### Sampling for mapping and monitoring of soil carbon stocks

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### Introduction of the authors and their work

- Brus, D.J. & J.J. de Gruijter, 1997. Random sampling or geostatistical modelling? Choosing between design-based and model-based sampling strategies for soil (with Discussion). *Geoderma* 80: 1-59.
- Hoogland, T., M. Knotters, D.J. Brus & P. Kuikman, 2005. Monitoring of changes in carbon stocks in the Dutch soil. Design of a monitoring strategy (in Dutch). Wageningen, Alterra report 1354, 56 pp.
- De Gruijter, J.J., D.J. Brus, M.F.P. Bierkens & M. Knotters, 2006. Sampling for natural resource monitoring. Berlin, Springer, 332 pp.

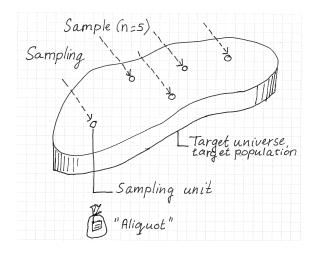


# Introduction of the authors and their work (continued)

- Brus, D.J. & J.J. de Gruijter, 2011. Design-based generalized least squares estimation of status and trend of soil properties from monitoring data. *Geoderma*, in press.
- Brus, D.J., B. Kempen & G.B.M. Heuvelink, 2011. Sampling for validation of digital soil maps. *European Journal of Soil Science* 62: 394-407.



### Clashing terms





#### To discuss sampling aspects of mapping, monitoring and validation



- Selection process of sampling locations deserves more attention: 'representative' is often ill-defined\*
- 'Start at the end, and reason backward': integrated planning of data collection and data processing, with respect to the required information

\*) Eight different meanings. See for an interesting analysis of the various meanings of 'representative sampling': Kruskal, W. and F. Mosteller, 1979. Representative sampling, I, II and III. International Statistical Review 47(1,2,3): 13-24, 111-127, 245-265.



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- Sampling for mapping of *changes* in soil carbon stocks (*monitoring*)
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#### Directed (targeted, purposive) sampling:

 collection of data can be optimized for spatial interpolation (fair spatial coverage).

#### Probability sampling:

- enables model-free estimation of means or totals and their standard errors (design-based inference).
- collected data are suitable for spatial interpolation if the sampling design guarantees fair spatial distribution of the selected sampling units.



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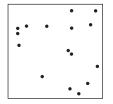
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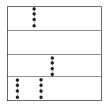
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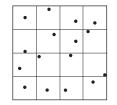
### Design types for probability sampling



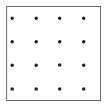
Simple random sampling



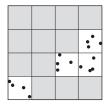
Cluster random sampling



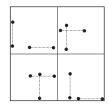
Stratified simple random sampling



Systematic random sampling



Two-stage random sampling



Stratified cluster random sampling



- we want to estimate the distribution function or parameters thereof (mean, median, P90 etc.) for the area as a whole or for subareas;
- objective estimates of target properties are required, i.e. no subjective judgement on 'representativeness', no subjective model choices;
- objective estimates of *estimation variance* or *confidence intervals* are required, i.e. validity.



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#### Model based method best choice when:

- we want to map the target property (making predictions for unvisited locations);
- ▶ sample size large enough for calibrating a model of variation (e.g. variogram: n > 100);
- strong autocorrelation exists, from which we may profit in mapping;

Further reading: D.J. Brus and J.J. de Gruijter (1997). Random sampling or geostatistical modelling? Choosing between design-based and model-based sampling strategies for soil (with Discussion). Geoderma, 80: 1-59.



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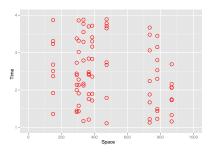
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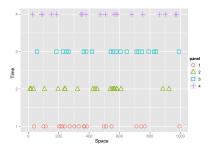
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#### 2. Sampling for mapping of changes in soil carbon stocks (monitoring)



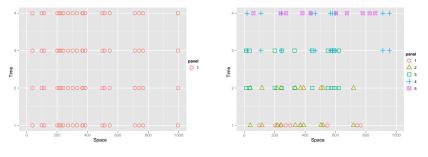
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Synchronous



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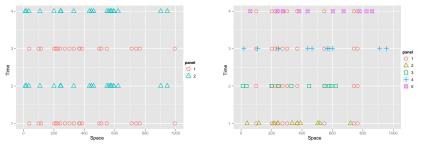


Static-Synchronous

Rotational



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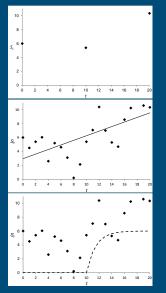


Serially alternating

Supplemented



### Changes: differences, trends or effects?



difference,  $d_{10-0} = -0.6 \text{ kg m}^{-2}$  $d_{20-10} = 4.9 \text{ kg m}^{-2}$ 

linear trend, 0.33 kg m $^{-2}$  year $^{-1}$ 

effect of policy, intervention at t = 10, step decay model, effect =6.0 kg m<sup>-2</sup> after 10 years



# 2. Sampling for monitoring

#### Possible selection modes for various aims of monitoring

space	time	
	purposive	random
purposive	space-time mapping	_
random	estimation of trends in spatial means, totals, areal proportions*	model-free estimation of space-time means

\*) 'Trend' can be defined as a time-series model parameter or as *a linear combination of true but unknown spatial means*, see: Brus, D.J., de Gruijter, J.J., Design-based Generalized Least Squares estimation of status and trend of soil properties from monitoring data, *Geoderma* (2011), doi:10.1016/j.geoderma.2011.06.001



- Validation: testing whether a map or a model satisfies its purpose
- Objectivity is crucial in validation (so that the validation procedure cannot get the blame for bad results)
- Collection of additional data by probability sampling is therefore recommended

For sampling aspects of map validation, see Brus, D.J., B. Kempen & G.B.M. Heuvelink, 2011. Sampling for validation of digital soil maps. *European Journal of Soil Science* **62**: 394-407.



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