SoilGrids250m and SoilInfo App: providing public access to soil information across borders



T. (Tom) Hengl <<u>tom.hengl@isric.org</u>>





Thanks to:

• ISRIC staff:

Jorge Mendes de Jesus, Maria Ruiperez Gonzalez, Bas Kempen, Robert McMillan, Niels Batjes, Eloi Ribeiro, Gerard Heuvelink, Johan Leenaars and others

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• External collaborators:

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AfSIS project

Bill & Melinda Gates Foundation

Business Operation

Bill & Melinda Gates Foundation is one of the largest private foundations in the world, founded by Bill and Melinda Gates. It was launched in 2000 and is said to be the largest transparently operated private foundation in the world. Wikipedia

Nonprofit category: Private Grantmaking Foundations

Founded: 2000

Assets: 36.79 billion USD (2010)

Income: 53 billion USD (2010)

Founders: Melinda Gates, Bill Gates

BILL&MELINDA GATES foundation



Also thanks to:





A bit of history



Why SoilGrids?





Why SoilGrids?





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Search Google Scholar	quantitative and have taken advantage of the digital revolution, conventional soil mapping delineates space mostly

Why SoilGrids? To fill in the gap.





Rome, FAO (Gerda Verburg twitter)



SoilGrids were officially launched on 4th of Dec 2013 at FAO, Rome



5 December 2013



SoilGrids1km — Global Soil Information Based on Automated Mapping

Tomislav Hengl , Jorge Mendes de Jesus, Robert A. MacMillan, Niels H. Batjes, Gerard B. M. Heuvelink, Eloi Ribeiro, Alessandro Samuel-Rosa, Bas Kempen, Johan G. B. Leenaars, Markus G. Walsh, Maria Ruiperez Gonzalez

Published: August 29, 2014 • DOI: 10.1371/journal.pone.0105992

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 Abstract Introduction 	Abstract				🔘 CrossMark		
Materials and Methods	Backgroun	d			Subject Area	as	?
Results	5						
Discussion	Soils are widely	recognized as a non-renev	wable natural resource and	l as biophysical carbon	Agricultural	oil scie	
Acknowledgments	sinks. As such, t	here is a growing requiren	nent for global soil informat	ion. Although several	Clay mineral	ogy	F
Author Contributions	limited spatial de	etail.	st, these tend to suffer from	n inconsistencies and	Deserts		P.
References	Methodolog	gy/Principal Findi	ngs		Forecasting		
					Remote sens	ing	Par.
Reader Comments (0)	We present Soil	Grids1km — a global 3D s	oil information system at 1	km resolution —	Surface temp	erature	in.
Figures	organic carbon (g kg-1), soil pH, sand, silt	and clay fractions (%), bul	k density (kg m-3), cation-			-
	exchange capac	ity (cmol+/kg), coarse frag	ments (%), soil organic ca	rbon stock (t ha-1), depth	Taxonomy		14
	to bedrock (cm), predictions are b	World Reference Base so	oil groups, and USDA Soil T ediction models which we fit	Faxonomy suborders. Our	Urban areas		





Global Soil Information Facilities



SoilGrids



- Soil property-depth curves (with confidence intervals)
- Dominant, 2nd and 3rd soil type
- Nutrient status (e.g. soil organic carbon stock)
- Available water capacity
- Current and past land use
- Closest soil data provider / agricultural extension service

SoilGrids1km: criticism

- Artifacts (mainly propagating from HWSD)
- Some soil properties are smoothed-out (regression only) over-prediction of low values (Organic carbon) / under-prediction of high values (see e.g. doi: 10.1016 /j.geoderma.2015.08.035)
- **Only linear** soil property-depth relationships
- No predictions for deserts (are deserts soil bodies?)



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Abstract						CrossMark		
Introduction	Abstract					Cubicat Area	. 0	
Materials and Methods	80% of arable la	nd in Africa has low soil fe	ertility and suffers from p	hysical soil prob	lems	Subject Area	5	
Results	Additionally, sign	ificant amounts of nutrier	its are lost every year du	ie to unsustainal	ole soil	Machine learni	ng 💿	l.
Discussion	management pro knowledge. To h	actices. This is partially th elp bridge the soil informa	e result of insufficient us ation gap in Africa, the Af	e of soil manage frica Soil Informa	ment tion Service	Agricultural so	oil scie 💿	E.
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Author Contributions	most compreher	nsive soil sample data set	s of the African continent	t to date. Utilizin	g these point	A C :-	0	2
References	data sets in com spatial prediction pH, sand, silt an	bination with a large num ns of soil properties releva d clay fractions, bulk dens	per of covariates, we hav ant to the agricultural mai sity, cation-exchange cap	ve generated a s nagement—orga pacity, total nitrog	series of anic carbon, gen,	Soil ecology	0	
Reader Comments (0)	exchangeable a	cidity, AI content and exch	angeable bases (Ca, K,	Mg, Na). We sp	ecifically			

Predictions soil org. carbon (1km vs 250m)



World Soil Information

ISRIC

1 km (left) vs 250 m (right). Location: Arusha, Tanzania

SoilGrids → SoilInfo App





GBIF's vision: "A world in which biodiversity information is freely and universally available for science, society and a sustainable future."

GSIF's vision: "... soil information ..."



SoilGrids250m



Aspect	SoilGrids1km	SoilGrids250m
Spatial resolution	1000	250 (<mark>16 x bigger</mark>)
Input point data	ca 100,000 points	ca 130,000 with points also in Russian Federation, Artic/Antartic
Covariates	ca 70 unique	ca 150 (<mark>2 x more</mark>)
Prediction domain (soil mask)	excludes deserts, permanent ice	excludes only permanent ice
Spatial prediction methods	linear regression (no kriging)	machine learning (random forests, neural networks etc) / ensemble predictions
Artifacts and	country borders, artifacts and noise from MODIS products, artifacts in the geological map and HWSD	*reduced to minimum
Systematic bias	(over-smoothing) over- prediction of low value	*to be tested



This is probably one of the biggest (statistical) soil mapping projects ever!







ca 150,000 points shown on this map





>850,000 measurements of soil organic carbon

ca 45,000 points shown on this map



FAO World Reference Base



"Sapric.Histosols", "Hemic. Histosols", "Fibric.Histosols", "Cryic.Histosols", "Histic. Albeluvisols

ca 52,000 points shown on this map



USDA Soil Taxonomy



"Saprists", "Hemists", "Folists", "Fibrists"



Soil_Taxonomy 🔺 🖿

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32	Haplic Arenosols Eutric	8	psamments	typic udipsamments	Cryopsamments
3	Haplic Podzols	8	aquods	typic haplorthods	Epiaquods
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General spatial prediction workflow (data-driven)



Regression matrix = <u>154 covariate</u> <u>layers</u> x 150,000 sites (soil profiles / samples)

Model fitting = Random Forest, neural nets / Deep learning



h2o software



Results



Model testing

- FAO / USDA Soil taxonomy classes:
 - Estimated Cohen Kappa (weighted): 29–40%
 - Map purity: 30–40%
- R-square for soil organic carbon, pH, soil textures, coarse fragments, bulk density (Random Forest models): 63–82%







Fibrists



ISRIC

World Soil Information

Fibrists



Soil organic carbon



Soil pH (5–15 cm)



/orld Soil Information





Conclusions



1. (Non-linear) machine learning algorithms show predictive potential for automated soil mapping



2. There is a potential in soil mapping automation and in using global soil data models (to fill-in the data gaps / transfer knowledge)



3. SoilGrids could be used to generate even hundreds of soil properties (automation is possible)!



Get ready for the Soil Data Revolution!





God's Word is like an icebergthere is more truth unseen than seen







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