

Chapter 21

Variograms of Soil Properties for Agricultural and Environmental Applications

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*“Think left and think right and think low and think high. Oh,
the things you can think up if only you try”!*

DR Suess

21.1 Geostatistics and Precision Agriculture

The previous chapter describes how precision agriculture can be used to improve farm management to achieve economic and environmental benefits. Short-range differences in soil attributes mean that spatially differentiated management can create economic or environmental benefits. Effective precision agriculture requires accurate soil mapping at subfield scales so that management practices can be modified. Improvements in farming technology, for instance, GPS-controlled farm equipment, decrease the difficulty and cost associated with spatially differentiated management. This improves the ease of implementation and makes high-resolution soil maps more valuable.

A key challenge for geostatistics in precision agriculture is the detection of soil variability at important subfield scales and the systematic incorporation of this variability into accurate field scale soil maps.

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21.2 Soil Survey, the Variogram and Kriging

Soil attributes are typically difficult and expensive to observe. As a result, soil attribute maps made for the purpose of precision agriculture are typically created from point observations which represent a small proportion of the area to be mapped. Estimates of soil attributes in unknown areas are based on the observations and an expectation of the regions between them. This process of predicting attributes in unobserved areas is known as kriging (explained in more detail in Chap. 10). Assumptions about spatial variability are typically derived from the variogram, which links spatial separation distance to expectations about variability. Chapters 10 and 11 explain the variogram, its uses and the different methods of calculating the variogram in more detail.

The variogram is sometimes called the ‘cornerstone of geostatistics’. Accurate estimation of the variogram is critical to the production of accurate soil maps. Because of the hidden nature of most soil attributes, we can usually only directly observe a small proportion of an area of interest. In Chap. 10, the distinction between the experimental or empirical variogram and the model variogram is described in some detail. The empirical variogram plots the average variance against separation distance for a number of distinct lags. The model variogram uses the information from the empirical variogram to estimate the expected variability at all lags. The purpose of the model variogram is to estimate the true underlying spatial variation at a level of detail that allows useful predictions.

Interpolation of results into unobserved points depends on the spatial structure estimated by the variogram. The closer the estimated variogram to the underlying spatial structure, the more accurate the subsequent interpolation. In general, a variogram computed from samples with finer spacing and more observations will estimate the underlying spatial structure more accurately than a variogram computed from samples with coarser spacing. Finer spacing allows the detection of spatial structure across more scales. The extent to which this is true will depend on the interaction of the spacing with the underlying spatial structure. For example, if there is no spatial relationship between points more than 5 m apart, then decreasing spacing from 50 m of separation distance to 10 m will not improve the variogram. We pause here to explain some key components of the variogram and how they are affected by survey design.

21.3 Key Components of the Variogram

While the variogram can take many forms (see Chap. 10, Sect. 10.1.1, for detail on some commonly used models), there are three components which are typically considered the most important indicators of spatial structure. Shown in the stylised diagram in Fig. 21.1 and subsequently described, these components are the nugget, the sill and the range. The estimation of each of these parameters depends strongly on the sampling design.

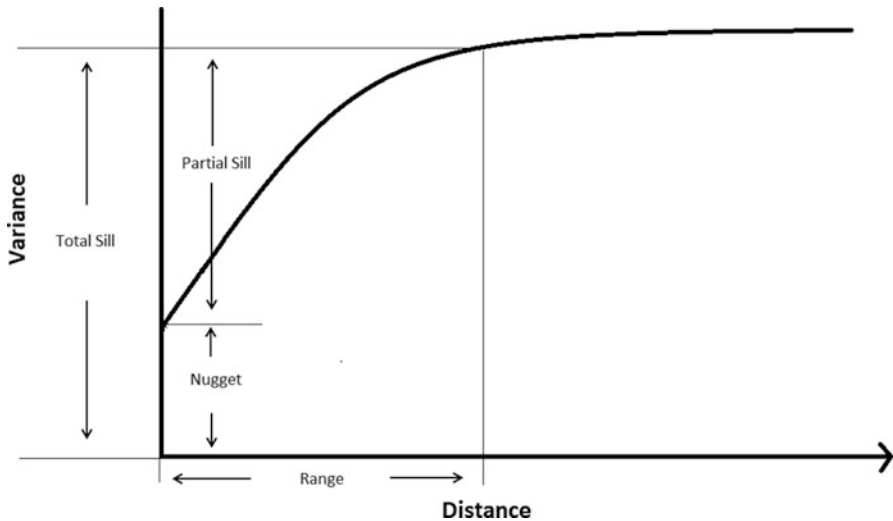


Fig. 21.1 Stylised diagram, showing the three most important indicators of spatial structure, the nugget, the sill and the range

21.3.1 Nugget

In principle the nugget effect captures nonspatial variation: measurement error; random variation. However, in practice the nugget will also capture spatial variation that occurs at scales less than the smallest sampling interval. If the sampling interval is wider than an important scale of spatial variation, then this will increase the nugget. Aliquotting (pooling of samples) will decrease the nugget. A larger support (area over which the sample is taken) will also decrease the nugget.

21.3.2 Total Sill

The total sill is defined as the maximum variability that can be expected for a particular soil property. Beyond a certain separation distance (the range, see below), the expected variability will not increase past the value of the total sill (or in some cases will increase only very slowly and slightly past the sill). Only bounded variogram models have a sill. The total sill is more likely to be affected by the number of samples and the wider separation distance than the minimum sampling spacing. When modelling spatial variability at a field scale, it is unlikely that the maximum variability of the soil will be reached. However, unbounded models are rarely fit. In the context of precision agriculture, we can think of the total sill as being the maximum variability within this particular context or a local maximum. If we extended the sampling to a regional level, it is likely we would reach another magnitude of variability.

21.3.3 *Partial Sill*

The distance between the nugget and the sill is known as the partial sill. The partial sill is the component of variation that can be spatially attributed. In Fig. 21.1, the semivariance increases linearly with distance at short separation distances. As the separation distance increases, the rate of increase in the semivariance decreases, before eventually reaching a plateau as the semivariance reaches the total sill. This pattern is commonly found in variogram models (see Chap. 10 for more detail on the functional forms). Accurately determining the change in variability between the nugget and the sill requires sufficiently dense sampling at appropriate scales of variability.

21.3.4 *Range*

Like the lag and the sill, the range estimated by the theoretical variogram will be strongly affected by the sampling design. As the lag increases, the confidence intervals widen (Oliver and Webster 2014) which can make it difficult to accurately fit a variogram at lags approaching the full extent. Precision agriculture surveys are typically conducted over areas from a few hectares to a few hundred hectares. This precludes them from capturing landscape or continental scale variability. Despite this, the majority of variogram fits from precision agriculture studies are bounded. The modelled range and sill can be thought of as the ‘local sill’ associated with the ‘local range’ associated with the particular extent and spacing of that survey. It is very likely that if the extent of the survey was increased, another degree of variability would be found. It is critical to consider the potential effect of both the extent and the spacing on the range when considering precision agriculture studies and how their findings may inform your own work.

21.4 Soil Survey Design: Capturing Spatial Variability

The task of ensuring that a soil survey captures the necessary scales of variability is not a trivial one. The expense of a soil survey, and the commonly destructive nature of soil sampling, creates a pressure to reduce the number of sampling points as much as possible. However, if sampling is insufficient or too sparse (relative to the underlying spatial structure), it will compromise the variogram and thus the accuracy of the maps which are calculated from it. It is noted in Chap. 11 that one of the chief difficulties associated with the design of a soil survey for estimating a

variogram is that the underlying spatial structure is unknown at the time the survey is designed. If the underlying spatial structure of a soil attribute is known, we can design a soil survey sufficiently to create a map at a particular level of detail.

Where budget allows it, the best practice is to undertake a preliminary soil survey to estimate important scales of variation before a more comprehensive soil survey is designed. This survey should be nested in its design to increase the chances of capturing important scales of variability (Pettitt and McBratney 1993; Webster and Oliver 2001). However, a preliminary survey will rarely be economically feasible. If it is not possible to conduct a preliminary survey, a soil survey design is likely to benefit from the consultation of alternative sources of information about soil variability, such as existing literature or covariates.

Budget or practical pressures may be sufficient to impede the collection of sufficient data to reliably calculate an empirical variogram. A stable variogram calculated from classical geostatistical methods (as described in Chap. 10) requires around 100 observations. More modern methods (as described in Chap. 11) typically require around 50. It is not possible to estimate spatial variability at distances less than the minimum separation distance. In cases where there are few or widely spread soil observations, alternative sources of information may be useful for the calculation of a variogram for kriging.

The expense of gathering soil observations creates a need for cheaper sources of information about soil variability, either to assist in the planning of a soil survey or to use in the process of kriging itself. Many authors have identified sources and strategies for the production of this information.

21.5 Variograms from the Precision Agriculture Literature

Variograms calculated for the same soil attribute may be a useful source of information. However, variograms calculated for the same property can vary significantly for a number of reasons that should be carefully considered. As outlined in Chap. 12, parent material, soil type, land use and climate will all have an effect on soil spatial variability. Consideration of these factors will be important in the selection of a variogram. Unfortunately, knowledge about soil spatial variability does not extend to the quantification of which of these factors are most important for determining spatial variability of different soil attributes.

The underlying variability of soil attributes might be different for the reasons mentioned above. In addition, the methods used to detect soil variability might create differences in the shape of the variogram. Different projects may focus on detecting variability at different scales for management or budget reasons. Even if the soil type is similar between two studies, the spatial variability might not be measured in a way that provides useful information.

It is also very important that the statistical methods used are assessed critically, before results are used or duplicated. There are a number of variograms included in Tables 21.1, 21.2, 21.3, 21.4, 21.5, 21.6, 21.7 and 21.8 which appear to use insufficient sample numbers for variogram estimation. There are several variograms which appear to assign spatial structure at a magnitude that appears to be meaningless relative to the units (i.e. total sill of less than 1% for the soil texture fraction).¹ These results have been included for completeness, but we wish to draw the readers' attention to the fact that some of the variograms included in this collection may have issues associated with them.

We describe below some key trends we have noted in the compilation of soil variograms and include a graphical summary of the variograms we have compiled. Variograms from McBratney and Pringle (1999) are included as well as those we have collected from the intervening period. Summary details and references for each variogram are given in Tables 21.1, 21.2, 21.3, 21.4, 21.5, 21.6, 21.7 and 21.8. We encourage the reader to consult each source directly for more detail about the sampling design and process.

Variograms for the same soil attribute from existing literature can be a useful source of information about field scale soil. It is important to exercise caution when consulting this literature, as variograms have been created from different soil types, for different purposes and possibly with important methodological limitations.

21.5.1 Field Scale Soil Variograms: Key Trends

21.5.1.1 Variogram Forms

Across all soil properties, a number of functional forms were fit to variograms. For each soil property, at least one study found no spatial structure (i.e. pure nugget) to be the best fit. This suggests that either the spatial structure occurs at scales finer than those surveyed or that the variability in the property in question is less than the experimental error. Spherical and exponential variograms were commonly used. Some papers described the functional form as 'experimental'. These models were fit with an exponential model.

¹In some cases the magnitude of the nugget and partial sill is extremely small compared to the magnitude of the standard deviation. This may suggest that the data has been transformed in some way before the variogram has been fit. We have reported the results as in the original article. These results should be interpreted with particular caution.

Table 21.1 Compilation of key properties for clay variograms

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
a	Lopez-Granados et al. (2002) (site two)	Caracol, Southern Spain		Vertic Xerochep	Crop	Before sowing	Boyoucos densimeter	6	84	26.3	12	0–0.10	Exponential	61	84	29.9
b	Adderley et al. (1997)	Nigeria			Trees			30	480	25.8	9.1	0–0.10	Experimental	30	110	250
c	Lopez-Granados et al. (2002) (site one)	Monclova, Southern Spain		Alfisol	Crop	Before sowing	Boyoucos densimeter	11.2	80	19.6	10.8	0–0.10	Nugget	113	0	0
d	Oliver and Webster (1987), (10–15)	Wyre Forest, England		Acid brown loam	Forest			0.025	100			0–0.15	Exponential	24.9	50.7	51.6
e	Oliver and Webster (1987), (0–5)	Wyre Forest, England		Acid brown loam	Forest			0.025	100	24.4	10.8	0–5	Exponential	14.34	51.8	65.7
f	Kerry and Oliver (2007)							15	118	47.5	6.2		Spherical	0.72	36.72	71.48
g	Nolin et al. (1996)	Quebec		Aquepts	Crop			10	130	44.3	6	0–15	Spherical	10.8	20.9	76
h	Shatar (1996)	Moree, Australia		Vertisol	Crop			15	114	49	5.5	15–30	Spherical	17	13.5	310
i	Kerry and Oliver (2007)							44	294	17.9	7.3		Spherical	4.19	22.9	152
j	Faroque et al. (2012) (site two)	Nova Scotia, Canada		Sandy loam, orthic humo-ferric podzols	Orchard		Boyoucos densimeter	1.6	86	9.6	4	0–15	Spherical	8.08	17.91	65.5
k	Miller et al. (1988)	Sacramento, California		Xererts, Xerochrepts	Crop			8	99	37.1	4.6	Surface	Spherical	7	14	75
l	Shouse et al. (1990)	Texas, USA		Vertisol				0.3	203	40	3.9	0–30	Exponential	11.7	4.8	8.47

(continued)

Table 21.1 (continued)

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget sill	Partial sill	Range
m	Kerry and Oliver (2007)	United Kingdom						10.5	205	30.2	5.8	0–15	Spherical	3.12	11.99	99.86
n	Liu et al. (2008)	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Clay loam to loam	Tobacco plantation	Before sowing	Bouyoucos densimeter	87	81	41.7	3.4		Spherical	2.72	11.78	609
o	Williams (1987) (four directions average)	Oklahoma		Paleustolls	Crop			1.62	108	22	2.3	0–30	Experimental	8.58	4.88	52.5
p	Shouse et al. (1990), (field 1)	Texas, USA		Vertisol				0.3	182	51	3.1	0–30	Exponential	3.8	6.4	11.89
q	Shukla et al. (2004)	Gross-Enzersdorf, Austria	510 mm, 10C	Loam to sandy loam	Crop	May	Pipette method	6.25	60	46.8	4.6	0–15	Spherical	5.24	4.27	347
r	Shouse et al. (1990)	Texas, USA		Vertisol				0.3	205	46.1	2.6	0–30	Spherical	3.5	3.1	36.58
s	Panagopoulos and Antunes (2008)	Algarve region, south Portugal						400	81	32.7	7	0–20	Gaussian	4.83	5.77	1.997.8
t	Kristensen et al. (1995) Riso	Riso, Denmark		Sandy loam/sandy clay loam	Crop			10.9	270			0–25	Exponential	0	6	37
u	Kristensen et al. (1995) (Vindum)	Vindum, Denmark		Sandy loam	Crop			10	302			0–25	Exponential	0	4.8	450
v	Kilic et al. (2012), (5 years of cultivation)	Kaz Lake of Tokat, Turkey	436 mm, 12C	Typic ustifluent Clay loam to sandy clay loam	Crop/horticulture		Bouyoucos hydrometer	0.8	46	21.7	6.3		Spherical	0.65	1.41	310.9
w	Tabor et al. (1985)	Arizona		Haplargid/mollisol	Crop			13	49	32.1	5.8	0–20	Linear	1.68	60.32	500

x	Kilic et al. (2012)	Kaz Lake of Tokat, Turkey	436 mm, 12C	Typic ustifluent Clay loam to sandy clay loam	Marshy plants		0.8	46	29.8	7.4	0–20	Spherical	0.69	0.81	207.8
y	Ayoubi et al. (2007)	Golestan Province, Iran		Fine, mixed, thermic, fluventic haploxerepts	Crop	Before planting	1.8	101	56.3	1.1		Spherical	0.67	0.66	25.83
z	Kilic et al. (2012)	Kaz Lake of Tokat, Turkey	436 mm, 12C	Typic ustifluent Clay loam to sandy clay loam	Crop/horticulture		0.8	46	20.4	5.2		Linear	0.99	0	85.66
aa	De Souza (2010)- site two	Araras, Southeast Brazil	Wet summer, dry winter	Oxisol (typic haplustox)	Cropping	Fall of 2004	22	90	34	0.2	0–20	Spherical	0	0.68	340
ab	de Souza et al. (2010) – site one	Araras, Southeast Brazil	Wet summer, dry winter	Oxisol (typic haplustox)	Cropping	Fall of 2003	20	80	26.1	0.5	0–20	Spherical	0.04	0.42	393
ac	Kerry and Oliver (2007)						6.9	109	4.1	1		Spherical	0.06	0.39	96.17
ad	Molin et al. (2013)- field one	Sao Palo, Southeast Brazil		Typic dystrodoxes			19	42	21.5	5.1	0–20	Spherical	0	0.15	153
ae	Molin et al. (2013) – field two	Sao Palo, Southeast Brazil		Typic hapludoxes			22	92	23.1	3.7	0–20	Spherical	0.01	0.02	307.3
af	Faroque et al. (2012); site one	Central Nova Scotia, Canada	Sandy loam, ferric podzols				1	56	41.9	4.4	0–15	Linear	0.01	19;16	23.7

^aUnless otherwise specified mm refers to annual average rainfall, and C refers to average annual temperature

Table 21.2 Compilation of key properties for sand variograms

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
a	Adderly et al. (1997)	Nigeria			Trees			30	157	56.8	10	0–10	Experimental	10	190	250
b	Lopez-Granados et al. (2002) site two	Caracol, Southern Spain		Vertic Xerochep	Crop		Bouyoucos densimeter	6	84	33.4	13.4	0–10	Spherical	85	99	84.8
c	Farooque et al. (2012) – site two	Central Nova Scotia, Canada		Sandy loam, orthic humo-ferric podzols	Orchard			1.6	86	48.7	12.5	0–15	Spherical	68.8	106.4	67.9
d	Lopez-Granados et al. (2002) site one	Monclova, Southern Spain		Alfisol	Crop		Bouyoucos densimeter	11.2	80	57.4	8.5	0–10	Nugget	69	0	0
e	Liu et al. (2008)	Henan Province, Central China	Monsoon 14.6C, 680 mm	Clay loam to loam,	Tobacco plantation	Before sowing	Bouyoucos densimeter	87	81	43.8	4.8	0–20	Spherical	7.7	32.43	657
f	Shatar (1996)	Moree, Australia		Vertisol	Crop			15	114			15–30	Spherical	20	17	300
g	Miller et al. (1988)	Sacramento, California		Xererts, Xerochrepts	Crop			8	99	20.2	5.9	Surface	Spherical	0.6	33.4	75
h	Farooque et al. (2012): site one	Central Nova Scotia, Canada		Sandy loam, orthic humo-ferric podzols	Orchard			1	56	49.5	4.5	0–15	Linear	18.75	18.75	85.86
i	Williams et al. (1987), four directions averaged	Oklahoma, USA		Paleustolls	Crop			1.62	108	27	4.8	0–30	Experimental	8	9.25	62.5

j	Tabor et al. (1985)	Arizona, USA		Haplargid/mollisol	Crop			13	49	41.7	8.4	0–20	Linear	15.4	101.4	300
k	Shouse et al. (1990)	Texas, USA		Vertisol				0.3	205	21.1	3.3	0–30	Spherical	6.2	5.2	22.86
l	Campbell (1977)	Kansas, USA						1.6	160	8.5	2.9		Experimental	2.2	0.8	40
m	Kilic et al. (2012)	Kaz Lake of Tokat, Turkey	436 mm, 12C	Typic ustifluent clay loam to sandy clay loam	Crop/horticulture			0.8	46	25.5	9.7	0–20	Exponential	0.75	1.13	200.4
n	Kilic et al. (2012) – 5 years of cultivation	Kaz Lake of Tokat, Turkey	436 mm, 12C	Typic ustifluent Clay loam to sandy clay loam	Crop/horticulture			0.8	46	33.5	10.3		Exponential	0.87	0.87	306
o	Kilic et al. (2012) – cultivation	Kaz Lake of Tokat, Turkey	436 mm, 12C	Typic ustifluent Clay loam to sandy clay loam	Crop/horticulture			0.8	46	50.8	7.9		Linear	1.02	0	85.66
p	Campbell (1977)	Kansas, USA						1.6	160	1.2	0.3		Experimental	0.04	0.04	30
q	Lady-smith Ayoubi et al. (2007)	Golestan Province, Iran		Fine, mixed, thermic, fluventic haploxerepts	Crop	Before planting		1.8	101	2.2	0.1	0–30	Gaussian	0.01	0	91.41

Table 21.3 Compilation of key properties for pH variograms

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
a	Kilic et al. (2012) (unmodified)	Kaz Lake of Tokat, Turkey	Typic ustifluent	Crop		Water 1:2.5	0.8	46	7.9	0.1	0–20	Exponential	0.8	0.82	209.6
b	Sidorova et al. (2012)	Northwestern Russia		Crop	Before sowing	KCl		72	5.6	0.2	23	Spherical	0.08	1	93.1
c	Kilic et al. (2012) (20 year)	Kaz Lake of Tokat, Turkey	Typic ustifluent.	Crop		Water 1:2.5	0.8	46	8.1	0.2	0–20	Linear	1.03	0	85.66
d	Kilic et al. (2012) (5 year)	Kaz Lake of Tokat, Turkey	Typic ustifluent	Crop		Water 1:2.5	0.8	46	7.8	0.2	0–20	Linear	1.01	0	85.66
e	Thompson et al. (2004) – site one	Alabama, USA		Crop			0.4	71			11*	Spherical	0.05	0.71	186.2
f	Pierce et al. (1995) (Durand)	Durand, USA	Alfisol, fine loam	Crop			16	165	6.6	0.9	0–5	Spherical	0.13	0.47	354
g	Machado et al. (2007)	Uberlandia, Brazil	Red latosol, moderate clayey texture	Crop		Water: 1:1	25	121	6	0.4	0–20	Spherical	0.07	0.51	150
h	Thompson et al. (2004) (site three)	Alabama, USA		Crop			0.4	48			11	Spherical	0.11	0.43	168.9
i	Mulla (1993)	Washington, USA	Ultic haploxeroll	Crop			8	172	6.1	0.7	0–30	Spherical	0.17	0.26	132
j	Pierce et al. (1995) (Adrian)	Adrian, USA	Alfisol, loam	Crop			10	74	6.5	0.9	0–5	Spherical	0.08	0.32	95
k	Camacho-Tamayo et al. (2007) (site one)	Puerto Lopez, Colombia	Typic haplustox	Crop	May 04	Water (1:1) potentiometer	1.875	42	5.1	0.4	0–10	Exponential	0.09	0.31	410
l	Webster and McBratney (1987)	Suffolk, England		Crop		Water	77	436	7.7	0.6		Spherical	0.02	0.33	185

m	Uehara et al. (1985)	Situng, Indonesia							0.0784	137							Experimental	0	0.35	4
n	de Souza et al. (2010) (site one)	Araras, Southeast Brazil	Wet summer, dry winter						20	80	Fall of 2003			4	5.6	0–20	Exponential	0.07	0.23	148
o	Adderley et al. (1997)	Nigeria							30	480				0.5	6.3	0–10	Exponential	0.15	0.1	250
p	de Souza et al. (2010) (site two)	Araras, Southeast Brazil	Wet summer, dry winter						22	90	Fall of 2004			0.5	5.7	0–20	Spherical	0.01	0.23	97
q	Pierce et al. (1995) (Plainwell)	Plainwell, USA							22	174				0.4	6.7	0–20	Spherical	0.06	0.15	190
r	Thompson et al. (2004) (site two A)	Alabama, USA							0.2	124						22	Spherical	0.09	0.1	393.2
s	Birrell et al. (1996)	Missouri, USA							28	504						0–15	Spherical	0.06	0.11	125
t	Thompson et al. (2004) (site two B)	Alabama, USA							0.4	58						22	Spherical	0.05	0.05	181.1
u	Kristensen et al. (1995) (Riso)	Riso, Denmark							10.9	270						0–25	Exponential	0	0.1	17
v	Kristensen et al. (1995) (Vindum)	Vindum, Denmark							10	302						0–25	Exponential	0	0.09	19
w	Liu et al. (2008)	Henan Province, Central China	Monsoonal 14.6C, 680 mm						87	81	Before sowing	Water (1:2.5)		0.3	7.7	0–20	Spherical	0.01	0.06	308
x	Mondo et al. (2012)	Sao Paulo, Brazil	Not specified						22	33	After			0.2	5	0–20	Gaussian	0.02	0.05	650

(continued)

Table 21.3 (continued)

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
y	Laslett et al. (1987)	Brisbane, Australia			Pasture		CaCl ₂	1	121	4.5	0.2		Spherical	0.02	0.03	55
z	Campbell (1977) (Pawnee)	Pawnee, Kansas, USA					CaCl ₂ (1:2)	1.6	160	6.5	0.3		Experimental	0.04	0	0
aa	Silva et al. (2003)	Santa Maria, Brazil		Ultisol dystrophic hapludalf	Crop	Before sowing	Water 1:1	0.3	192	4.9	0.1	0–20	Spherical	0	0.02	18.66
ab	Campbell (1977) (Ladysmith)	Ladysmith, Kansas, USA					CaCl ₂ (1:2)	1.6	160	6.5	0.2		Experimental	0.02	0	0
ac	Shatar (1996)	Moree, Australia		Vertisol	Crop		CaCl ₂ (1:5)	15	114	7.4	0.2	5:30	Spherical	0.01	0.01	310
ad	Tabor et al. (1985)	Arizona		Haplagid/Mollisol	Crop			13	49	7.3	0.2	0–20	Linear	0.02	0.06	500
ae	Lopez-Granados et al. (2002) (site one)	Monclova, Southern Spain		Alfisol	Crop	Before sowing	0.1 mol KCl	11.2	80	7.8	0.1	0–100	Spherical	0	0.01	66
af	Lopez-Granados et al. (2002) (site two)	Caracol, Southern Spain		Vertic Xerochep	Crop	Before sowing	0.1 mol KCl	6	84	7.7	0.1	0–100	Nugget	0.01	0	0
ag	Shukla et al. (2004)	Gross-Enzersdorf, Austria	510 mm 10C	Loam to sandy loam	Crop	May		6.25	60	NA	NA	0–15	Spherical	0.01	0	158
ah	Ayoubi et al. (2007)	Golestan Province, Iran		Fluventic haploxerepts	Crop	Before sowing	0.1 mol KCl	1.8	101	7.9	0	0–30	Spherical	0.001	0.001	24.39

*Assuming 11 cm were taken from topsoil.

Table 21.4 Compilation of key properties for carbon variograms

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
a	Farooque et al. (2012) (site one -1)	Central Nova Scotia, Canada	Sandy loam, podzols	Orchard	May July 2009	Om (LOI)	1	56	6.6	1.5	0-15	Sph.	1.14		
b	Farooque et al. (2012) (site one -2)	Central Nova Scotia, Canada	Sandy loam, podzols	Orchard	Jun 10	Om (LOI)	1	56	6.6	1.4	0-15	Sph.	1.11	2.28	76.1
c	Panagopoulos and Antunes (2008)	Algarve region of south Portugal	Mostly lithosols	Forestry, crop		OM (WB-wet)	400	81	1.2	0.5	0-20	Exp.	1.63	2.14	70.23
d	Farooque et al. (2012) (site two -2)	Central Nova Scotia, Canada	Sandy loam, podzols	Orchard	Jun 11	Om (LOI)	1	56	24	5.7	0-16		0.25	1.95	1997.8
e	Farooque et al. (2012) (site two -1)	Central Nova Scotia, Canada	Sandy loam, podzols	Orchard	May July 2009	Om (LOI)	1.6	86	4.9	1.3	0-15	Exp.	0.22	1.86	13.7
f	Kilic et al. (2012) (5 years)	Kaz Lake of Tokat, Turkey	Typic ustifluent	Crop		OM (WB-wet)	0.8	46	1	0.3	0-20	Exp.	0.09	1.82	14.6
g	Nolin et al. (1996)	Quebec, Canada	Aquepts	Crop		OC	10	130	2.8	0.7	0-15	Sph.	0.08	1.23	224.5
h	Kristensen et al. (1995), (Riso)	Riso, Denmark	Sandy loam-sandy clay loam	Crop		OC	10.9	270			0-25	Exp.	0	0.61	343
i	Kilic et al. (2012) (10 years)	Kaz Lake of Tokat, Turkey	Typic ustifluent	Crop		OM (WB-wet)	0.8	46	0.7	0.3	0-20	Exp.	0	0.43	99
j	Kilic et al. (2012) (uncultivated)	Kaz Lake of Tokat, Turkey	Typic ustifluent	Crop		OM (WB-wet)	0.8	46	1.6	0.4	0-20	Lin.	0.33	0.35	9.6
k	Mulla (1993)	St. John, Washington, USA	Ultic haploxeroll	Crop		OC	8	172	1.2	0.5	0-30	Sph.	0.05	0	85.66

(continued)

Table 21.4 (continued)

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
l	Gutierrez et al. (2010)	Municipality of Pasca, Colombia	1,800 mm, 16°C	Entisols and others	Horticulture		OC	1.5	64	6.4	0.3	0-20	Sph.	0.17	0.19	114
m	Nanni et al. (2011)	Sao Paulo state, Brazil	Meso-thermic climate	Oxisols, entisols, alfisols, ultisols, inceptisols and mollisols	Crop		OM ^b	184	184	0.7	0.4	0-20	Sph.	0.05	0.24	10,240
n	Zanão Júnior et al. (2010)	South-eastern Brazil	1,500 mm	Oxisol hapludox (medium clay)	Crop	May to June 2003	OM	25	121	1.7	0.2	0-10	Gaus	0.05	0.15	691
o	Cahn et al. (1994)	Central Illinois		Mollisol	Crop		OC	0.25	200	1.7	0.3	0-15	Exp.	0	0.15	539
p	Kristensen et al. (1995), (Vindum)	Vindum, Denmark		Sandy loam	Crop		OC	10	302			0-25	Exp.	0	0.18	50
q	Shukla et al. (2004)	Gross-Enzersdorf, Austria	Temperate and continental, 510 mm	Loam to sandy loam	Crop	May	OC	6.25	60	1.4	0.2	0-15	Sph.	0.04	0.12	45
r	Rowlands 1998; cited in McBratney & Pringle (1999) ^c	Wyalkatchem, WA	10C	Duplex	Crop		OC	75	56	0.7	0.2	0-10	Sph.	0.03	0.03	163
s	de Souza et al. (2010) (site two)	Araçás, Southeast Brazil	Wet summer, dry winter	Oxisol (typic haplustox)	Crop	Fall of 2004	OM	22	90	15.1	9.7	0-20	Sph.	0.02	0.03	638,985
t	Mondo et al. (2012)	Sao Paulo, Brazil			Crop	After harvest	OM	22	33	0.7	0.2	0-20	Gaus	0.01	0.04	304
u	Camacho-Tamayo et al. (2008) (site I)	Puerto Lopez, Colombia	2,375 mm; 27C	Typic haplustox	Crop	May 04	OC (mod-WB)	1.9	42	1.3	0.2	0-10	Exp.	0	0.05	498.2

v	Shukla et al. (2004)	Gross-Enzersdorf, Austria	Temperate and continental, 510 mm 10C	Loam to sandy loam	Crop	May	OC	6.25	60	1	0.2	0–15	Sph.	0.02	0.04	22.1
w	Shatar (1996)	Moree, Australia		Vertisol	Crop		OC	15	114	1.1	0.2	15–30	Sph.	0.01	0.02	184
x	Zhang et al. 2016	Jiangsu Province, China			Crop	2012	OM	7	136				Sph.	0	0.03	280
y	de Souza et al. 2010 (site one)	Araras, Southeast Brazil	Wet summer, dry winter	Oxisol (typic haplustox)	Crop	Fall of 2003	OM	20	80	10.7	1.4	0–20	Sph.	0.01	0.04	38.6
z	Amirnejad et al. 2011	Uttar Pradesh, India		Inceptisol	Crop		OC (WB)	19.6	145			0–15	Gaus	0.01	0.02	324
aa	Kumhalova et al. 2011 (sampled 2005)	Prague, Ruzyně, Czech Republic	526 mm; 7.9C	Haplic luvisol.	Crop		OC	11.5	70	2.1	0.2		Sph.	0.02	0.02	380.73
ab	Carnacho -Tamayo et al. 2008 (site two)	Puerto Lopez, Colombia	2,375 mm, 27 C	Typic hapludox	Crop	Aug 04	OC (mod-WB)	1.875	42	1.6	0.1	0–10	Sph.	0	0.01	240.5
ac	Liu et al. 2010	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Sandy to medium clay loam	Crop	After harvest,	OM (WB-wet)	4	111	1	0.1	0–20	Sph.	0.01	0.02	33.5
ad	Liu et al. 2008	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Clay loam to loam	Crop	Before sowing	OM ^d	87	81	0.7	0.1	0–20	Sph.	0.01	0.01	56.5
ae	Ayoubi et al. 2007	Golestan Province, Iran		Fluventic haploxepts	Crop	Before planting	Om (WB)	1.8	101	1.5	0.1	0–30	Sph.	0.01	0.01	556
af	Silva et al. 2003	Santa Maria, Brazil		Ultisol dystrophic hapludalf	Crop	Before sowing	OM ^e	0.3	195	1.6	0.1	0–20	Gaus	0.01	0.01	29.28

(continued)

Table 21.4 (continued)

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
ag Kumhalova et al. 2011 (sampled 2004)	Prague, Ruzyně, Czech Republic	526 mm: 7.9C	Haplic luvisol	Crop		OC	11.5	70	1.7	0.2		Sph.	0.01	0.01	9.5
ah Kumhalova et al. 2011 (sampled 2006)	Prague, Ruzyně, Czech Republic	526 mm: 7.9C	Haplic Luvisol	Crop		OC	11.5	70	1.9	0.2		Sph.	0.01	0.01	274.6
ai Goovaerts and Chiang 1993 (average)	Belgium		Typic hapludalf			OC	0.16	73	0.7	0.1	0–20	Sph.	0.01	0.05	244.7
aj Chung et al. 2008	Korea	12.7C, 1,560 ^µ	Coarse, loamy, mixed, non-acid, mesic	Crop	After harvest	OM ^f	0.3	246	1.3	0.1	0–15	Lin.	0.01	0	11.5
ak Lopez-Granados et al. (2002) site one	Monclova, Southern Spain		Alfisol	Crop	Pre planting	OM ^g	11.2	80	0.9	0.1	0–10	Nug.	0.01	0	NA
al Lopez-Granados et al. (2002) site two	Caracol, Southern Spain		Vertic Xerochep	Crop	Pre planting	OM ^g	6	84	0.8	0.1	0–10	Sph.	0	0	0
am Chatterjee et al. 2015	West Bengal, India	1,443 mm, hot and humid		Crop	After harvest	OC (WB)	81	100	0.5	0.1	0–15	Lin.	0	0.01	44.8
an Kumhalova et al. 2011 (sampled 2007)	Prague, Ruzyně, Czech Republic	526 mm: 7.9C	Haplic luvisol	Crop	Not specified	OC	11.5	70	2	0.2		Sph.	0	0.01	58
ao Bai and Wang 2011	Shaanxi Province, China	393 mm	Silty loams	Orchard		OC ^h	0.275	125	0.3	0.1	0–10	Exp.	0	0.07	247.1

^aMeasurement method refers to the original property which was measured. All values have been converted into OC

^bOrganic matter (OM), total and effective acidity (determined by 1 M calcium acetate – Ca(CH₃COO)₂ H₂O and 1 M KCl titulometric method, respectively),

^cUnpublished data provided by Pringle for the publication of McBratney & Pringle (1999)

^dPotassium bichromate titrimetric method

^ePhotocolorimetry

^fLaboratory analysis was completed by the Soil Management Division – NIAST, RDA

^gRedox-Electrode (Methrom Titroprocessor)

^hOil bath titration

Table 21.5 Compilation of key properties for available nitrogen variograms

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range	
a	Liu et al. (2010)	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Sandy to medium clay loam	Crop	After harvest	Alkaline hydrolysable N (AN)	4	111	75	14.3	0-20	Spherical	105.6	127.3	112.6
b	Liu et al. 2008	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Clay loam to loam	Crop	Before sowing	Alkalytic N (AN)	87	81	70.6	10.3	0-20	Linear	110.46	0	NA
c	Lopez-Granados et al. (2002) (site two)	Caracol, Southern Spain	Vertic Xerochep	Crop	Before winter sowing	Nitrate		6	84	23.2	10.2	0-10	Exponential	10	91	31.7
d	Chatterjee et al. (2015)	West Bengal, India	Hot and humid	Crop	After harvest	Available N		81		76.7	7.8	0-15	Spherical	17.38	45.26	43
e	Shire 1997; cited in McBratney & Pringle (1999) ^d	Wyalkatchem, Western Australia	Duplex	Crop		Nitrate		90	88	22.9	7.6	0-10	Spherical	22.14	36.86	281.432
f	Lopez-Granados et al. (2002) site one	Monclova, Southern Spain	Alfisol	Crop	Before winter sowing	Nitrate		11.2	80	7.1	5.3	0-10	Nugget	27.2	0	0
g	Cahn et al. (1994) 0.25 ha	Central Illinois	Mollisol	Crop		Nitrate		0.25	200	6.2	3.7	0-15	Experimental	7	3	5
h	Everett and Pierce (1996)	Michigan	Loamy sand, sandy loam	Crop		Nitrate		22.6	60	6.2	2.2	0-30	Spherical	3.05	3.7	167.05

(continued)

Table 21.5 (continued)

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
i	Tabor et al. (1985)	Arizona		Haplargid/mollisol	Crop		Nitrate	13	49	13.6	4.2	0–20	Linear	6.66	11.12	200
j	van Meirvenne and Hofman (1989)	Belgium		Udifluent	Horticulture		Nitrate	1	247	8.9	2.7	0–100	Experimental	2.25	4.25	25
k	Wade et al. (1996) (pasture)				Pasture		Nitrate		86	0.8	0.8	0–10	Experimental	0.44	0.57	120
l	Cambardella et al. (1994)	Iowa		Udic/argic mollisols	Crop		Nitrate	10	72	7.2	1.3	0–15	Spherical	0.14	0.2	201
m	Wade et al. (1996), (arable)	Warwickshire, England					Nitrate		81	0.7	0.5	0–10	Experimental	0	0.3	60

^aAlkaline hydrolysable N was measured using the alkaline hydrolysis diffusion method

^bNitrate determined by colorimetry in SKALAR

^cAvailable N content was determined by alkaline permanganate method as provided in Chatterjee et al. (2015)

^dUnpublished data provided by Pringle for the publication of McBratney & Pringle (1999)

Table 21.6 Compilation of key properties for total nitrogen variograms

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
a Ganawa et al. (2003)	Sawah Sempadan, Malaysia		Jawa, Teluk, Karang, Sempadan and Sedu	Crop	Before planting	^a	2,300	120	4140	870	0–20	Spherical	100,000	700,000	460
b Yana et al. (2000)	Takatsuki, Japan	15.8 °C, 1,240 mm	Clay loam		After transplanting	Dry combustion	0.5	91	3,100	449.5		Spherical	30,000	270,000	35.5
c Yana et al. (2000)	Takatsuki, Japan	15.8 °C, 1,240 mm		Crop	After harvest	Dry combustion	0.5	91	3,400	452.2		Spherical	80,000	200,000	19.5
d Zhang et al. (2016)	Jiangsu Province, China			Crop			7	136				Exponential	0	180,000	39,08
e Shukla et al. (2004)	Gross-Enzersdorf, Austria	510 mm, 10C	Loam to sandy loam	Crop	May	Kjeldahl method	6.25		1,328,6381		0–15	Spherical	106,576	31,746	243.6
f Nouri et al. (2010)	Fesaran village, Estfahan		Clay loam and loam	Horticulture	Before fertigation	Nitrate	87	60			0–30	Spherical	30,000	30,000	141.8
g Liu et al. (2008)	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Clay loam to loam	Crop	Before sowing	Kjeldahl method	87	81	730	100	0–20	Spherical	2,000	9,000	274
h Ayoubi et al. (2007)	Golestan Province, Iran		Fine, mixed, thermic, fluventic haploxerepts	Crop	Before sowing	Kjeldahl method	1.8	101	1,300	110	0–30	Gaussian	600	10,000	23.99
i Chung et al. (2008)	Korea	12.7C, 1,560 mm ^b	Coarse loamy	Crop	After harvest	^b	0.3	246	1,500	100		Exponential	0	0	4.5

^aSulfuric-salicylic acid digestion method

^bLaboratory analysis was completed by the Soil Management Division – NIAST, RDA

Table 21.7 Compilation of key properties for phosphorus variograms

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
a	McBratney et al. (1985)	Narrabri, NSW, Australia	Vertisol	Crop		P (NaHCO ₃ soluble)	1	386	129.6	58.8	0–7.5	Gaussian	152.85	7,489.55	109.64
b	Chatterjee et al. (2015)	Hot and humid 1443 mm		Crop/ horticulture	After harvest	Available P2O ₅	81		172.4	82.4	0–15	Spherical	3,700.68	3,925.85	283
c	Mondo et al. (2012)	Sao Paulo, Brazil		Crop	After harvest	P resin	22	33	239.6	59.6	0–20	Spherical	1,484.69	2,030.1	89.9
d	Cambardella et al. (1994) (pothole field)	Iowa, USA	Udic/aquic mollisols	Crop		P (bray no. 1)	6.25	241	126	55.6	0–15	Spherical	596.9	2,839	71
e	Delcourt et al. (1996)	Leeftdaal, Belgium	Silty	Crop		P (colorimetric)					0–25	Gaussian	310	850	74.5
f	Cahn et al. (1994), 0.25 ha	Central Illinois, USA	Mollisol	Crop		P (bray no. 1)	0.25	200	74	26.8	0–15	Experimental	390	760	50
g	Pierce et al. (1995), (Plainwell)	Plainwell, USA	Entisol, loamy sand	Crop		P (not specified)	22	174	124	32	0–20	Spherical	233	844	172
h	Camacho -Tamayo et al. (2008) (site two)	Puerto Lopez, Colombia	Typic hapludox	Crop	Aug 04	P (bray II)	1.875	42	28.8	25.8	0–10	Spherical	2	660.5	38.4
i	Ganawa et al. (2003)	Sawah Sempadan, Malaysia	law, Teluk, Karang, Sempadan and Sedu	Crop	Before planting	Available P NH ₄	2,300	120	50.2	22.5	0–20	Spherical	69	422	597
j	Nolin et al. (1996)	Quebec, Canada	Aquepts	Crop		P (Melich III)	10	130	52	18.5	0–15	Exponential	35.1	255.5	13

k	Chung et al. (2008)	Korea			Crop	After harvest	c	0.3	246	119	15.5		Spherical	17.8	210.3	4
l	Pierce et al. (1995) Durand	Durand, USA			Crop		P (not specified)	16	165	35	21	5:20	Linear	223	163	294
m	Camacho-Lamayo et al. (2008) (site one)	Puerto Lopez, Colombia	2,375 mm	27 C	Crop	May-04	P (bray II)	1.875	42	11.1	14.2	0-10	Linear	154.4	262.9	373.4
n	Frogbrook et al. (2002) (sampled 1998)					Before fertiliation		16.5					Exponential	17.95	131.5	17.07
o	Frogbrook et al. (2002) (sampled 1999)							16.5					Spherical	54.58	66.61	54.9
p	Frogbrook et al. (2002) (sampled 1997)							16.5					Spherical	56	59.72	49.95
q	Kristensen et al. (1995) (Riso)	Riso, Denmark			Crop		Available P	10.9	270			0-25	Exponential	0	110	60
r	Shatar (1996)	Moree, Australia			Vertisol		Available P	15	114	18	1	15-30	Spherical	55	50	290
s	Sidorova et al. (2012)	Northwestern Russia			Crop		Available P ^d		72	484.9	65.6		Spherical	11	93	63.7
t	Pierce et al. (1995) (Adrian)	Adrian, USA			Alfisol, loam		P (not specified)	10	74	23	9	0-5	Exponential	23.6	60.49	68
u	Machado et al. (2007)	Uberlandia, Brazil	Tropical humid with dry season,		Crop			25		12.4	9.1	0-20	Linear	82.33	0	NA
v	Mulla (1993)	St. John, Washington, USA			Crop		P (acetate extractable)	8	172	15.2	7.7	0-30	Spherical	27.58	35.8	145

(continued)

Table 21.7 (continued)

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
w	Rowlands 1998; cited in McBratney & Pringle (1999) ^e	Wyalkatchem, WA, Australia		Duplex	Crop		P (not specified)	75	56	31.9	7	0–10	Spherical	22.32	28.72	414.604
x	Kristensen et al. (1995) (Vindum)	Vindum, Denmark		Sandy loam	Crop		Available P	10	302			0–25	Exponential	0	45	148
y	Liu et al. (2008)	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Clay loam to loam,	Crop	Before sowing,	Available P ^f	87	81	16.1	5.9	0–20	Spherical	13.4	22.59	337
z	Lopez-Granados et al. (2002) (site two)	Southern Spain (Caracol)		Vertic Xerochep	Crop	Before crop	Available P ^g	6	84	11.3	5.6	0–10	Exponential	0.7	33.3	28.3
aa	Lopez-Granados et al. (2002) (site one)	Southern Spain (Monclova)		Alfisol	Crop	Before r crop	Bray extractable P ^e	11.2	80	15.5	4.4	0–10	Exponential	0.6	18.6	27.4
ab	Ayoubi et al. (2007)	Golestan Province, Iran		Fine, mixed, thermic, fluventic haplocrepts	Crop	Before crop	Available P ^h	1.8	101	27.2	1.3	0–30	Spherical	1.08	0.57	35.58

^aAvailable P205 Olsen method by way of extracting 2.5 g of soil with 50 ml of 0.5MNaHCO₃ (pH 8.5) for 30 min and determining the phosphorus in the extract by the L-ascorbic acid method

^bAvailable P NH₄F-HCl extraction method

^cLaboratory analysis was completed by the Soil Management Division – NIAST, RDA

^dAvailable P (the TsI-NAO modification of the Kirsanov method)

^eUnpublished data provided by Pringle for the publication of McBratney & Pringle (1999)

^fAvailable P Olsen extraction method using alkaline sodium bicarbonate as the extractant in a 20:1 ratio

^gBray extractable P measured by colorimetry using ascorbic acid-ammonium molybdate reagents

^hAvailable P measured by colorimetry using ascorbic acid-ammonium molybdate reagents

Table 21.8 Compilation of key properties for potassium variograms

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
a Lopez-Granados et al. (2002) (site two)	Caracol, Southern Spain		Vertic Xerochep	Crop	Before planting	Exchangeable K	6	84	741	156	0-0.10	Spherical	0	15,210	54.6
b Lopez-Granados et al. (2002) (site one)	Monclova, Southern Spain		Alfisol	Crop	Before planting	Exchangeable K	11.2	80	468	117	0-0.10	Exponential	0	15,210	34.5
c Cahn et al. (1994) (0.25 ha)	Central Illinois, USA		Mollisol	Crop		K (not specified)	0.25	200	268.2	114.9	0-15	Experimental	5000	7500	50
d Pierce et al. (1995) (Adrian)	Adrian, USA		Alfisol, loam	Crop		K (not specified)	10	74	210	71	0-5	Exponential	1850	4168	93
e Frogbrook et al. (2002) (sampled 1997)			Clay loam		Before fertilisation		16.5	110				Spherical	994.5	1831	55.09
f Chatterjee et al. (2015)	West Bengal, India	Hot and humid, 1443 mm		Crop/ horticulture	After harvest	Available K ₂ O	81		134.8	51.9	0-15	Exponential	1864.51	923.4	55
g Chung et al. (2008)	Korea	12.7C, 1560 ^b	Coarse, loamy, mixed, non-acid, mesic	Crop	After harvest		0.3	246	269.1	27.3		Gaussian	304.2	2281.5	32.9
h Frogbrook et al. (2002) (sampled 1998)					Before fertilisation	K ^c	16.5					Circular	1121	1266	57.71

(continued)

Table 21.8 (continued)

	Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
i	Frogbrook et al. (2002) (sampled 1999)					Before fertilisation		16.5					Penta-spherical	1050	1034	79.03
j	Nanni et al. (2011)	Sao Paulo state, Brazil	Mesothermic climate	Oxisols, entisols, alfisols, ultisols, inceptisols and mollisols	Crop			184	184	61.3	44.6		Spherical	706.18	1373.56	353
k	Pierce et al. (1995) (Plainwell)	Plainwell, USA		Entisol, loamy sand	Crop		K (not specified)	22	174	121	35.9	0-20	Spherical	887	391	157
l	Ferraz et al. (2012) (sampled 2008)	Tres Pontas, Brazil	Mild tropical altitude	Red-yellow latosol	Crop		K Mehlich 1	22	48	7.1	6.5	0-20	Spherical	0	1268.37	437
m	Liu et al. (2010)	Henan Province, Central China	Monsoon climate 14.6C, 680 mm	Sandy to medium clay loam	Crop	After harvest	Available K ^d	4	111	161.4	30.8	0-20	Spherical	435	744.1	312.4
n	Silva et al. (2003)	Santa Maria, Brazil		Ultisol dystrophic hapludalf	Crop	Before sowing	Available K (flame photometry)	0.3	194	111	31.9	0-20	Gaussian	527	619	16.6
o	Pierce et al. (1995), (Durand)	Durand, USA		Alfisol, fine loam	Crop		K (not specified)	16	165	97	31	5-20	Spherical	302	833	174
p	Delcourt et al. (1996)	Leefdaal, Belgium		Silty	Crop		K (AES)			NA	NA	0-25	Spherical	425	565	149.7
q	Rowlands 1998; cited in McBratney & Pringle (1999) ^b	Wyalkatchem, WA		Duplex	Crop		K (not specified)	75	56	59.1	24.7	0-10	Spherical	185.17	581.23	800
r	Camacho-Tamayo et al. (2008) (site two)	Puerto Lopez, Colombia	2375 mm 27C	Typic hapludox	Crop	Aug 04	K ^e	1.875	42	74.1	15.6	0-10	Exponential	182.52	349.83	366.6

s	Machado et al. (2007)	Uberlandia, Brazil	Tropical humid with dry season	Red latosol, moderate clayey texture	Crop		K ^f	25	121	79	21.5	0-20	Linear	466.4	0	NA
t	Kristensen et al. (1995) (Riso)	Riso, Denmark		Sandy loam/sandy clay loam	Crop		Available K	10.9	270			0-25	Exponential	0	389	129
u	Kristensen et al. (1995) (Vindum)	Vindum, Denmark		Sandy loam	Crop		Available K	10	302			0-25	Exponential	0	377	25
v	Ferraz et al. (2012) (sampled 2007)	Tres Pontas, Brazil	Mild tropical altitude	Red-yellow latosol	Crop		K Mehlich 1	22	54	11.5	12.5	0-20	Exponential	0	365.82	165
w	Liu et al. (2008)	Henan Province, Central China	Monsoon climate 14-6C, 680 mm	Clay loam to loam, and the soil is slightly alkaline (pH, 7.7)	Crop	Before sowing	Available K (AK) ^d	87	81	105.9	14.3	0-20	Spherical	66.9	149.2	345
x	de Souza et al. (2010) (site one)	Araras, Southeast Brazil	Wet summer, dry winter	Oxisol (typic haplustox)	Crop	Fall of 2003		20	80	3.2	1.3	0-20	Spherical	26.1	103.9	185
y	Sidorova et al. (2012)	Northwestern Russia			Crop		Exchangeable K ^g		72	202.5	36.2		Exponential	7	106	72.8
z	Ayoubi et al. (2007)	Golestan Province, Iran		Fine, mixed, thermic, fluventic haplocrepts	Crop	Before sowing	Available K ^h	1.8	101	334.6	8.1	0-30	Spherical	17.16	62.4	93.92
aa	Camacho-Tamayo et al. (2008) (site 1)	Puerto Lopez, Colombia	2375 mm 27C	Typic haplustox	Crop	May 04	K ^e	1.875	42	70.2	7.8	0-10	Nugget	60.84	0	0

(continued)

Table 21.8 (continued)

Reference	Location	Climate	Soil type	Land use	Timing	Measurement method	Study area (ha)	Sample size	Mean	Standard deviation	Depth (cm)	Model	Nugget	Partial sill	Range
ab Thompson et al. (2004) (site two A)	Alabama, USA			Crop	January		0.2	124			22	Exponential	0.1	55.1	228.3
ac Thompson et al. (2004) (site two B)	Alabama, USA			Crop	January		0.4	58			22	Spherical	23.16	27.52	299
ad de Souza et al. (2010) – (site two)	Araras, Southeast Brazil	Wet summer, dry winter	Oxisol (typic haplustox)	Crop	Fall of 2004		22	90	3.7	1.9	0–20	Exponential	0.9	3.2	121
ae Thompson et al. (2004) (site one)	Alabama, USA			Crop	September		0.4	71			11	Spherical	0.01	0.06	295.2
af Thompson et al. (2004) (site three)	Alabama, USA			Crop	September		0.4	48			11	Spherical	0	0.03	160.4

^aExchangeable K atomic absorption spectrophotometry (AAS)

^bUnpublished data provided by Pringle for the publication of McBratney & Pringle (1999)

^cLaboratory analysis was completed by the Soil Management Division – NIAST, RDA

^dAvailable K (AK) neutral ammonium acetate extraction method

^eK extraction with ammonium acetate pH 7.0

^fHCl 0.05 mol L⁻¹ + H₂SO₄ 0.025 mol L⁻¹

^gExchangeable K Tsl-NAO modification of the Kirsanov method

^hAvailable K extraction with ammonium acetate (1 N)

21.5.1.2 Total Sill

Between studies the total sill (nugget plus partial sill) changes by several orders of magnitude for each property. As expected, this variability is the least pronounced for bounded properties (pH, OM %, sand % and clay %) and much more pronounced for micronutrients, which vary by around three to five orders of magnitude.

Variability in total sill is similar to that observed by McBratney and Pringle (1999). Inspection of the summary tables (Tables 21.1, 21.2, 21.3, 21.4, 21.5, 21.6, 21.7 and 21.8) indicates that for the majority of soil properties, the range in the total sill is similar for the variograms collected by McBratney and Pringle (1999) and the more recently collected properties. The maximum variability reached within 1 km has remained within an order of magnitude for all properties. For soil pH, organic carbon and potassium, the maximum variability found in the more recent literature search is two to three times greater than the maximum variability found in the literature reviewed by McBratney and Pringle (1999). The other properties have a very similar maximum.

21.5.1.3 Nugget

Like McBratney and Pringle (1999), we find wide variability (several orders of magnitude) in the nugget parameter. McBratney and Pringle (1999) suggest that this variability is largely due to the strong effect of sampling design on the nugget. A variogram can only model the spatial structure that is detectable by the sampling design. In general, the wider the spacing, the more spatial variability will be attributed to the nugget component of the model. If the survey spacing is wider than the spatial structure, the variogram will appear as a pure nugget model. A wider support and the use of aliquotting will reduce the 'noise' in the data and decrease the nugget. It has been often proposed, and is quite likely, that the majority of soil properties would have more than one layer of soil structure. The differences in estimated nugget likely reflect both the underlying differences in spatial structure at the field scale and the capacity of different survey designs to capture this variability.

21.5.1.4 Range

Like the lag and the sill, the range estimated by the theoretical variogram will be strongly affected by the sampling design. As the lag increases, the confidence interval around the variance increases (Oliver and Webster 2014). Precision agriculture surveys are typically conducted over areas from a few hectares to a few hundred hectares which limits the extent of spatial variability they can capture. Despite this, the majority of variograms fit from precision agriculture studies are bounded. The modelled range and sill can be thought of as the 'local sill' and 'local range' associated with the particular extent and spacing of that survey. It is very likely that if the extent of the survey was increased, another degree of variability would be

found. It is critical to consider the potential effect of both the extent and the spacing on the range when considering precision agriculture studies and how their findings may inform your own work.

21.5.2 Field Scale Soil Variograms: Methodological Differences

21.5.2.1 Survey Design

Another substantial difference between the survey designs was whether or not aliquotting was used. This is particularly significant when comparing these spatial studies because some studies model the range at distances that other studies were combining soil at. This is typically done for samples taken within 1 m of each other. This practice is likely to reduce the nugget (or white noise) and also to reduce any short-term spatial trends which might be occurring.

There is significant variation in the survey design which may influence the mapping of spatial variability. Nested designs are better placed to capture spatial trends across a variety of scales than designs with even spacing; however, because of the additional costs associated with these, they are less common.

21.5.2.2 Model Fitting Process

Oliver and Webster (2014) wrote an explanatory piece of work, describing the best methods for soil scientists to model variograms for kriging. They also described common mistakes made by soil scientists when calculating variograms. The majority of papers we assess do not follow all of Oliver and Webster's recommendations for reporting methods. This makes it difficult to assess how well a fitted variogram captures underlying spatial variability. Few papers report summary statistics for a variety of models, and few papers present variogram clouds to illustrate the utility of the fit. This does not necessarily mean that the fitted models are not accurate, but it does make it difficult to assess the model.

Another point worth considering is the possibility that trends are being overfitted. Perhaps some of the models presented in this chapter would have been better represented by a nugget. These issues around model quality are not new, but a degree of caution is required when interpreting the results.

The more recent literature has included studies which have found much lower values for the total sill for several soil attributes. For clay the lowest value found for the total sill is two orders of magnitude lower than the lowest value reported by Petit and McBratney. Sand is one order of magnitude lower. Some modelled variograms occur over a very tiny range of variability relative to the magnitude of the property they are measuring. Whether it is necessary or feasible to model a spatial structure of less than 1% for soil texture properties is questionable.

21.5.2.3 Measurement Methods

The properties we have included here are commonly measured soil properties with known agronomic implications. However, measurement of these properties is rarely simple or consistent. Differences in measurement methods and differences in which component of the property is being measured will influence both the shape and magnitude of the variogram.

pH is an extremely commonly measured property. However, within the studies we have assessed, there are differences in the solution, the dilution rate and the equipment used to measure the pH. This problem becomes more complex when considering more difficult-to-measure properties such as potassium and phosphorus. Different studies have used different extractants to target different fractions for these nutrients.

The variogram is affected by the distribution of the property it is being calculated for. The variograms calculated for potassium and phosphorus show the greatest differences in total sill. We expect that this is because the target of the measurements varies as well as the measurement method used.

Some articles were found which estimated total carbon or inorganic carbon. However, there were relatively few such studies, so we have not included them here. We have included studies which measured organic matter as a proxy for organic carbon. We converted these using the van Bemmelen factor (1.724). Pribyl (2010) illustrates that an accurate conversion factor for different soils can vary from 1.4 to 2.5. Error in the conversion will be small relative to the overall spread of the variograms.

21.5.3 Field Scale Soil Variograms: A Compilation

Figures 21.2, 21.3, 21.4, 21.5, 21.6, 21.7, 21.8, and 21.9, provide a visual compilation of field scale variograms for each of the soil properties initially examined by McBratney and Pringle (1999). We include both the original variograms used by McBratney and Pringle and variograms published since then. We only include variograms which were calculated from untransformed data and which were based on physical observations (i.e. not from remotely observed data). The black bold lines represent the average variogram (Sect. 21.7). Tables 21.1, 21.2, 21.3, 21.4, 21.5, 21.6, 21.7 and 21.8 correspond to each figure and include reference details and key parameters for each variogram. Because of the wide range of values of the source variograms, the scales used in the figures cannot include all of them, and some low sill variograms have not been included. We include the details in the tables for completeness, but suspect that they are unlikely to provide useful information. The figures affected and the number of source variograms not included are pH 1, carbon 3, total nitrogen 1 and potassium 2.

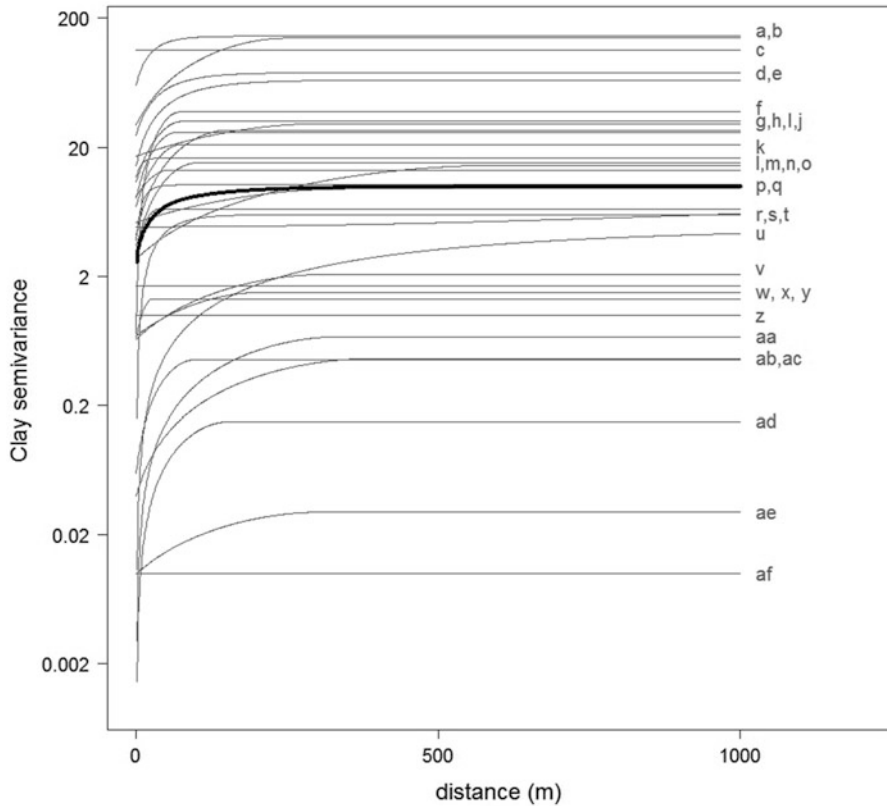


Fig. 21.2 Compilation of field scale variograms for clay. The *bold black line* represents the average variogram. Summary details and references for each variogram (*a–ag*) are given in Table 21.1

21.6 Estimation of the Variogram from Proportional Variograms

McBratney and Pringle (1999) observe a strong relationship between mean squared and variance for several soil properties and develop a method for estimating a ‘proportional variogram’. This method has the advantage of capturing the much higher levels of variability that tend to occur when the mean values of the property are extreme. Similar to McBratney and Pringle (1999), we find that some soil properties (phosphorus, nitrogen, potassium and carbon) appear to have a strong linear relationship between the mean and the standard deviation. This could imply that the calculation of a proportional variogram would be useful for these properties. However, closer interrogation reveals that this relationship is largely driven by high leverage points. It is not possible to fit a robust curve to link the mean and standard deviation. Proportional variograms are based on the relationship between the mean

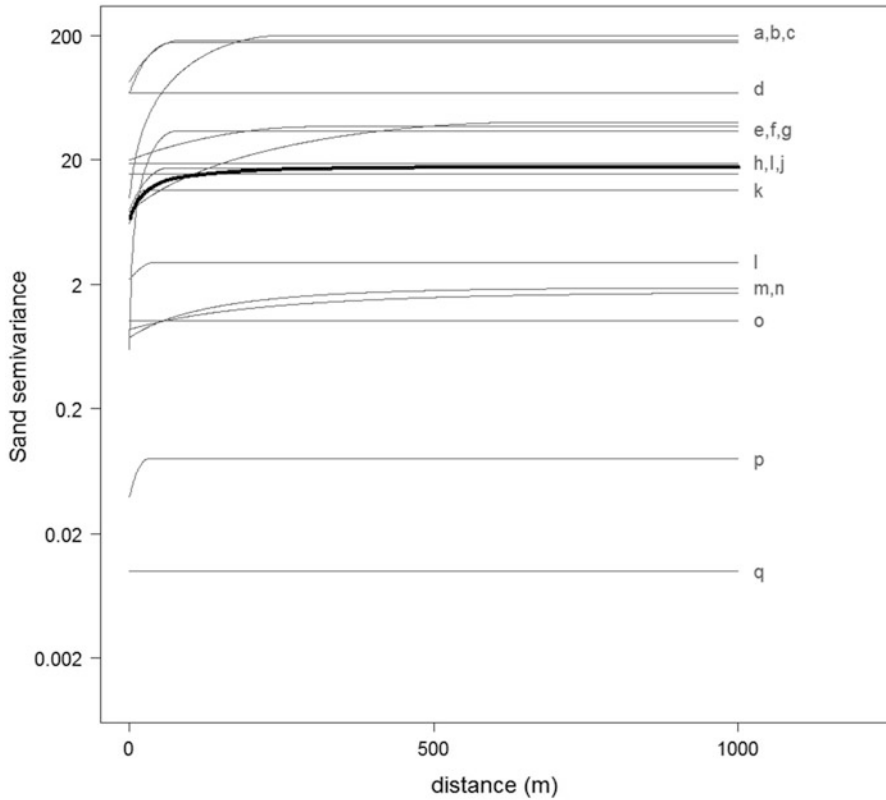


Fig. 21.3 Compilation of field scale variograms for sand. The *bold black line* represents the average variogram. Summary details and references for each variogram (*a–r*) are given in Table 21.2

and the variance. Because we cannot be confident about this relationship, it is not prudent to calculate proportional variograms.

We advise against the use of proportional variograms as an estimate for variability and as such do not update McBratney and Pringle's (1999) estimates of proportional variograms.

21.7 Estimation of the Variogram from Average Variograms

McBratney and Pringle (1999) calculate average variograms for seven soil properties. These average variograms are calculated from the sample of variograms they compile. The fourth root transform of each approximated by the spherical model is taken and then backtransformed. Exponential or spherical models are fitted.

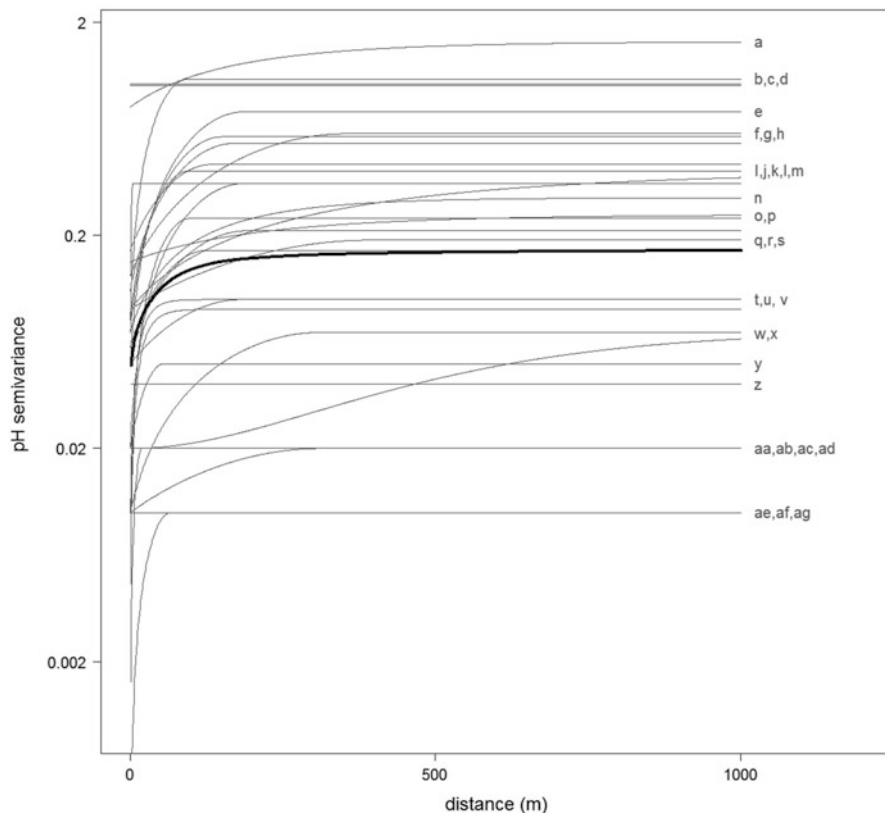


Fig. 21.4 Compilation of field scale variograms for pH. The **bold black line** represents the average variogram. Summary details and references for each variogram (*a–ag*) are given in Table 21.3

McBratney and Pringle (1999) note the broad spread of variability in the variograms they collected and suggest that this might make the ‘average’ variograms less useful. Despite this, they suggest that the average variogram is a useful starting point where no other information is available. Kerry and Oliver (2003) find evidence that average variograms can be useful for prediction when parent material and soil forming factors are similar, but emphasise that they do not expect a global average variogram to provide much useful information.

We do not believe a global average provides useful information for predictive purposes. However, as McBratney and Pringle (1999) suggested, the average variogram does provide a useful reference for those interested in soil variability. We produce average variograms for illustrative purposes (by averaging the fourth root transform of each variogram at finely spaced intervals and then plotting the backtransformed values), but we do not fit these with a functional form.

As suggested by McBratney and Pringle (1999) and illustrated by Kerry and Oliver (2004), the concept of the average variogram has the most use for prediction

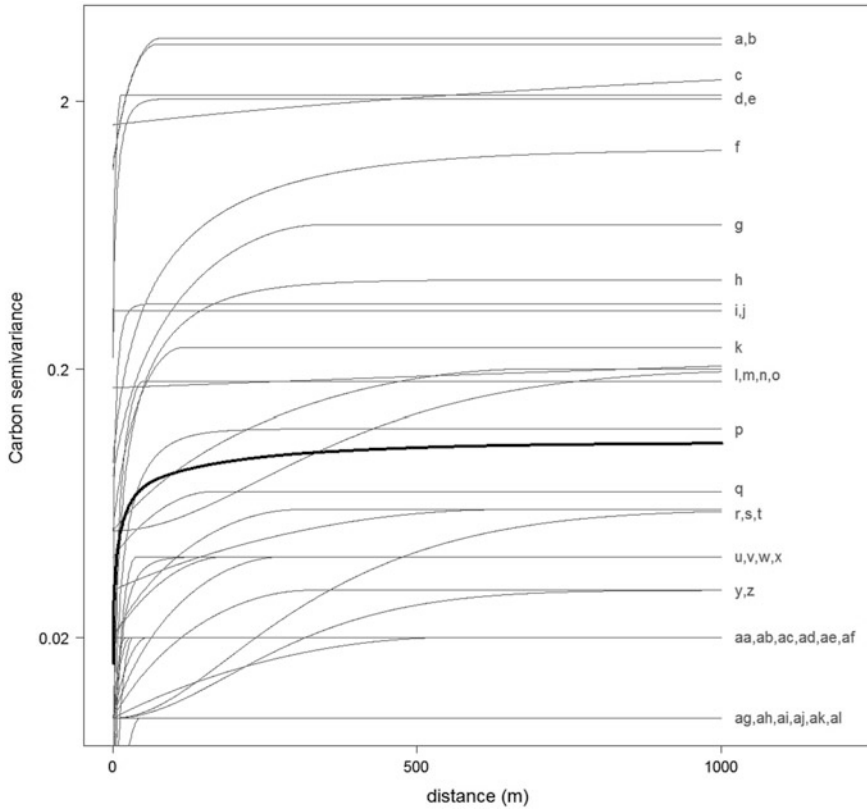


Fig. 21.5 Compilation of field scale variograms for carbon The *bold black line* represents the average variogram. Summary details and references for each variogram (*a–al*) are given in Table 21.4

when it is calculated from a select subset of existing variograms. The diversity of climate regimes, soil types and land use for which field scale variograms have been calculated means that discretion is essential in the selection process. Differences in sampling regime, soil measurement protocols and geostatistical methods add another layer of complexity that needs to be navigated in appropriate selection. We discuss these issues further in the next section.

We advise against the use of the ‘global average’ variogram as an estimate of local soil variability. Instead, where suitable variograms are available, an average of variograms with similar conditions is taken. Discretion and expert knowledge will need to be used in this selection process. The process outlined by McBratney and Pringle (1999) for calculation of an average variogram can be followed.

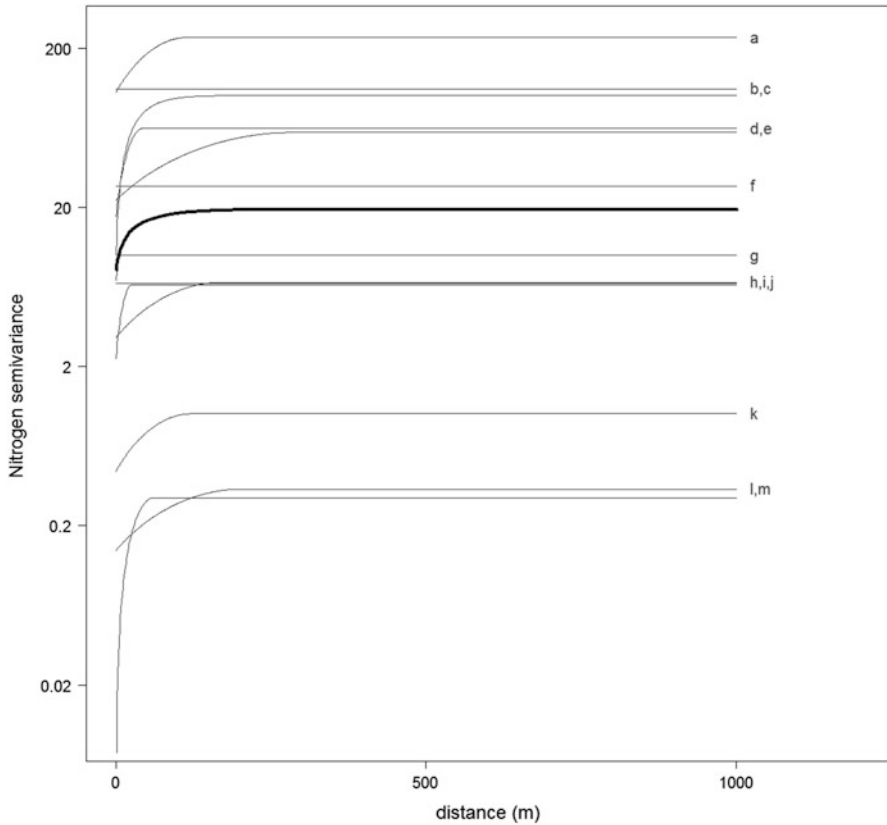


Fig. 21.6 Compilation of field scale variograms for available nitrogen. The *bold black line* represents the average variogram. Summary details and references for each variogram (*a–m*) are given in Table 21.5

21.8 Ancillary Information

Kerry and Oliver have written several papers investigating the potential of using cheaper more densely available ancillary information to supplement expensive and sometimes sparse soil survey data. In 2004, they compared the spatial structure of a number of ancillary data sources to the spatial structure of a number of fixed soil properties. They found that variograms calculated from aerial colour photographs of 3.4 m ‘sampling density’ can estimate range with sufficient accuracy to helpfully guide sampling density of the soil survey. Kerry and Oliver (2008) paper extends the use of ancillary information to kriging. They suggest that the primary requirement for using data is that the ancillary variogram has a similar sill-to-nugget ratio to the property being studied.

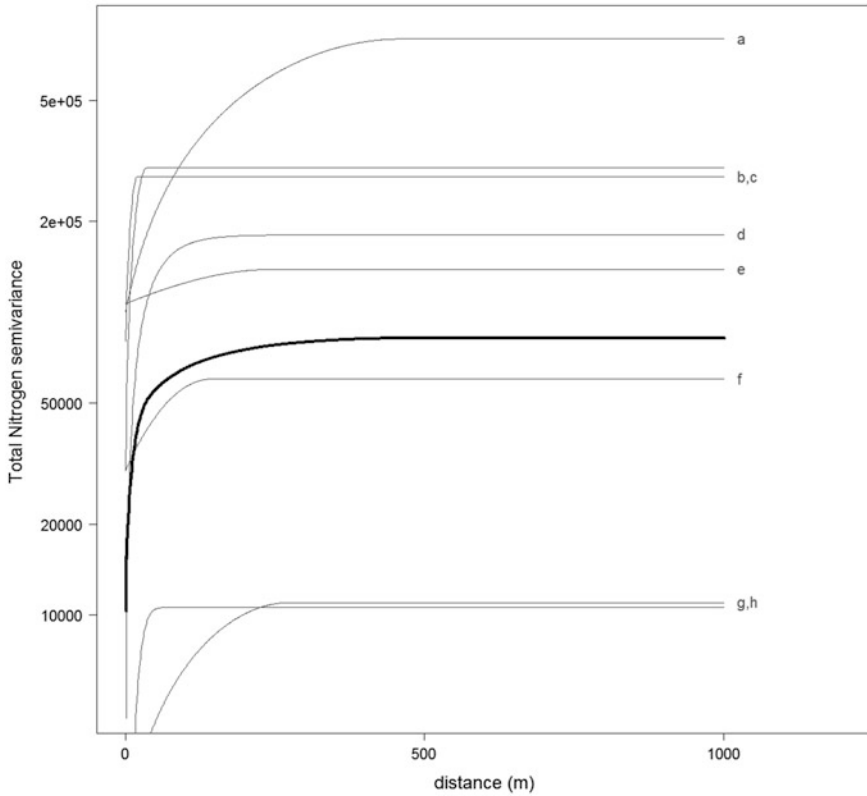


Fig. 21.7 Compilation of field scale variograms for total nitrogen. The *bold black line* represents the average variogram. Summary details and references for each variogram (*a–h*) are given in Table 21.6

Where it is available, ancillary information can be a useful source of information. Care must be taken in the selection of appropriate ancillary information. It may be useful to consult a range of ancillary variables.

21.9 Expert Knowledge

Truong et al. (2013) propose the use of expert knowledge as a means to estimate the variogram when there are not enough observations to calculate a reliable variogram using geostatistics. They point to an increasing realisation from other disciplines that experts' knowledge provides a useful source of information that can be incorporated into statistical models. Truong et al. (2013) also suggest that expert knowledge may be useful when there is no data available or even when the available data for some reason are unreliable or unsuitable.

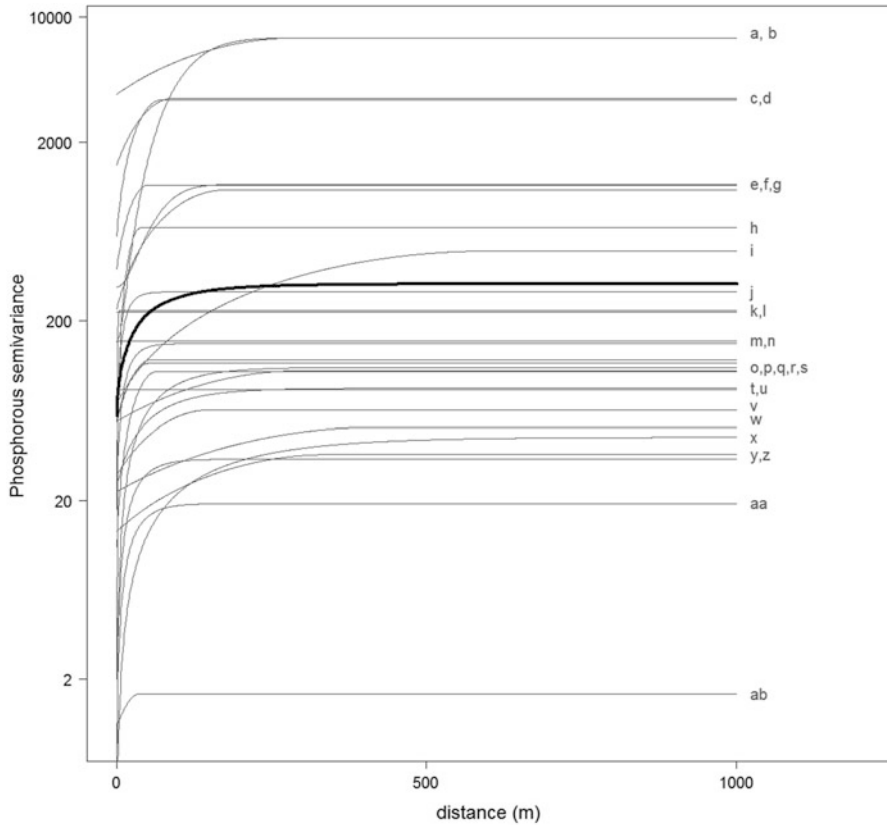


Fig. 21.8 Compilation of field scale variograms for phosphorus. The *bold black line* represents the average variogram. Summary details and references for each variogram (*a–ab*) are given in Table 21.7

Truong et al. (2013) propose a strict methodology for eliciting knowledge from experts in order to construct a variogram. Their methods are designed to avoid bias. This process is still in the prototype stage. Currently, those seeking to supplement data with expert knowledge will not be able to avoid some bias. However, in many cases, subjective expert knowledge may be the best available option.

Even when there are data available, a degree of subjectivity will be required to assess the usefulness and representativeness of these data. Where possible, it will obviously be preferred that these subjective decisions are informed by those with expertise in the area of interest (geographical or topical).

We strongly encourage the use of expert knowledge in the selection of appropriate datasets for the modelling of spatial variability. Where datasets are unavailable or deemed inappropriate, it may be necessary to rely entirely on expert knowledge to estimate the variogram. Eventually, it may be possible

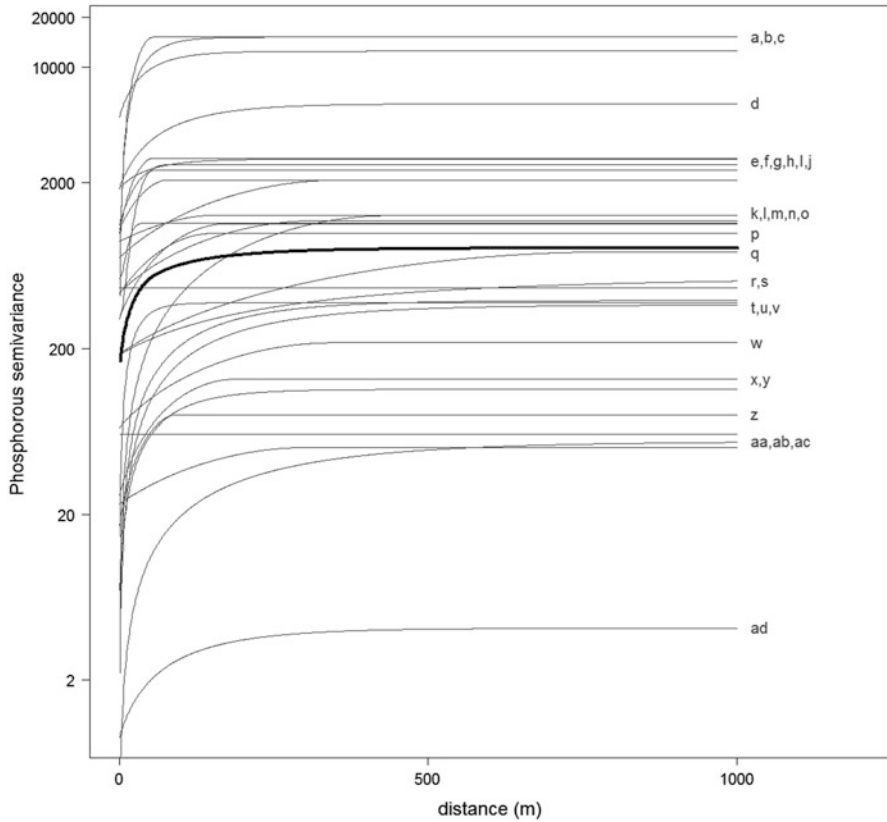


Fig. 21.9 Compilation of field scale variograms for potassium. The *bold black line* represents the average variogram. Summary details and references for each variogram (*a–ad*) are given in Table 21.8

to elicit expert knowledge using a formal process such as that described by Truong et al. (2013).

21.10 Quick Variograms

There may be situations when ancillary information, variograms from literature or even expert knowledge are unavailable or unreliable. In these cases, we would like to propose the following sampling approach that can be used to estimate a rough variogram at very low cost. The method proposed will necessarily be imprecise, but is a better alternative than not having any information. We anticipate that this method would be particularly useful in cases where alternative sources of information are available but unreliable.

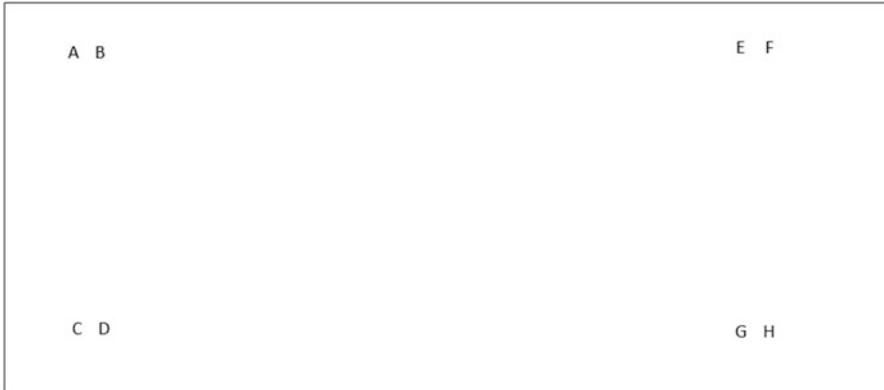


Fig. 21.10 Sampling approach for estimating a ‘rough’ variogram. Here, it is shown that sampling is recommended in eight locations (four widely spaced points, each with a closely spaced pair, i.e. *A-B*, *C-D*, *E-F* and *G-H*)

We suggest sampling in eight locations (four widely spaced points, each with a closely spaced pair as per the diagram shown in Fig. 21.10). Obviously, the sampling design will be affected by the shape of the field. We advise that sampling occurs as close to the boundaries as possible while avoiding the edge effect.

In Fig. 21.10, one can calculate four bin sizes.

Close spacing (proxy for nugget): four pairs *A-B*, *C-D*, *E-F*, *G-H*

Maximum spacing (proxy for sill): eight pairs (*A-G*, *A-H*, *B-G*, *B-H*, *C-E*, *C-F*, *D-E*, *D-F*)

Intermediate spacing 1: eight pairs *A-C*, *A-D*, *B-C*, *B-D*, *E-G*, *E-H*, *F-G*, *F-H*

Intermediate spacing 2: eight pairs *A-E*, *A-F*, *B-E*, *B-F*, *C-G*, *C-H*, *D-G*, *D-H*

The close spacing (the close pairs) can be used as a proxy for the nugget, and the maximum spacing (the diagonals) can be used as a proxy for the total sill.

If the nugget and the sill are similar, we can assume that the appropriate model is the nugget model.

If the nugget and the sill are different, we will need to estimate the range and select a model for the variability.

There is no obvious proxy for the range that can be calculated from a small number of data points.

The two intermediate spacings may be useful to indicate where the range should occur. If they are similar to the total sill, then the range should be less than the intermediate spacings. If they are smaller than the total sill, then the range should be greater than the intermediate spacings.

We suggest that the intermediate spacings be used to determine the limits of where the range could occur. The range should then be taken as the halfway point between the limits.

For example, if the smallest intermediate bin has a variance similar to the sill, then we should set the range to be equal to the halfway point between the nugget and the smaller intermediate bin. If both of the intermediate bins have variances smaller than the sill, we should set the range to be equal to halfway between the total sill and the intermediate bin.

Modelling a range larger than the maximum separation distance is always unlikely in variograms, because of the tendency of models to break down beyond half of the field extent. We do not think that unbounded models (i.e. ranges greater than the maximum separation distance) are necessary to consider here.

Quick variograms are not able to provide a precise estimation of soil variability. Their primary and important advantage is that they are calculated from data for the property of interest in the location of interest. Quick variograms provide a useful source of information when either (i) there are no other sources of information available or (ii) for checking alternative sources of information against a reference point.

21.11 Recommendations

- Where it is economically and practically feasible, the best practice for estimating variograms is to conduct a preliminary survey and then a comprehensive field survey at the spatial scales of interest.
- Where resources for field survey are limited, it is desirable to find a variogram for the same property which has been calculated for a similar soil type. This variogram may be useful as a source of information for survey design. It may also be possible to use this variogram for kriging. Due to the large variability in soil variograms calculated for each soil property, it is critical to use discretion in this selection process.
- If more than one existing variogram from a similar soil type is identified, the information from both should be used. An averaging process may be a useful way to combine this information. Alternatively, they could be used separately to provide a range of predictions.
- Ancillary information (such as that from aerial photographs) should be considered to estimate variograms for soil properties. It can be used for survey planning and for kriging. Care must be taken to ensure that these variograms have a similar nugget-to-sill ratio to the property of interest. If available, it is preferred to use information from soil survey over ancillary information.
- If it is not possible to find a variogram for a similar area, other options are available. For example, expert knowledge could be consulted. Consultation of expert knowledge may occur in a formal process-based manner or in a more informal way.
- Quick field surveys, with limited sampling, may provide a cost-effective way to estimate rough variograms where other information is unavailable. These

methods may also be useful when it is desirable to supplement information with observations directly from the field.

- Variograms are useful for designing detailed sampling surveys for agricultural and environmental purposes.

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