

# Digital soil mapping in Australia. Can it achieve its goals?

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## ABSTRACT

Digital soil mapping (DSM) has been used from the national to paddock scale in Australia over the past 20 years. However, there has been insufficient objective evaluation of the limitations of DSM. The continual evolution of DSM methods over time have led to a lack of operational stability that creates an ongoing risk associated with the method. The inherent modelling paradigm of DSM (a reliance on systematic variation) is a key factor that creates potentially significant constraints to the use of DSM in Australia, particularly in the context of different scales of application. Inherent covariate limitations create a further potential ceiling to what can be achieved with DSM at any point in time. As part of a more critical and objective approach to the use of the method in Australia, there is a need for more effective exploration and discussion of these and other constraints in the DSM approach. This will hopefully ensure that it is used in a fit-for-purpose and effective manner in the future.

**Keywords:** communication, covariates, expert knowledge, geomorphic, mapping, pedology, pedometrics, soil survey.

## Introduction

Digital soil mapping (DSM) methods have been in development and use in Australia and elsewhere for more than two decades. Their purpose and use in Australia has been described in numerous articles, such as [McBratney \*et al.\* \(2000\)](#), [McBratney \*et al.\* \(2003\)](#), [Holmes \*et al.\* \(2015\)](#), [Kidd \*et al.\* \(2018\)](#) and [Kidd \*et al.\* \(2020\)](#). The use of DSM fits within a broader pedometric goal of providing soil related technology and data that are faster, cheaper and continually updated ([McBratney \*et al.\* 2019](#)). Conventional survey is generally viewed as soil survey via the ‘mental model’ of the soil surveyor ([Hudson 1992](#)) and is therefore regarded (in the DSM literature) as inherently flawed and biased by the ‘human factor’. The literature regarding DSM also promotes a view that DSM is an alternative to, or replacement for, ‘conventional’ soil survey (for example, see [Grundy \*et al.\* 2020](#)).

Early in the development of DSM, [McBratney \*et al.\* \(2003\)](#) listed seven topics that needed further research in order for DSM to achieve its intended goals. These topics were: (1) environmental covariates; (2) spatial decomposition; (3) sampling methods; (4) quantitative modelling; (5) quality assessment; (6) (re)presentation of digital soil maps; and (7) economics of digital soil mapping. [Brevik \*et al.\* \(2016\)](#) provided a similar list of topics for attention in their review of DSM. Authors such as [Zhang \*et al.\* \(2017\)](#) have described other challenges that exist for DSM, in particular those associated with the trend of DSM being applied to ever-larger areas, with an associated increase in model complexity. [Grunwald \*et al.\* \(2011\)](#) considered the philosophical views of mapping soils and the issues that impact upon DSM, suggesting that a new generation of DSM was needed that explicitly considers scaling across space and time, with improved linkages between the natural and human domains.

There has been significant progress in relation to some of these matters over the past two decades. For example, the TERN facility (<https://www.tern.org.au/tern-observatory/tern-landscapes/>) has made accessing spatial data of environmental covariates across the nation easy. Developments have also occurred in spatial decomposition ([Odgers \*et al.\* 2014](#)) and

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statistically based sampling strategies (Minasny and McBratney 2006; Kidd *et al.* 2015). The thoughts of McBratney *et al.* (2003); Brevik *et al.* (2016) and others remain valid to this day, but it is apparent that not all topics have been sufficiently raised or addressed, leading to some gaps in the understanding of the abilities of DSM to be a viable operational approach to soil survey. We suggest there are some additional matters that need attention and a significant question remains as to whether DSM is a viable alternative or replacement to conventional soil survey in a wide range of settings. The full analysis of such a topic is beyond the scope of this paper; our purpose here is to highlight the need for it.

### Chasing the new horizon vs operational stability

The development and evaluation of DSM in an operational sense in Australia commenced with the Enhanced Resource Assessment (ERA) program in Queensland in the late 1990s (Slater and Grundy 1999). ERA was seen as a holistic toolkit of improvements to soil survey methods, of which DSM was only one part. Since then, there has been shift away from the holistic view of ERA, to a focus on DSM as a singular interest. There has been a continual evolution of paradigms, methods and tools (software) associated with DSM, from crisp soil surveyor driven decision trees in the 1990s, to fuzzy rule sets in the early 2000s; e.g. Grundy and Searle (1998), Brough *et al.* (2002), Claridge and Grundy (2003). Disaggregation (Odgers *et al.* 2015) became popular for a while, while the current focus is on machine learning (Taghizadeh-Mehrjardi *et al.* 2021). Expert knowledge was a critical feature in the early days, as the paradigm concerned capturing explicitly ‘the knowledge

in the soil surveyor’s head’; i.e. a hybrid approach. A shift away from this occurred post-2000, but in more recent times, there has been a return to the concept of incorporating expert knowledge (Bui *et al.* 2020). The continual change in DSM methods is invariably portrayed as a positive illustration of the ongoing research, development and improvement of the approach. However, there is a corollary that from an operational sense, any use of a DSM method has invariably been a trial; i.e. the particular method had not been used in that context or landscape previously and consequently, the outcomes of its use were not predictable or certain. To a large extent, this remains the case today.

The continual evolution of DSM has led to a paradigm we call ‘chasing the new horizon’. It is perhaps best illustrated by the frequency with which the phrase ‘DSM has the potential to ...’ is used in DSM related publications and conference presentations. This is in comparison to conventional survey, in which the outputs and outcomes are well known and understood by both producers and users, before the survey is undertaken. This is in part because of long existence of conventional soil survey but also the large number of soil survey related standards published over time. In particular, the 1990s onwards saw the culmination (in print) of >60 years of evolution of standards and methods within conventional survey in Australia (Table 1). By comparison, DSM has only produced the GlobalSoilMap standard (Arrouays *et al.* 2014). Unfortunately, evidence from published DSM surveys within Australia suggests that adherence to standards is *ad hoc* at best and in some instances, poor; e.g. the publication of DSM data to a precision beyond that specified in the standard (and common sense) in the NAWRA explorer website (<https://nawra-explorer.csiro.au/#soil> accessed 13 February 2020) and in downloadable data for the Soil Grid of Australia (Viscarra Rossel *et al.* 2014).

**Table 1.** Soil survey related standards developed over time in Australia, and digital soil mapping standards relevant to Australia.

Standard	1980–1989	1990–1999	2000–2009	2010–2019	2020–
Guidelines for conducting surveys		First edn McDonald <i>et al.</i> (1990)	Second edn McKenzie <i>et al.</i> (2008)		
Field handbook	First edn McDonald <i>et al.</i> (1984)	Second edn McDonald <i>et al.</i> (1990)	Third edn NCST (2009)		Fourth edn in progress
Soil chemical methods		First edn Rayment and Higginson (1992)	Second edn Rayment and Lyons (2011)		
Soil physical methods		First edn McKenzie <i>et al.</i> (2002)			
Australian soil classification		First edn Isbell (1996)	Revised edn Isbell (2002)	Second edn Isbell and NCST (2016)	Third edn Isbell and NCST (2021)
Data standards		Sites v1 Peluso and McDonald (1995)	ASRIS 1.1 McKenzie <i>et al.</i> (2004)	ASRIS 1.6 McKenzie <i>et al.</i> (2012); Sites v2 Jacquier <i>et al.</i> (2012)	
Digital soil mapping				GlobalSoilMap Arrouays <i>et al.</i> (2014)	

Many of these standards are potentially just as relevant to DSM as conventional survey, and some, such as [McKenzie et al. \(2008\)](#), explicitly cover both. However, it is a reality that in the same era that soil survey in Australia has matured and documented its standards, there has also been the appearance and use of a new mapping approach lacking clear standards. It is certainly true that some standards for conventional soil survey are ill-defined, but there have historically been sufficient informal and formal standards in place that soil survey organisations generated high quality, consistent products produced to a common set of standards at that point in time. This by no means suggests that conventional soil survey products are 'perfect'. There are many known problems of consistency between conventional surveys, within and between states. The question is whether DSM is the only solution to this.

A common criticism of conventional surveys is that they are not detailed enough, not accurate enough or did not capture the data needed today. While these are valid statements, they potentially also reflect a lack of understanding of the context of historical surveys in terms of project scope, methods and technology available at the time. For example, to expect a 1:500 000 scale land systems survey from 1970 to represent modern soil attributes at the paddock scale is entirely unrealistic. Is this a fault of the survey methodology or a lack of understanding by the user of the nature and limits of the data?

Some authors, such as [Kidd et al. \(2020\)](#), would argue that DSM has achieved operational stability. This is in part because DSM is seen by many as a general approach, rather than a specific method, thus frequent changes in the specific software/algorithms/data used is seen as advantageous flexibility rather than problematic. There are some texts published that attempt to provide standardised approaches, for example, [Malone et al. \(2017\)](#). However, these run a real risk of becoming outdated within a short period of time of being published. One indicator of operational stability is that products of the work are used beyond the purpose of their original creation. DSM products of the past 20 years have often not met this criteria compared to the ongoing use of soil survey products that are more than 60 years old.

We argue in a separate paper (the authors' unpubl. data) that use of DSM does not represent operational stability. We also suggest the risks associated with continual operational implementation of new approaches have not been sufficiently evaluated. Whether it be from the perspective of planning and funding surveys, teaching/training or consistency between products/outputs, constantly changing methods does create operational problems. Operational stability is important for funding bodies and clients, as they have some certainty as to what they are getting and that it is consistent with other surveys. Operational certainty is also important from a technical/productivity perspective; the ability to produce survey outputs in long-term programs is hampered if effort

is continually expended developing and learning new approaches. This does not mean that method development is unimportant; rather, that a balance must be struck between producing known products to recognised consistent standards vs the aspirational goal of producing 'something new'. We suggest that there has been an imbalance and consequently there are ongoing risks.

## Limitations and limits

The paradigm of 'chasing the new horizon' has at times led to an overly optimistic view of the capabilities of DSM as an operational tool. The future use of DSM in Australia requires a much clearer understanding of the operational limitations of the approach and the fundamental limits of the DSM paradigm. Operational limitations are a function of the operational environment at any point in time and may be resolved via changes such as the development of new tools/methods/technology or increased resources. Some of them are related to the translation of academic theory to operational use; statistically-based sampling strategies are a good example. Limits to the DSM approach are, however, inherent within the paradigm and consequently are less likely to be easily resolved. Below, we briefly discuss three examples that are highly relevant to the current implementation of DSM in Australia.

- The ability of the modelling and mapping frameworks to represent the landscape continuum and the outcomes of pedogenetic processes.
- The inherent limitations associated with covariates.
- Lack of primary (site) data.

## The mapping/modelling framework

[Wilding and Drees \(1983\)](#) suggested that variability in the natural landscape is inherently systematic, but the degree to which it appears non-random is influenced by the scale of observation. It is rare to find a landscape where the apparent variation in soil types or attributes will be entirely systematic ([Wilding 1994](#)), particularly at the normal scales of investigation in Australia (1:50 000 to 1:250 000). Consequently, there is always an inherent potential for the inability to fully predict all soil and landscape features within the self-imposed limits of the survey scale (this should not, however, be used as an excuse to produce poor quality survey outputs). DSM appears to be at risk of being inherently limited by the presence of non-systematic variability, as the method relies entirely upon the presumption that the analysis of covariate data will yield systematic relationships. Where does this limit sit and how does it vary with differing soil and land attributes, locations and scale? DSM has been used at the continental scale, but

at the same time there is a push for it to be used at the paddock scale; e.g. [Zhao \*et al.\* \(2020\)](#). How these limits in relation to systematic variability express at different scales has not been explored in sufficient detail.

While DSM surveys report uncertainty statistics for individual attributes being predicted, there has been little analysis of the meaning of these results in the context of systematic vs non-systematic variability at particular scales. There is little value in using complex means to predict highly predictable attributes. Similarly, there is little value in complex attempts to predict highly unpredictable attributes. These are key questions requiring investigation, as they help set the bounds for when DSM is (or is not) appropriate, prior to a methodology choice in a project planning phase.

Another aspect of the modelling/mapping framework concerns the geomorphic construct. This is a widely used basis of most soil survey but is largely absent within DSM. Its absence creates model inefficiencies but also hampers communication of products. A key virtue of a conventional soil survey map is the ability to see differing scales within the landscape at once; simply via the use of good cartographic practise and the geomorphic construct. This ability has yet to be proven with DSM and remains a key limitation to the way in which DSM products are currently used and communicated.

A final limitation worth considering is the nature of covariates. The suite of available covariates has improved over the past 20 years in terms of distribution, resolution and diversity. There is, however, a fundamental assumption within DSM that the available suite will somehow represent the outcomes of both past and present pedogenic environments. Whether expressed in its original form as CLORPT ([Jenny 1941](#)) or in the DSM form as SCORPAN ([McBratney \*et al.\* 2003](#)), the pedogenetic model is often cited as underpinning the DSM approach. Despite this, no operational DSM projects in Australia have utilised a pedogenetic framework. Rather they have been models that sought correlation in data, with an associated assumption that the correlation will sufficiently represent pedogenetic outcomes. Correlation does not represent causation and in this regard, DSM is no different to conventional survey that relies on visual and mental correlation. The current approach within DSM is to utilise as many covariates as possible and 'hope for/pick the best'. The poor predictive outcomes within some DSM projects suggests that this is not the best approach. There are also key limitations in relation to covariate resolution and locational accuracy that remain operationally significant and are likely to do so for a long time. Importantly, few covariates represent the below-surface environment effectively. Given many of the soil attributes of interest concern the subsoil, the ability to predict these in the absence of covariates representing that domain is likely to remain limited.

When acknowledged, the issues associated with covariates are viewed as an operational limitation rather than an inherent limit. Once again, the 'new horizon' comes into

play, with the most common response being 'soon we will have more/better covariates'. Given that the government programs that collect/derive the fundamental data supporting most covariates are outside of the pedological domain, the likelihood that the operational limitations of covariates will be resolved any time soon is a risk for DSM. Importantly, the inability of common covariates to represent pedogenic outcomes in the landscape are likely to remain into the future.

## Lack of primary site data

The operational impact of a lack of site data on the predictive capacity of DSM models is well recognised ([Brevik \*et al.\* 2016](#); [Zhang \*et al.\* 2017](#)). It is a problem that is frequently commented upon as a constraint to outcomes achieved in DSM projects; e.g. [Bui \*et al.\* \(2006\)](#), [Bartley \*et al.\* \(2013\)](#), [Thomas \*et al.\* \(2018\)](#). Despite these acknowledgements, DSM has been promoted as a tool for use in the absence of site data; e.g. the disaggregation of existing polygonal data. Various strategies have been used in DSM surveys to resolve the problem of insufficient site data; for example, the inclusion of data from outside the survey area by [Thomas \*et al.\* \(2018\)](#). However, this potentially comes with a cost in increased model effort and complexity. More importantly, such an approach is only possible if suitable data are available. Given the known paucity of site data across much of Australia, there is an obvious question on whether the absence of site data is an inherent practical limit as much as an operational limitation. Yes, more site data can be collected to improve models, but the history of operational DSM in Australia to date suggests that DSM projects typically do not collect large quantities of site data. After all, one of the key virtues of DSM widely promoted is the ability to model without investing significantly in data collection. This is particularly the situation for large area modelling exercises such as the Soil Grid of Australia. If this is the case, then how will the issue be resolved? To answer this question, a better understanding is required of where the data gaps are.

One of the largest data gaps in Australia concerns cropping lands. The nature and history of soil survey in most cultivated lands in Australia is such that the data was either collected pre-development or there was a specific intent to describe the natural state. There is a significant absence of data representing soil attributes of cropping lands in soil survey databases. Thus, the use of data from state and territory soil survey databases for DSM will invariably lead to difficulties in depiction of cropping lands, in particular, for dynamic soil attributes such as pH, organic carbon and nutrients. There are multiple possible solutions to this, one of which is the targeted collection/collation of data in cropping lands. The National Soil Strategy has identified this opportunity and is investigating options regarding the collation of the large body of private sector data collected in cropping lands. Will this be the solution for representation of soil attributes in cropping lands via conventional or DSM means? Will it



provide the feedstock for a new realm of paddock-scale DSM? There are many issues of data quality, sampling, bias, incomplete attribution etc. that make this an aspirational outcome rather than an easy, realistic one. It would be unfortunate if the very limited soil science capacity in Australia started chasing this new horizon rather than dealing with known problems with easy solutions.

## The role of ‘the expert’

Early exploratory DSM works in Australia put the soil surveyor’s knowledge at the centre of the equation with attempts to capture soil surveyor knowledge explicitly through rule-based frameworks. The need for pedological expertise in DSM has long been argued (Walter *et al.* 2006) and most operational DSM projects in Australia have involved traditionally trained soil surveyors. In the past 10–15 years, there has been a strong focus on machine-based approaches and reduced input of expert knowledge. The recognition of the value of pedological knowledge in DSM has, however, been reinvigorated in recent literature (Holmes *et al.* 2015; Thomas *et al.* 2018; Arrouays *et al.* 2020a, 2020b), although the reasons suggested for the use of that knowledge are varied. One reason may be a recognition that the machine does not have all the answers.

Rossiter (2018) warned that the combination of computing power and increasing data availability enables pedology-free inference of soil and attribute distribution. Biggs *et al.* (2018) provided a similar caution in their discussion of the current state of pedology in Australia, and suggested that there is a risk that the increased adoption of DSM will in fact lead to the decline of skills and knowledge within the pedological and soil survey community. There is a risk that DSM methods will lead to a perception that soil scientists are not needed and it only requires a ‘modelling expert’ to make a soil map. It is ironic that at a time when universities around Australia are struggling to maintain the teaching of soil science (Rogers *et al.* 2020), elements of the soil science profession have developed a methodology that may remove a potential career pathway for soil science graduates. A ‘modelling expert’ is not necessarily a soil scientist.

The renewed recognition of the role of experts in soil survey has unfortunately come too late. The retirement of the ‘baby boomer’ generation has led to a large reduction in the number of skilled pedologists/soil surveyors in the country; noting that these people gained their expertise via decades of conventional survey. Given that DSM inherently contains a paradigm of less fieldwork than conventional survey, its promotion and use in the past 20 years has led to a reduction in the pedological skills of newer generations. The supply and training of new pedologists is a larger issue than just the question of survey method, but it is indisputable that the skill of a soil surveyor/pedologist is inherently related to

time spent in the field, and that DSM involves a paradigm of less fieldwork. There is consequently a real risk that in the future, soil survey (in the form of DSM) will more and more become the domain of desktop theorists with little real field experience.

## A new era?

In their pedometrics timeline, McBratney *et al.* (2019) described a number of stages in the evolution of pedometrics and DSM, and proposed that we are now in the ‘Global Mapping’ era. We posit that perhaps this should be re-titled the ‘coming down to earth’ era. Apart from the matters we have discussed above, this new era would involve acknowledgement that DSM is not the universal panacea to deriving soil data across the country. More importantly, it would involve critical analysis of the risks and potential perverse outcomes associated with the over-promotion and inappropriate use of DSM. The view promoted in the literature to date has been entirely ‘positive’ and there has been a lack of balanced evaluation.

There are some hints that this recognition is occurring; for example, both Arrouays *et al.* (2020a) and Ma *et al.* (2019) have acknowledged some of the problems associated with DSM. A more critical approach to determining the value of DSM will help identify where it fits best and where the greatest value may be obtained from the methodology in conjunction with existing methods. Recognition of the value of hybrid approaches is nothing new; it was suggested by McKenzie and Grundy (2008) and is implicit in the framework set out by Slater and Grundy (1999). Hybrid approaches do not, however, feature from an operational perspective. Divergence between academic theory and operational reality is a common thread in the DSM world. Progress would perhaps be measured by a decrease in the frequency of use of the statement ‘DSM has the potential to ...’ and its replacement with ‘DSM can ...’ Importantly, we would see a reduction in DSM projects that yield results that do not meet mapping standards or last the test of time.

We indicated earlier, as have authors such as Ma *et al.* (2019) and Kerry and Oliver (2011), that one of the key potential values of pedometric methods is the identification of relationships between soil attributes and pedogenetic factors. Knowledge of these relationships is of significant value for many purposes, irrespective of whether they are spatially depicted. Knowledge transcends scale. Investigating these relationships in greater detail should be a priority activity in Australia, as the knowledge generated is applicable at many scales and to many purposes, one of which is DSM. The work of Malone and Searle (2021) in considering relationships between field texture and particle size analysis is a recent example of this type of analysis. Importantly, such knowledge is useful at the paddock/farm scale without spatial

depiction and is likely to be of value now, compared to the unrealistic goal of deriving paddock scale spatial soil attribute data.

Moving forwards, a more balanced approach to the merits and use of DSM (or not) must be embraced. The concerns of soil surveyors must be heard and understood by DSM practitioners. Arrouays et al. (2020a) commented that it is expected that conventional surveyors will oppose quantitative algorithmic approaches. Kidd et al. (2020) described 'barriers' to the adoption of DSM in Australia. Such a depiction of conventional soil surveyors is by no means beneficial and presumes that the questions and concerns raised by them are inconsequential, rather than valid questions of scientific rigour. Practising soil surveyors (such as the authors) are not against new methods or technologies, provided the public and scientific benefit is clearly demonstrated. This has been well illustrated by the early adoption of a variety of technologies and associated method changes in conventional soil survey in the past 30 years; e.g. GIS, GPS, digital cameras, digital field devices etc. The DSM discipline has not sufficiently and objectively demonstrated the scientific or public benefit associated with the approach, which is why there is 'resistance' amongst parts of the soil survey community. The soil survey community is quite rightly sceptical of blackbox approaches, which ironically suffer from many of the same problems that have been ascribed to conventional survey: the output has been derived using opaque, inefficient processes that cannot be replicated.

Authors such as Bui et al. (2020) have argued that soil surveyors must become geostatisticians. We suggest otherwise. The typical soil surveyor is not a geostatistician or mathematician and should never be. Their core skill is, and should remain, pedology. Otherwise, they cannot call themselves a soil surveyor. Their skill lies in their field craft and ability to observe and understand landscapes and importantly, communicate that knowledge to a wide range of audiences through a variety of pathways. As discussed earlier, DSM methods do not lead to extensive training of field pedologists and this problem must be recognised and dealt with. The future era must ensure that public investment in soil survey leads to the creation of soil surveyors/pedologists and primary data to service future questions, rather than the extinction of the discipline.

## Conclusions

Over the past 25 years, considerable energy, time and money has been devoted to the research, development and operational implementation of DSM in Australia. Has it been value for money? Have the outcomes and outputs been better/greater than if the same time and money was invested in conventional survey? Will it be value for money into the future? An objective view would suggest the

answer is unclear, which indicates that DSM has not yet proven its worth. More than two decades have been spent chasing the 'new horizon'. While there are some obvious outcomes, history has shown that many of the outputs have not withstood the test of time; a clear difference and benefit compared to conventional soil survey products. It is time for a more objective analysis and review of the merits (or not) of DSM and where it is (or is not) appropriate to use in Australia. This must be coupled with a strategic review of where the pedometric toolkit can answer the most important soil-related questions and where/how the significant data gaps in the nation are most efficiently filled. The National Soil Strategy offers a potential pathway to achieve these outcomes, but to do so requires a more balanced view of technical methods and standards by our profession.

## References

- Arrouays D, McBratney A, Minasny B, Hempel JW, Heuvelink G, Macmillan RA, Hartemink AE, Lagacherie P, McKenzie N (2014) The GlobalSoilMap project specifications. In 'GlobalSoilMap: basis of the global spatial soil information system – Proceedings of the 1st GlobalSoilMap conference, Introduction' (Eds D Arrouays, N McKenzie, J Hempel, A Richer de Forges, AB McBratney) pp. 9–12. (CRC Press) doi:10.1201/b16500-5
- Arrouays D, McBratney A, Bouma J, Libohova Z, Richer-de-Forges AC, Morgan CLS, Roudier P, Poggio L, Mulder VL (2020a) Impressions of digital soil maps: the good, the not so good, and making them ever better. *Geoderma Regional* 20, e00255. doi:10.1016/j.geodrs.2020.e00255
- Arrouays D, Poggio L, Salazar Guerrero OA, Mulder VL (2020b) Digital soil mapping and *GlobalSoilMap*. Main advances and ways forward. *Geoderma Regional* 21, e00265. doi:10.1016/j.geodrs.2020.e00265
- Bartley R, Thomas M, Clifford D, Philip S, Brough D, Harms B, Willis R, Gregory L, Glover M, Moodie K, Sugars M, Eyre L, Smith D, Hicks W, Petheram C (2013) Land suitability: technical methods. A technical report to the Australian Government from the CSIRO Flinders and Gilbert agricultural resource assessment, part of the North Queensland irrigated agriculture strategy. p. 125. (CSIRO: Brisbane, Qld) doi:10.4225/08/584d97f34ddf4
- Biggs A, Holz G, McKenzie D, Doyle R, Cattle S (2018) Pedology – a vanishing skill in Australia? *Soil Science Policy Journal* (1), 24–34. Available at <https://www.soilscienceaustralia.org.au/publications/soil-policy-journal/>
- Brevik EC, Calzolari C, Miller BA, Pereira P, Kabala C, Baumgarten A, Jordán A (2016) Soil mapping, classification, and pedologic modeling: history and future directions. *Geoderma* 264, 256–274. doi:10.1016/j.geoderma.2015.05.017
- Brough DM, Wilson PR, Burt SM (2002) Soil attributes and agricultural suitability of the Burnett riparian lands, Ceratodus-Auburn River. (Department of Natural Resources and Mines: Brisbane) QNRM01072.
- Bui EN, Simon D, Schoknecht N, Payne A (2006) Chapter 15 Adequate prior sampling is everything: lessons from the Ord River Basin, Australia. *Developments in Soil Science* 31, 193–205.
- Bui EN, Searle RD, Wilson PR, Philip SR, Thomas M, Brough D, Harms B, Hill JV, Holmes K, Smolinski HJ, Van Gool D (2020) Soil surveyor knowledge in digital soil mapping and assessment in Australia. *Geoderma Regional* 22, e00299. doi:10.1016/j.geodrs.2020.e00299
- Claridge J, Grundy MJ (2003) Spatial soil properties in the Lower Balonne Airborne Geophysics project. (Queensland Department of Natural Resources and Mines: Brisbane)
- Grundy MJ, Searle RD (1998) New directions in soil survey: soil landscape modelling in Bundaberg. In '14th Australian Geological Convention, Townsville, July 1998.' p. 190. (Geological Society of Australia), Abstracts No. 49.
- Grundy MJ, Searle R, Meier EA, Ringrose-Voase AJ, Kidd D, Orton TG, Triantafyllis J, Philip S, Liddicoat C, Malone B, Thomas M, Gray J,

- Bennett JM (2020) Digital soil assessment delivers impact across scales in Australia and the Philippines. *Geoderma Regional* **22**, e00314. doi:10.1016/j.geodrs.2020.e00314
- Grunwald S, Thompson JA, Boettinger JL (2011) Digital soil mapping and modeling at continental scales: finding solutions for global issues. *Soil Science Society of America Journal* **75**(4), 1201–1213. doi:10.2136/sssaj2011.0025
- Holmes KW, Griffin EA, Odgers NP (2015) Large-area spatial disaggregation of a mosaic of conventional soil maps: evaluation over Western Australia. *Soil Research* **53**(8), 865–880. doi:10.1071/SR14270
- Hudson BD (1992) The soil survey as paradigm-based science. *Soil Science Society of America Journal* **56**(3), 836–841. doi:10.2136/sssaj1992.03615995005600030027x
- Isbell RF (1996) 'The Australian soil classification.' (CSIRO Publishing: Collingwood)
- Isbell RF (2002) 'The Australian soil classification.' Revised Edn. (CSIRO Publishing: Collingwood)
- Isbell RF, NCST (2016) 'The Australian soil classification.' 2nd edn. (CSIRO Publishing: Collingwood)
- Isbell RF, NCST (2021) 'The Australian soil classification.' 3rd edn. (CSIRO Publishing: Melbourne)
- Jacquier D, Wilson P, Griffin E, Brough D (2012) Soil Information Transfer and Evaluation System (SITES) – Database design and exchange protocols. Australian Collaborative Land Evaluation Program, CSIRO, Canberra.
- Jenny H (1941) 'Factors of soil formation: a system of quantitative pedology.' (McGraw-Hill: NY)
- Kerry R, Oliver MA (2011) Soil geomorphology: identifying relations between the scale of spatial variation and soil processes using the variogram. *Geomorphology* **130**(1–2), 40–54. doi:10.1016/j.geomorph.2010.10.002
- Kidd D, Malone B, McBratney A, Minasny B, Webb M (2015) Operational sampling challenges to digital soil mapping in Tasmania, Australia. *Geoderma Regional* **4**, 1–10. doi:10.1016/j.geodrs.2014.11.002
- Kidd D, Searle R, Wilson P (2018) Digital soil mapping: application, opportunity and challenges. *Soil Science Policy Journal* (1), 35–40. Available at <https://www.soilscienceaustralia.org.au/publications/soil-policy-journal/>
- Kidd D, Searle R, Grundy M, McBratney A, Robinson N, O'Brien L, Zund P, Arruays D, Thomas M, Padarian J, Jones E, Bennett JM, Minasny B, Holmes K, Malone BP, Liddicoat C, Meier EA, Stockmann U, Wilson P, Wilford J, Payne J, Ringrose-Voase A, Slater B, Odgers N, Gray J, van Gool D, Andrews K, Harms B, Stower L, Triantafyllis J (2020) Operationalising digital soil mapping – lessons from Australia. *Geoderma Regional* **23**, e00335. doi:10.1016/j.geodrs.2020.e00335
- Ma Y, Minasny B, Malone BP, McBratney AB (2019) Pedology and digital soil mapping (DSM). *European Journal of Soil Science* **70**(2), 216–235. doi:10.1111/ejss.12790
- Malone B, Searle R (2021) Updating the Australian digital soil texture mapping (Part 1\*): re-calibration of field soil texture class centroids and description of a field soil texture conversion algorithm. *Soil Research* **59**(5), 419–434. doi:10.1071/SR20283
- Malone BP, Minasny B, McBratney AB (2017) 'Using R for digital soil mapping.' (Springer: Switzerland)
- McBratney AB, Odeh IOA, Bishop TFA, Dunbar MS, Shatar TM (2000) An overview of pedometric techniques for use in soil survey. *Geoderma* **97**(3–4), 293–327. doi:10.1016/S0016-7061(00)00043-4
- McBratney AB, Mendonça Santos ML, Minasny B (2003) On digital soil mapping. *Geoderma* **117**(1–2), 3–52. doi:10.1016/S0016-7061(03)00223-4
- McBratney A, de Gruijter J, Bryce A (2019) Pedometrics timeline. *Geoderma* **338**, 568–575. doi:10.1016/j.geoderma.2018.11.048
- McDonald RC, Isbell RF, Speight JG, Walker J, Hopkins MS (1984) 'Australian soil and land survey field handbook.' (Inkata Press: Melbourne)
- McDonald RC, Isbell RF, Speight JG, Walker J, Hopkins MS (1990) 'Australian soil and land survey field handbook.' 2nd edn. (Inkata Press: Melbourne)
- McKenzie NJ, Grundy MJ (2008) 2. Approaches to land resource survey. In 'Guidelines for surveying soil and land resources.' 2nd edn. (Eds NJ McKenzie, MJ Grundy, R Webster, AJ Ringrose-Voase) pp. 15–26. (CSIRO Publishing: Collingwood, Vic.)
- McKenzie N, Coughlan K, Cresswell H (Eds) (2002) 'Soil physical measurement and interpretation for land evaluation.' Australian soil and land survey handbook series. (CSIRO Publishing: Collingwood, Vic.)
- McKenzie NJ, Jacquier DW, Simon D (2004) ASRIS technical specifications v1.1. (Australian Collaborative Land Evaluation Program: Canberra). Available at <https://www.asris.csiro.au/downloads/ASRIS%20Tech%20Specs%201.5.pdf>
- McKenzie NJ, Grundy MJ, Webster R, Ringrose-Voase AJ (2008) 'Guidelines for surveying soil and land resources,' 2nd edn. (CSIRO Publishing: Collingwood, Vic.)
- McKenzie NJ, Jacquier DW, Maschmedt DJ, Griffin EA, Brough DM (2012) The Australian Soil Resource Information System (ASRIS) technical specifications. Revised Version 1.6. June 2012. (The Australian Collaborative Land Evaluation Program)
- Minasny B, McBratney AB (2006) A conditioned Latin hypercube method for sampling in the presence of ancillary information. *Computers & Geosciences* **32**(9), 1378–1388. doi:10.1016/j.cageo.2005.12.009
- NCST (Ed.) (2009) 'Australian soil and land survey field handbook.' 3rd Edn. (CSIRO Publishing: Collingwood, Vic.)
- Odgers NP, Sun W, McBratney AB, Minasny B, Clifford D (2014) Disaggregating and harmonising soil map units through resampled classification trees. *Geoderma* **214–215**, 91–100. doi:10.1016/j.geoderma.2013.09.024
- Odgers NP, Holmes KW, Griffin T, Liddicoat C (2015) Derivation of soil-attribute estimations from legacy soil maps. *Soil Research* **53**(8), 881–894. doi:10.1071/SR14274
- Peluso E, McDonald W (1995) Soil information transfer and evaluation system: database design and exchange protocols. ACLEP Technical Report No. 3. (CSIRO Division of Soils: Canberra)
- Rayment GE, Higginson FR (1992) 'Australian laboratory handbook of soil and water chemical methods.' (Inkata Press: Melbourne, Australia)
- Rayment GE, Lyons DJ (2011) 'Soil chemical methods : Australasia.' (CSIRO Publishing: Collingwood, Vic.)
- Rogers F, Leckie C, Pozza L, McWilliams J, Field D, Bennett JM (2020) National inventory of soil science teaching on behalf of the Australian soil network (WG6). (Soil Science Australia, University of Southern Queensland, The University of Sydney, Centre for Sustainable Agricultural Systems Publication 1007360/20/01: Toowoomba)
- Rossiter DG (2018) Past, present & future of information technology in pedometrics. *Geoderma* **324**, 131–137. doi:10.1016/j.geoderma.2018.03.009
- Slater BK, Grundy MJ (1999) Enhanced resource assessment: integration of innovative technologies by agency soil scientists in Australia. In 'Second approximation international conference on soil resources, their inventory, analysis and interpretation in the 21st century, 10–12 June 1999'. (Minneapolis, MN, USA)
- Taghizadeh-Mehrjardi R, Hamzehpour N, Hassanzadeh M, Heung B, Ghebleh Goydaragh M, Schmidt K, Scholten T (2021) Enhancing the accuracy of machine learning models using the super learner technique in digital soil mapping. *Geoderma* **399**, 115108. doi:10.1016/j.geoderma.2021.115108
- Thomas M, Brough D, Bui E, Harms B, Hill J, Holmes K, Morrison D, Philip S, Searle R, Smolinski H, Tuomi S, Van Gool D, Watson I, Wilson PL, Wilson PR (2018) Digital soil mapping of the Fitzroy, Darwin and Mitchell catchments. A technical report from the CSIRO Northern Australia water resource assessment, part of the National Water Infrastructure Development Fund: water resource assessments. (CSIRO: Australia) doi:10.25919/5b86ed6426d7a
- Viscarra Rossel R, Chen C, Grundy MJ, Searle R, Clifford D (2014) Soil and landscape grid Australia wide 3D soil property maps (3" resolution). Dataset. Release 1 v3. Available at <https://data.gov.au/dataset/ds-dap-csiro%3A11379/details?q=>
- Walter C, Lagacherie P, Follain S (2006) Chapter 22 Integrating pedological knowledge into digital soil mapping. *Developments in Soil Science* **31**, 281–301.
- Wilding LP (1994) Factors of soil formation: contributions to pedology. In 'Factors of soil formation: a fiftieth anniversary retrospective: proceedings of a symposium cosponsored by the Council on the History of Soil Science (S205.1) and Division S-5 of the Soil Science Society of America, 28 October 1991, Denver, CO.' (Eds R Amundson, J Harden, M Singer) pp. 15–30. (The Society: Madison, WI, USA) doi:10.2136/sssaspecpub33.c2

- Wilding LP, Drees LR (1983) Chapter 4 spatial variability and pedology. *Developments in Soil Science* **11**, 83–116. doi:[10.1016/S0166-2481\(08\)70599-3](https://doi.org/10.1016/S0166-2481(08)70599-3)
- Zhang G-L, Liu F, Song X-D (2017) Recent progress and future prospect of digital soil mapping: a review. *Journal of Integrative Agriculture* **16**(12), 2871–2885. doi:[10.1016/S2095-3119\(17\)61762-3](https://doi.org/10.1016/S2095-3119(17)61762-3)
- Zhao X, Arshad M, Li N, Zare E, Triantafilis J (2020) Determination of the optimal mathematical model, sample size, digital data and transect spacing to map CEC (Cation exchange capacity) in a sugarcane field. *Computers and Electronics in Agriculture* **173**, 105436. doi:[10.1016/j.compag.2020.105436](https://doi.org/10.1016/j.compag.2020.105436)

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