estimation of
- bulk density and volumetric water content are elaborated below.
- 216 Bulk density
- 217 Missing bulk density (BD) values were assigned the average of all measured values from the
- 218 SPADE 2 (Table 1). For soil horizons with organic matter (OM) content >10%, BD values were
- calculated from the OM content grouped into 10% intervals. For soils with OM contents <5%,
- BD values were averaged over depth intervals of 25 cm down to 100 cm. Deeper horizons were
- assigned a value of 1.5 g cm<sup>-3</sup> unless geomorphology or overlying horizons indicated a
- significantly different value. For soils with OM contents between 5 and 10%, the BD was
- estimated a value in the range  $1.1-1.2 \text{ g cm}^{-3}$  based on surrounding horizons and profiles.

- 225
- 226
- 227 Volumetric water content (VWC)
- 228 National stakeholders were requested to specify the water content at 1, 10, 100 and 1500 kPa
- suction for each soil horizon enabling the calculation of functions such as root zone capacities. In
- 230 order to assign realistic data to missing estimates, we regressed (multivariate linear regression)
- water retention data, i.e. VWC at 1, 10, 100 and 1500 kPa suction, from countries with complete
- 232 datasets against multiple explanatory variables; bulk density (BD), particle size fractions (TEXT,
- 233  $<2 \mu m$ , 2-20  $\mu m$ , 20-50  $\mu m$ , 50-200  $\mu m$ , 200-2000  $\mu m$ ) and organic matter content (OM, %).





- 234 Member States with complete estimated datasets were Belgium, United Kingdom (UK) and
- 235 Denmark. As data from DK were used for validation, the derived equations were based on data
- 236 from Belgium and the UK. Fluvisols were omitted as they often have complicated water
- 237 retention properties due to their geomorphological origin. Only 7 % (9 of 132) of the
- observations from DK deviated more than 10% VWC from the 1:1 line between observed and
- 239 calculated values using the linear models. The adjusted correlation coefficients were 0.85, 0.86,
- 240 0.87 and 0.91 for VWC<sub>1</sub>, VWC<sub>10</sub>, VWC<sub>100</sub>, and VWC<sub>1000</sub>, respectively (P < 0.001), and the
- 241 resulting regression equations were:

242 
$$VWC_1 = (-27.653 \times BD + 1.463 \times OM + 0.208 \times TEXT_2 + 0.017 \times TEXT_{20} + 0.154 \times TEXT_{50} + 0.154$$

- 243  $0.013 \times TEXT_{200} + 0.003 \times TEXT_{2000} + 57.783) \times BD$
- 244  $VWC_{10} = (-20.231 \times BD + 1.110 \times OM + 0.262 \times TEXT_2 + 0.029 \times TEXT_{20} + 0.193 \times TEXT_{50} 0.029 \times TEXT_{20} + 0.0193 \times TEXT_{50} 0.029 \times TEXT_{20} + 0.0193 \times TEXT_{50} 0.000 \times TEXT_{50} + 0$
- 245  $0.026 \times TEXT_{200} 0.072 \times TEXT_{2000} + 41.072) \times BD$
- 246  $VWC_{100} = (-4.246 \times BD + 1.356 \times OM + 0.335 \times TEXT_2 + 0.071 \times TEXT_{20} + 0.105 \times TEXT_{50} 0.000 \times TEXT_{50} + 0.0$
- 247  $0.002 \times TEXT_{200} 0.015 \times TEXT_{2000} + 8.380) \times BD$
- 248  $VWC_{1500} = (-0.330 \times BD + 1.088 \times OM + 0.358 \times TEXT_2 + 0.125 \times TEXT_{20} + 0.072 \times TEXT_{50} + 0.$
- 249  $0.056 \times TEXT_{200} + 0.053 \times TEXT_{2000} 4.719) \times BD$
- 250

#### 251 Traceability and quality check

- 252 In order to ensure traceability of all proposed changes, we developed a colour coding system to
- the Excel spreadsheets submitted to stakeholders that allowed them to identify what kind of
- changes had been applied to each data element. Moreover, a tracing document keep track of





255	whether the dominating STUs contained original stakeholder estimated data, a dataset copied
256	from another profile in the database, or a dataset modified by the working group. For the latter, a
257	separate tracing document kept track of profiles and parameters modified to anticipate criticism
258	and corrections by national stakeholders. Finally, the data quality was evaluated as prior to the
259	modifications, and a new cross-database-check was introduced to make sure whether the topsoil
260	texture class specified in the estimated profile database matched the actual topsoil texture class
261	specified in the estimated horizon database. When inconsistencies were identified, the topsoil
262	texture class in the estimated horizon database was adjusted accordingly.
263	
264	Evaluating, updating and publishing the SPADE-14 database
265	Table 2 provides an overview of the origin of the data for each country. The first column

266 (Original SPADE-8) shows how many profiles were available from both SPADE 1 and 8. The

267 second column (SPADE 14 - Profiles from other countries) shows how many profiles were

268 copied from other countries, and the third column (SPADE 14 - Modified profiles) shows how

269 many profiles that were created by the working group by adjusting existing profiles in order to270 complete the national datasets.

270 complete the national dat

271

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Overall, the SPADE 18 (level 2) database contains soil analytical data from 1831 profiles which
is about 60% more than the number of profiles in SPADE 14 (level 1) containing soil analytical





276	data from 1078 profiles, which is, almost a doubling of the number of profiles available in
277	SPADE 1 and 8. Most of the profiles originally lacking data had allocated datasets from
278	complete profiles from other countries. Yet, ~15% of the dominant profiles specified by soil type
279	and texture were not present in either SPADE 1 nor 8 and had to be constructed by modifying
280	other existing profile datasets to fit the required soil classification. Eight countries did not deliver
281	data to SPADE 1 nor 8. Thus, datasets for these countries were exclusively based on imported or
282	constructed datasets. Stakeholders have been notified throughout this project that they may
283	update their national datasets at any time by contacting the responsible ESDAC office.
284	
285	Creating a pilot version of the SPADE 18 level 2 database (SPADE-18)
286	As described previously, the SPADE framework has four levels. The level 2 database contains
287	the same type of analytical data as the level 1 database, but in addition to the dominating soil
288	types, the inclusions and associations have been assigned estimated analytical data. This allows
289	for a substantial improvement in the accuracy of estimated soil characteristics (e.g. irrigation
290	need or carbon stocks within each SMU).
291	
292	In 2017, a working group from the European Soils Bureau and University of Copenhagen
293	discussed the methodology for creating a level 2 SPADE database (SPADE 18). Given that it
294	took about 20 years to create the level 1 database, it was decided to speed up the process by
295	following the route used to finalise SPADE-14. The following concept has been developed based
296	on data from two member states, Denmark and UK.
297	





298	1: For each country unique combinations of all soil types and topsoil textures present as
299	dominant, associated or included STUs were listed. For UK 79 new soil types had to be added to
300	the 62 at level 1, and for Denmark this left 29 unique combinations compared to 13 at level 1,
301	where only dominant soil types were considered. Thus, 16 new soil types had to be added to the
302	Danish database.
303	
304	2: For each missing soil type, the entire level 1 database was scrutinised for the particular soil
305	type. If multiple countries contained the soil type, profiles from neighbouring countries had
306	preference. If more than one neighbouring country had the desired soil type, agricultural land use
307	had preference.
308	
309	3: In cases where the soil type did not exist as a dominating soil type for any other country in the
310	database, the soil types were taken from a database containing modified soil profile data. This
311	database was created by compiling a list of all combinations of soil and topsoil texture in the
312	entire SPADE database that did not exist as dominating in any country, and therefore had no
313	estimated data assigned at level 1 (129 unique combinations in total). In the same way as
314	described for the dominating soil types, data were estimated for these profiles by making minor
315	modifications to existing profiles. For example, a Podzol with a topsoil texture class 2 (Po-2)
316	could be created from a slight modification of the topsoil particle size distribution for a Po-1.
317	Other characteristics affected by the change in soil texture were adjusted accordingly.
318	
319	4: After completion, the level 2 database will be shared with national stakeholders for evaluation,

320 and changes can be made to any data not found to be valid or meaningful. All comments and





- 321 changes must be reported to the committee within a given period. If no responses are provided,
- the proposed database will be published, but the stakeholders are always welcome to submit their
- 323 national change requests to JRC.

324

- 325 5: The final version will be published through JRC's European Soil Data Centre (ESDAC)
- 326 website (<u>http://esdac.jrc.ec.europa.eu/</u>).
- 327
- 328

## 329 SPADE applications: Root zone capacity and soil organic matter stocks in Europe

- 330 Earlier versions of the SPADE have been used to estimate soil organic C-stocks (European
- 331 Commission, 2005). More recently, it was used to map wheel load carrying capacity in Europe
- 332 (Schjønning et al. 2015).

333

## 334 Root zone capacity to 100 cm

335 To demonstrate the use of the complete SPADE level 1 database for a relevant soil property, we

- calculated the plant available water for crops having an effective root depth of 100 cm (e.g.
- barley), also called root zone capacity (RZC<sub>100</sub>) (Figure 2). Crop production on soils with RZC<sub>100</sub>
- 338 <50 mm in Northern Europe and <100 mm in Southern Europe is highly dependent on irrigation.
- 339 RZC was estimated from the following equation:

340

341 
$$RZC_{100} = \sum_{i=100} (VWC_{100i} - VWC_{1500i}) \times D_i$$





343	where $RZC_{100}$ is the cumulated root zone capacity (mm) within the upper 100 cm , $VWC_{1500i}$ is
344	the volumetric water content at 1500 kPa suction for horizon $i$ (%), $VWC_{100i}$ is the volumetric
345	water content at 100 kPa suction for horizon $i$ (%), and $D_i$ is the depth of horizon $i$ (mm).
346	Areas with very high $RZC_{100}$ (> 300 mm), relate mainly to the occurrence of Histosols, Gleysols
347	and Fluvisols, which are affected by shallow groundwater tables and few well-drained soils with
348	high silt and fine sand content (Figure 2). Soils with high $RZC_{100}$ are common in the Loess Belt,
349	just south of the ice margin from the previous ice ages, e.g. Belgium and Germany. The medium
350	RZC <sub>100</sub> , 100-200 mm, corresponds mainly to loamy soils, for instance dominating in Eastern
351	Denmark, England and Poland, while sandy soils and some shallow loamy soils have a low
352	RZC100 of 50-100 mm, e.g. Western Denmark and Sweden. Very shallow soils (Leptosols) have
353	a very low $RZC_{100}$ of 0-50 mm, which are found primarily in mountainous regions such as the
354	Alps, coastal Norway and large parts of Greece.
355	
356	
357	
358	SOC stock to 100 cm for Europe

359 We estimated the SOC stock for Europe from the following equation:

$$360 \quad SOC_{100} = \sum_{i=1}^{n} p_i SOC_i D_i A$$

361 where  $SOC_{100}$  is the cumulated SOC stock to 100 cm depth,  $p_i$  is the bulk density of horizon *i*,

362  $SOC_i$  is the SOC concentration for horizon *i*,  $D_i$  is the depth of horizon *i*, and A is the area of the

363 particular STU (Figure 3). The regional distribution of soil organic C stocks is similar to what





364	was found previously (European Environmental Agency 2012; Panagos et al. 2013) with the
365	highest stocks concentrated in areas dominated by histosols (e.g. Northwestern British Isles and
366	Finland, Figure 3). Intermediate stocks are situated in the wet Northwestern Iberian peninsula, in
367	the Massif Central region in France, and in the interior parts of the Scandinavian Peninsula,
368	while soils with relatively low SOC-stocks are situated in mountainous areas (e.g. coastal
369	Norway), dry Mediterranean areas, and areas under intensive cultivation (e.g. Northern France,
370	Germany, Denmark).
371	Our estimated cumulated SOC stock for Europe (0-100 cm) based on SPADE 18 level 1 is 76 x
372	$10^{15}$ g. This corresponds to the estimate of 75 x $10^{15}$ g obtained by the European Environment
373	Agency (2012) and the EC Joint Research Centre (Panagos et al. 2013) based on an earlier
374	version of the database, showing that our approach produces a similar result as obtained from
375	pedotransfer functions. We did not find other estimates of European SOC stocks across
376	landscape types in the scientific literature. However, as an approximation we may sum up the
377	recent estimates of SOC stocks in agricultural and forest soils. The forest SOC stock in Europe
378	(0-100 cm) was estimated to $22 \times 10^{15}$ g (De Vos et al. 2015), while the agricultural SOC stock
379	(0-30 cm) was estimated to $18 \times 10^{15}$ g (Lugato et al. 2014). This sums up to $40 \times 10^{15}$ g SOC,
380	which is only about half of our SPADE 18 level 1 estimate. However, over-/underestimation of
381	~40-100% when comparing to other studies is similar to what was discovered by others (De Vos
382	et al. 2015; Guevara et al. 2018; Lugato et al. 2014). Hence, work still remains on elucidating the
383	underlying sources of variation to find the best approach, as estimates of SOC is considered an
384	important indicator of environmental health (European Environment Agency 2012; Panagos et
385	al. 2013).





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- 389 Better estimates with SPADE 18 level 2: SOC stock in Denmark
- 390 The application of SPADE 18 level 2 data has been tested in a pilot study calculating the RZC
- for wheat in Denmark (Jensen et al. 1998). They found a substantial difference of up to 49% in
- 392 estimated national RZC values when comparing level 1 to level 2 data. To demonstrate the added
- value from including the associations and inclusions in another example, we calculated the soil
- organic carbon stock (SOC) to 1 m depth for Denmark based on SPADE 14 level 1 (Figure 4a)
- and SPADE 18 level 2 (Figure 4b) data.

Overall, the comparison shows that the estimated total SOC stock in the upper metre of Danish 396 soils increases by 14% from 478 x  $10^{12}$  to 545 x  $10^{12}$  g C when using level 2 data instead of level 397 398 1. This number is comparable to the most recent estimate obtained from digital soil mapping of about 570 x 10<sup>12</sup> g C (Adhikari et al. 2014) and previous estimates ranging from 563-598 x 10<sup>12</sup> 399 C (Krogh et al. 2003), suggesting that using level 2 data yields more reliable results. The 400 401 increase in SOC-stock using level 2 compared to level 1 data is mostly due to SOC-rich soils 402 such as Histosols, Gleysols and Fluvisols primarily present as associations or inclusions. The 403 spatial distribution of the changes reveals that particularly in areas dominated by loamy soils, the inclusion of subordinate soil types increased the SOC stock substantially (Figure 4c), 404 405 occasionally more than 40%. For sandy soils, the carbon gain was modest, typically less than 20%. Only in small loamy areas in Western Jutland and on the raised seafloors in Northern 406

- 407 Jutland dominated by wetland soils, did the carbon content decrease by using the level 2
- 408 database, probably due to the inclusion of sandy soils with low organic matter content. This





- 409 study demonstrates the added accuracy of estimating an environmentally relevant soil property
- 410 like SOC stock by the more detailed level 2 database.
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### 414 Limitations of our approach

415 Digital soil mapping (DSM, reviewed in Mulder et al. (2011); Minasny and McBratney (2016);

416 Zhang et al. 2017) is the future of soil mapping, and is constantly developing and improving (e.g.

417 Møller et al. 2019; Pouladi et al. 2019; Stockmann et al. 2015; Zeraatpisheh et al. 2019). The

418 great advantage of these formalised approaches are their reproducibility and ability to estimate

the accuracy of their predictions. However, as mentioned earlier, challenges to such inference

420 techniques persist, and no adequate and harmonised methodology for large-scale analyses has yet

- 421 been developed (Mulder et al. 2011; Zhang et al. 2017). Until these tools are developed, we
- 422 argue that databases with analytical soil properties estimated or evaluated by local expert
- 423 stakeholders is still a feasible way of assessing large-scale soil property patterns. Similar

424 conclusions underlie data harmonisation initiatives at the global scale lead by ISRIC, which has

led to the construction of the Global Soil Map (Arrouays et al. 2014) and the SoilGrids1km

426 (Hengl et al. 2014) related to the Harmonized World Soil Database (HWSD, Nachtergaele et al.

427 2014). The HWSD contains soil properties data gathered in various ways, resulting in

428 considerable variation in confidence levels. The European dataset for the HWSD is retrieved

- 429 from the European Soil Database, which comprises the information from the most recent SPADE
- 430 dataset (i.e. the one presented in this paper).





431	A consideration with respect to the interpretation of outputs from bottom-up harmonised
432	databases, like SPADE, is how well the mapping units actually reflect real landscape
433	delineations (Figure 1a). Efforts have been made by the ESDAC to let mapping units overlap
434	arbitrary administrative limits, such as national borders, to best fit the SMU delineations on both
435	sides (e.g. European Commission 2005). However, the inherent variation in level of detail from
436	the national datasets is still evident in certain areas (see for instance the Danish-German border).
437	Hence, the predictions based on the current dataset might be substantially improved by modern
438	downscaling techniques (as an example, see Peng et al. 2017 for a review of the downscaling of
439	soil moisture). This was beyond the scope of the current work, but should be a priority in future
440	large-scale soil mapping efforts.

441

# 442 **Conclusions**

We document the development of a full-covered EU-wide soil database, containing analytical
data connected to the Soil Map of Europe at scale 1:1,000,000. We demonstrate the benefits of

445 careful analysis of legacy data, wherever possible with the help of national soil experts.

446 The application of the current soil analytical database at level 1 was demonstrated by calculating

the root zone capacity for the Europe and associated countries, mapping out areas where severe

- 448 need of irrigation for crop production might occur. Moreover, we estimate the SOC stock for
- Europe to 76 x  $10^{15}$  g, which is larger than previous estimates. The increased accuracy obtained
- 450 by including associated and included soil types in the SPADE database, was demonstrated by
- 451 comparing the SOC stock of Denmark calculated from level 1 and level 2 data, showing an
- 452 increase of 14 % from 478 x  $10^{12}$  to 545 x  $10^{12}$  g C, which is more comparable to literature





- 453 estimates obtained with other methods. This exercise highlights the need for a level-2 database
- 454 for the entire European continent.
- 455 Perhaps the greatest contribution of this research to the management and protection of Europe's
- soils is the harmonisation of detailed soil profile data, hitherto unavailable across regions, but
- 457 now connected to the latest soil mapping.

458

## 459 Acknowledgements

- 460 We want to warmly thank our late colleague and friend, Professor Henrik Breuning-Madsen,
- 461 who passed during the preparation of this manuscript. He has been a key figure in moving
- 462 European soil science forward over more than three decades. This work was financially
- supported by the European Union through the EC Joint Research Centre. We thank all national
- 464 stakeholders for their contributions to the development of the SPADE database. For a full list of
- 465 stakeholders we refer to ESDAC's homepage <u>http://esdac.jrc.ec.europa.eu/.</u>





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- 657 Table 1: Average bulk densities calculated from the SPADE 2 database. The mean, standard
- 658 deviation and the number of observations (n) are shown.

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OM	Depth	<b>Bulk Density</b>	Std. dev.	n
%	cm	g cm <sup>-3</sup>	g cm <sup>-3</sup>	
90-100		0.1	0.13	165
80-90		0.1	0.05	81
70-80		0.2	0.11	64
60-70		0.2	0.13	36
50-60		0.3	0.13	25
40-50		0.4	0.08	28
30-40		0.4	0.17	19
20-30		0.8	0.31	35
10-20		1.0	0.72	176
5-10		1.1-1.2	n/a	n/a
<5	0-25	1.3	0.18	400
	25-50	1.4	0.18	726
	50-75	1.4	0.17	719
	75-100	1.5	0.14	468
	>100	1.5	0.18	714

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- 662 Table 2: The origin of SPADE data at the national level. *Original* shows the soil profiles to
- 663 which the stakeholders originally provided data; *Profiles from other countries* show the soil
- profiles for which data was copy-pasted from a similar country; *Modified profiles* show the soil
- profiles to which slight adjustments were made; Level 1 Total shows the total number of
- dominating soil profiles; Level 2 Total shows the total number of profiles, when associated soil
- 667 types were included.

Country	Original	Profiles from	Modified	Level 1	Level 2
5	(SPADE 8)	other countries	profiles	Total	Total
	· · · ·	(SPADE 14)	(SPADE 14)	(SPADE14)	(SPADE 18)
AL	14 (AL 1-14)	13 ( AL 15-27)	3 (AL 28-30)	30	49
AT	0	23 (AT 1-23)	4 (AT 24-27)	27	35
BE	42 (BE 1-42)	14 (BE 43-56)	0	56	74
BG	0	16 (BG 1-16)	7 (BG 17-23)	23	40
СН	28 (CH 1-28)	2 (CH 29-30)	7 (CH 31-37)	37	51
CZ	0	19 (CZ 1-19)	7 (CZ 20-26)	26	73
DE	60 (DE 1-60)	15 (DE 61-75)	2 (DE 76-77)	77	149
DK	13 (DK 1-13)	0	0	13	29
EE	11 (EE 1-11)	2 (EE 12-13)	4 (EE 14-17)	17	26
ES	26 (ES 1-26)	15 (ES 27-41)	8 (ES 42-49)	49	65
FI	6 (FI 1-6)	1 (FI 7)	0	7	12
FR	118 (FR 1-118)	35 (FR 119-153)	22 (FR 154-	175	230
			175)		
GB	41 (GB 1-41)	15 (GB 42-56)	6 (GB 56-62)	62	141
GR	10 (GR 1-10)	15 (GR 11-25)	4 (GR 26-29)	29	66
HU	40 (HU 1-40)	10 (HU 41-50)	11 (HU 51-61)	61	92
IE	18 (IE 1-18)	4 (IE 19-22)	3 (IE 23-25)	25	44
IT	21 (IT 1-21)	11 (IT 22-32)	9 (IT 33-41)	41	91
LT	0	20 (LT 1-20)	8 (LT 21-28)	28	52
LU	0	10 (LU 1-10)	2 (LU 11-12)	12	26
LV	26 (LV 1-26)	0	0	26	39
NL	20 (NL1-20)	12 (NL21-32)	0	32	42
NO	15 (NO1-15)	0	1 (NO16)	16	23
PL	0	28 (PL1-28)	12 (PL29-40)	40	63
РТ	18 (PT 1-18)	10 (PT 19-28)	4 (PT 29-32)	32	66
RO	28 (RO 1-28)	29 (RO 29-57)	21 (RO 58-78)	78	116
SE	0	9 (SE 1-9)	3 (SE 10-12)	12	21
SI	0	15 (SI 1-15)	9 (SI 16-24)	24	31
SK	17 (SK 1-17)	6 (SK 18-23)	1 (SK 24)	24	73
Total	571 (31%)	349 (19%)	158 (9%)	1079 (59%)	1819 (100%)







- **Figure 1**: a) Structure of the European Soil Database to which SPADE provides data (after
- Lambert et al., 2003), b) Timeline of the establishment of the Soil Profile Analytical Database of
- 672 Europe (SPADE). See text for details.
- 673









**Figure 2:** Plant available water content in mm within the uppermost one metre of the soil. Very

- 676 low 0-50 mm; low 50-100 mm, medium 100-200 mm; high 200-300 mm; very high >300
- 677 (mainly Histosol, Gleysol and Fluvisol).
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- **Figure 3:** The soil organic carbon stocks (t ha<sup>-1</sup>) in Europe within the upper 100 cm of soil
- calculated based on level 1 data (dominating soil types only).







**Figure 4:** Soil organic carbon stocks (t ha-1) in Denmark within the upper 100 cm of the soil

- calculated based on a) SPADE 18 level 1 data, and b) SPADE 18 level 2 data. c) Shows the
- relative change from level 1 to level 2 in %.