

# Optimization of Sampling Designs for Validating Digital Soil Maps

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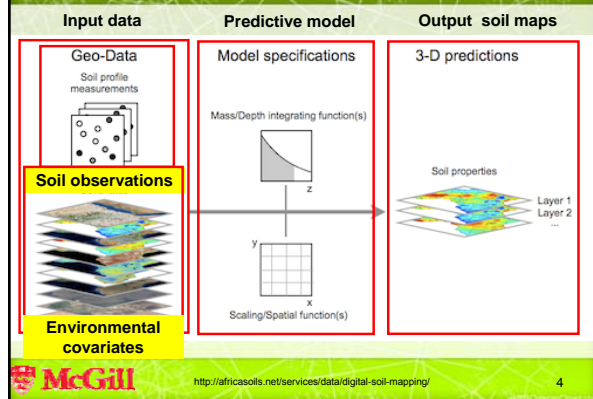
## Outline

- Introduction and objectives
- Methods
- Results
  - 3D digital soil mapping of SOC content (theory and results)
  - Comparison of sampling designs

## Digital soil mapping

- Spatial variability of soil properties
  - Jenny's theory:  $S = f(c, o, r, p, t, \dots)$  (Jenny 1941)
- Traditional soil maps
  - Delineation of soil type or properties
- Digital soil maps
  - Georeferenced soil database
  - High resolution
  - Quantitative relationship between soil properties and environmental covariates

## Digital soil mapping- Workflow



## Sampling design

- Sample size
  - Trade-off between budget and accuracy
- Sample locations
  - determined by sampling design
  - Sampling design is not the actual pattern of soil locations, but the **procedure** used to select it.


**Provide reliable input for predictive model**

## Challenges and opportunities

- Due to the technical constraints, most studies focus on **top soil**. However, the **vertical variability**, and **horizontal variability of deep soil** need more work.
- Development of **proximal soil sensing** allows us to accurately obtain the deep soil properties, which needs more detection and application.


### Objectives

- Produce 3D digital soil maps to quantify horizontal and vertical variability of soil properties
- Compare different sampling designs for digital soil maps in multiple layers (3D)




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### Study area




Field 26 of Macdonald farm  
McGill University  
Ste-Anne-de-Bellevue

Field size-11 ha




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### Sampling designs



Grid sampling	56
Grid random sampling	45
Transect sampling	37
Random sampling	44
Stratified random sampling	42
All designs together	148

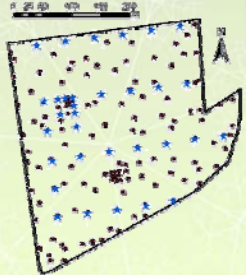




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### Sample collection

**Input data**

- 148 vis-NIR spectral data
- 32 soil samples



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### Digital soil mapping- Procedure

**Input data**

- Soil properties at 32 locations
- Spectral data at 32 locations
- Spectral data at 148 locations


**Prediction**


**Partial Least Square Regression (PLSR)**

**Output**

- Soil properties at 148 locations
- Environmental covariates

**Maps**





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### Sampling designs- Procedure

**Output data**


**Soil properties at 148 locations**

**Sampling designs**

- Grid sampling 56
- Grid random sampling 45
- Transect sampling 37
- Random sampling 44
- Stratified random sampling 42

**Sub maps**

- GS
- GRS
- TS
- RS
- StRS



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## Soil properties and environmental covariates

### > SOC content

Laboratory: loss-on-ignition

**Normalization and transformation:**  $y = \log\left(\frac{\frac{x}{100}}{1 - \frac{x}{100}}\right)$

### > Environmental covariates

- Elevation(RTK)
- Gamma-ray radiation- Total count, K, U, Th, Cs
- Apparent electrical conductivity (DUALEM)- 1m-HCP, 1m-PRP, 2m-HCP, 2m-PRP

## Results- 1

### 3D digital soil mapping of SOC content by using regression-kriging

## 3D mapping procedure

### > Regression Kriging (RK)

Prediction = Trend Prediction (Regression) + Residual Prediction (Kriging)

### > 3D- Regression Kriging (RK)

$$\hat{z}(s_0, d_0) = \sum_{j=1}^p \beta_j \cdot X_j(s_0, d_0) + \hat{g}(d_0) + \sum_{i=1}^n \lambda_i \cdot e(s_i, d_i)$$

Regression
Depth function
Residual kriging

## 3D mapping procedure

### > 3D variogram

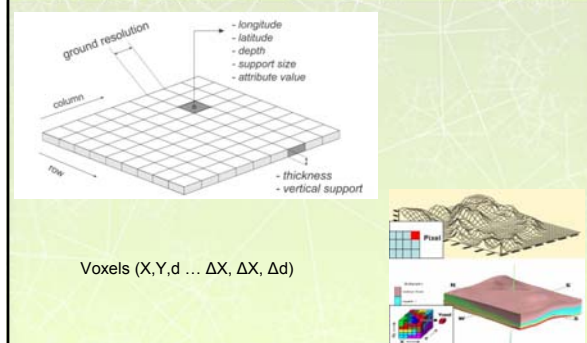
Exponential model, consisting of three standard parameters ( $c_0$ ,  $c_1$ ,  $r$ ), and 2 to 3 geometric anisotropy parameters ( $a_x$ ,  $a_y$ ,  $a_d$ ).

$$\gamma(h) = \begin{cases} 0 & \text{if } |h|=0 \\ c_0 + c_1 \cdot [1 - e^{-\frac{|h|}{r}}] & \text{if } |h|>0 \end{cases} \quad h=[h_x, h_y, h_d]$$

The isotropic lag ( $h$ ) is calculated by scaling the horizontal and vertical distances by using anisotropy parameters:

$$h = \sqrt{\left(\frac{h_x}{a_x}\right)^2 + \left(\frac{h_y}{a_y}\right)^2 + \left(\frac{h_d}{a_d}\right)^2}$$

## 3D mapping procedure



Voxels (X,Y,d ... ΔX, ΔY, Δd)

## Spectral model- PLSR

### > Cross-validation results

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \hat{x}_i)^2} \quad RPD = \frac{\text{Standard Deviation}}{RMSE}$$

### Original SOC maps- Google earth

- Soil SOC maps
- Sampling variance (standard error)
- 90% Confidence Interval (Lower boundary and upper boundary)

General  
linear model  
(GLM)

Regression  
tree  
(RT)

Random  
forest  
(RF)

- Depth intervals: 0-5cm, 5-15cm, 15-30cm, 30-60cm, 60-100cm.

McGill 19

### Results- 2

## Comparison of different sampling designs for calibrating digital soil maps

McGill 20

### Comparison of sampling designs

GS

GRS

RS

StRS

TS

Ori

Spatial coverage

Distribution of values

Accuracy var+RMSE

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