

Advancing Digital Soil Mapping tools in support of forest resource inventory, planning and decision-making



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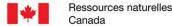


Land Acknowledgement

I'd like to respectfully acknowledge that the areas covered by the work I am presenting today are on the inherent and treaty lands of First Nations and within the traditional lands of Indigenous peoples. For thousands of years, Indigenous peoples inhabited and cared for this land, and continue to do so today.

My office at GLFC is located in Robinson-Huron Treaty territory, this land is the traditional territory of the Anishinaabek, specifically the Garden River and Batchewana First Nations, as well as Métis People.

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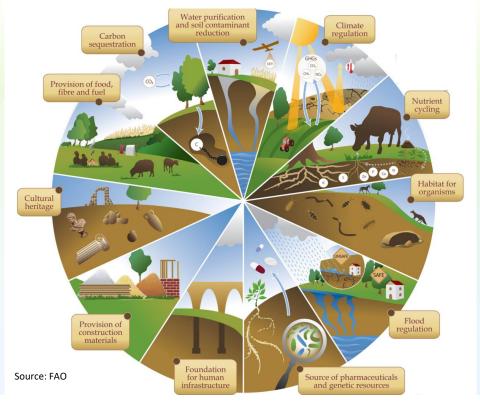




Soil: A foundation for healthy forests



Senate committee on Agriculture and Forestry report on status of soil health in Canada (due by the end of this year)







Soil: A foundation for healthy forests





- Moisture
- Depth
- **Bulk density**



- Nitrogen
 - рН
- **Cations**





Biological

Chemical

- Microbial activity/diversity
- Soil fauna

Physical





Soil information is essential for many different applications

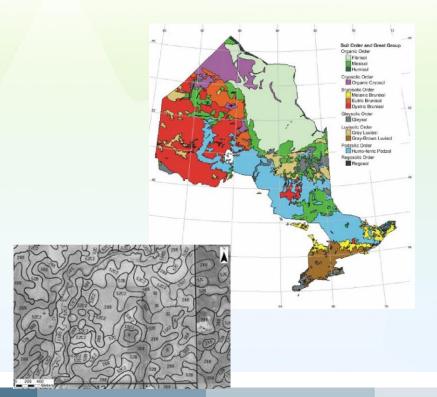
Soil Management Forest Management Site sensitivity Biomass harvesting Site productivity • Veg. Competition Slope stability hazard Tree species selection Waste management Pest hazard Trafficability Silvicultural practices Interpretive **Applications of Biodiversity Management** Soil property Landscape unit planning Wildlife Management Maps Biodiversity guidelines Critical wildlife habitat •Old growth management Wildlife suitability •Rare ecosystem ID Forage potential Non-timber forest products **Policy** Carbon accounting Climate change adaptation

- The type of soil information (i.e., specific properties) might be different for different applications.
- Understanding how these properties vary across the landscape will improve our ability to practice Sustainable Forest Management



But... soils are not easy to map

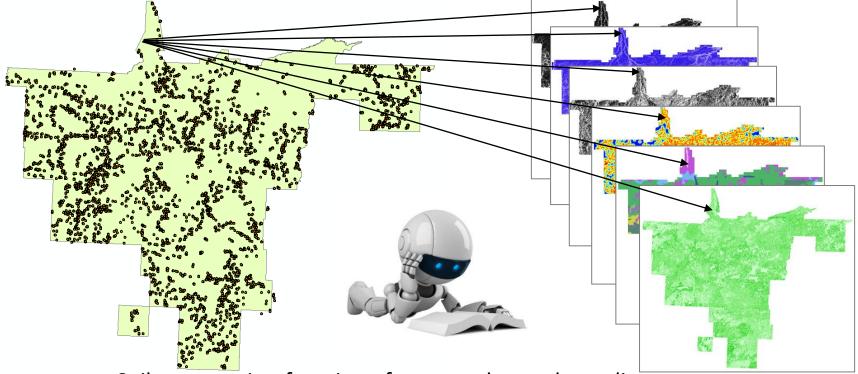
- Existing maps are coarse
- Not easily remotely sensed
- Inferred from landscape position and vegetation associations
- Forest soils offer additional challenges (e.g., tree canopy, forest floor, complex terrain)







Digital soil mapping



Soil property is a function of topography, geology, climate, ...

How do we improve on our forest soil mapping?

- Adapting digital soil mapping to forest systems by mapping at relevant scales for forest management
 - Data, data, data
 - What are the key properties (or proxies for the key properties)
 - Workflow
- Mapping soil properties for different applications





Project Goals

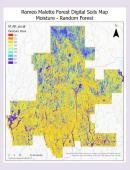
Compiling legacy data and facilitating soil data acquisition.

| MAP | Lager | File Nume & Location | Project |
|-----|-----------|---|---|
| | FEO_ELO-M | ENFANCETY Francoultrick, SR_MART BurkletenArthurdEC_ ELC-Highp | Forvet Emerytum Chardinates aydum Emerytum Chardinatha Incertain the Distrant Oute Forganismy (EDR) database |

2. Summarizing key indicators or metrics of soil properties.

$$D_B = f(Depth, SOC, CF, pH,)$$

3. Expanding soil property map products and their applications.



Dataset Information

C/N ratio

Exch catio

Publications

Year(s): 2003-2016, ongoing?

Humus form

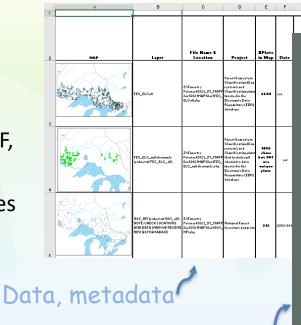
Structure

Roundan

Acid test (K)

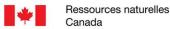
Compiling legacy soil data

- Large datasets (NFI, FEC/ELC, G&Y, FRI, VSN)
- Regional datasets (ARNEWS)
- Project datasets (LTSP, TLW, PRF, AFRIT, university)
- Field-determined soil properties vs. lab-determined soil properties

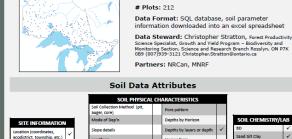


Fact sheets

Shelagh Yanni



Natural Resources Canada



Landform

Stone/Pock outcron

National Forest Inventory (NFI)

Project Details

This dataset includes the Ontario ground plots that are assessed for the ongoing National Forest Inventory (NFI) sampling program that was originally established between 2003 and 2006. The NFI program is a joint effort between federal, provincial, and territorial governments, and plot remeasurements are spread over a 10-year cycle. The 212 Ontario

plots were assessed for 25 key attributes as well as for additional variables that met provincial growth and yield requirements. The NFI program in Ontario is a component of the

Forest Soil Datasets

in Ontario

Ontario Growth and Yield program led by the MNRF.

Plot Map

Site Treatment

Facilitating soil data acquisition

- Encouraging soil collections as part of other studies, community-based monitoring
- Flexible sampling protocols and guidance documents

Overview

Planning

Field collection

Processing soils

Laboratory analyses

Data management



David Paré, Charlotte Norris, Shelagh Yanni, Stephanie Nelson, Dave Morris, Kim Chapman

Quick

booklets





Compiling legacy data and facilitating soil 12 data acquisition

Key challenges

- Locational accuracy
- Subjective/qualitative measures
- Non-standard methods
- Expense of field collection of samples and processing and analyses.

Opportunities

- Linking soil databases
- Rapid assessment methods
- In-situ quantitative sensors
- Rapid laboratory methods



Summarizing key indicators or metrics of soil properties

- Some properties are critical but may be difficult to measure.
- Can more easily measured properties serve as a proxy?
- Can we build "Pedo-transfer functions" from these proxies?

Example of Bulk Density

 $D_B = f(Depth, SOC, CF, pH, Silt, Sand, Clay)$

 $D_B = f(Depth, SOC, CF, pH,)$

 $D_B = f(Depth, SOC, CF)$

 $D_B = f(Depth, SOC)$

 $D_B = f(SOC)$

 $D_B = f(Depth)$



Summarizing key indicators or metrics of soil properties



Article

A quantitative approach to defining soil nutrient regimes within ecosystem classifications for Northwestern Ontario

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Abstract

Soil nutrient regimes (SNRs) are often incorporated in ecosystem classifications. Evaluation of actual nutrient levels associated with these SNRs and the development of complementary soil chemistry regimes (SCRs) could broaden their utility. Using data from 618 forest stands in northwestern Ontario, we developed five-category SCRs using K-means clustering and examined relationships among individual nutrients, SCRs, and the SNRs of the Canadian National Vegetation Classification Associations and the Ontario Ecological Land Classification Ecosites. F, A, and B horizon samples were analyzed for organic C (OrgC), total N (TotN), CN ratio (C:N), cation exchange capacity (CEC), exchangeable bases, base saturation (BaSat), and pH. CEC, pH, and BaSat showed good correspondence across horizons, and together with C:N accounted for much of the variation in chemical properties. There was broad agreement between Association and Ecosite SNRs and B horizon (BHorz) and All horizon (AllHorz) SCRs. C:N decreased while pH and cation metrics increased with increasing SNR and SCR richness. User's accuracies (SNRs vs. SCRs) for the classifications ranged from 31%-39% but increased to 80%-86% for SNR values within ±1 SCR class. Classification trees identified pH class, soil texture, and overstory composition as the principal field-measure factors related to BHorz SCRs.

Key words: ecosystem classification, nutrient regime, soil chemistry, soil properties, Ontario

Building pedotransfer functions for soil nutrient regime

J. Zhang¹, B. Heung¹, Shelagh Yanni², Kara Webster²





Texture and pH are important to nutrient regime

Rob Fleming, Jin Zhang





their applications – Step 1: Workflow

Workflow

- Plot data
- Spatial layers and covariates (LiDAR DEM acquisition)
- Model choice (kNN, RF, SVM)
- Variable Inflation Factor, Recursive Feature Elimination to remove multi-collinearity
- Resampling to improve balance across categories
- Variable importance

of DSM workflow best

"Tips and Tricks" report practices

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Jeremy Arkin, Grant McCartney, Brandon Heung, Ian Sinclair





Test area – Romeo Malette Forest

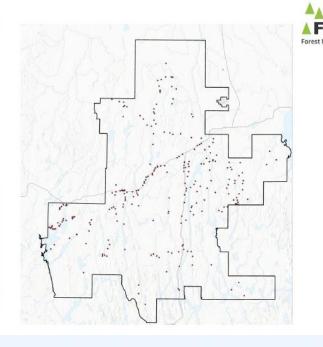
RMF VSN Plots

○ Plot Data

258 Plots

| Moisture | | |
|----------|-------------|--|
| Class | # of points | |
| CO | 9 | |
| C1 | 39 | |
| C2 | 97 | |
| C3 | 56 | |
| C4 | 19 | |
| C5 | 17 | |
| C6 | 5 | |
| C7 | 7 | |
| C8 | 9 | |

| Text | ure |
|--------------|-------------|
| Class | # of points |
| Clayey | 11 |
| Coarse loamy | 128 |
| Fine loamy | 12 |
| Folic | 25 |
| Organic Peat | 15 |
| Sandy | 51 |
| Silty | 16 |



Covariates

- Topography (25 variables)
- Geology (surficial and bedrock)
- Hydrology (2018 LiDAR derived water bodies and stream network)
- Biology data (forest type)

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Expanding soil property map products and their applications – Step 2: Outputs

Outputs

- How good was the model? Accuracy and Kappa with confusion matrices
- What are the most important covariates? Variable importance
- What is the distribution of the property? Maps of soil property
- How confident are we in the prediction? Certainty/entropy maps

Jeremy Arkin, Grant McCartney, Brandon Heung

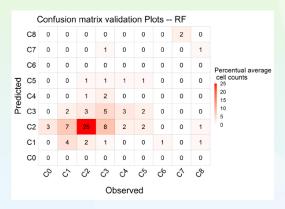




Accuracy, Kappa and Confusion Matrices

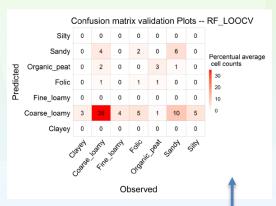
Moisture Regime

| Accuracy | Карра |
|----------|-------|
| 0.42 | 0.19 |



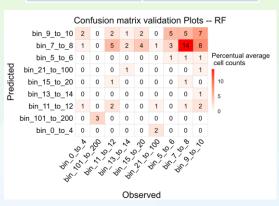
Texture

| Accuracy | Карра |
|----------|-------|
| 0.46 | 0.12 |



Overpredicting coarseloamy category Depth of AB

| Accuracy | Карра |
|----------|-------|
| 0.36 | 0.15 |



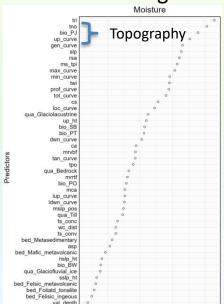
Jeremy Arkin, Grant McCartney



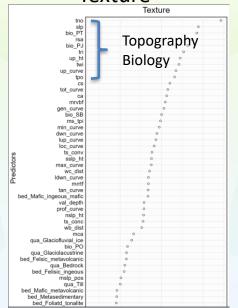


Variable Importance

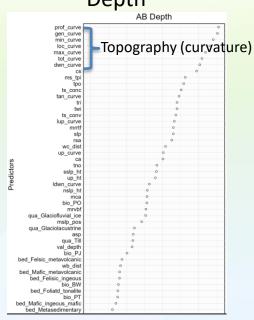
Moisture Regime



Texture



Depth



Jeremy Arkin, Grant McCartney

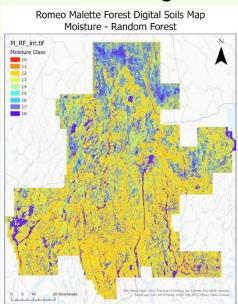
Increasing importance



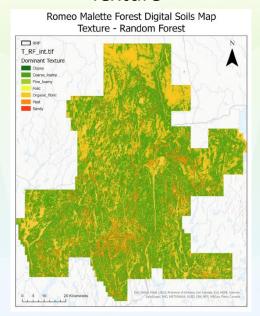


Raster Maps

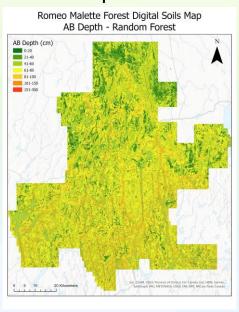
Moisture Regime



Texture



Depth



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Certainty/Entropy

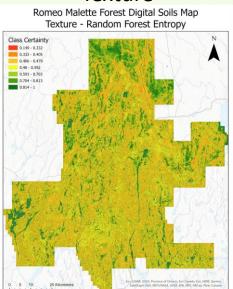
Moisture Regime

Romeo Malette Forest Digital Soils Map Moisture - Random Forest Entropy

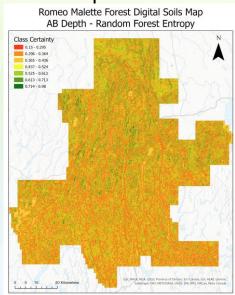
Class Certainty

0.112 - 0.258
0.039 - 0.031
0.030 - 0.0515
0.040 - 0.0515
0.0516 - 0.057
0.024 - 1

Texture



Depth



Jeremy Arkin, Grant McCartney





Expanding soil property map products and their applications

Key challenges

- Covariate layers
 - What are the quality of the covariate layers?
 - Which are most important?
 - Are more covariates better?
- Combining different data sources

Opportunities

- LiDAR acquisition for topography and biota/inventory data
- Scripts that streamline the process
 - Options for different model types
 - Consistency in the workflow and outputs





Recommendations and Next Steps

- Encourage opportunistic sampling of soils
- Rapid, easy and standardized field methods
- Testing new technology and funding for lab analyses
- Sample archives (additional soil properties or new technologies)
- Data repositories
- Best practices, scripts, and training (test data)
- Continued testing and application at FMU scales.







Thank you!

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