Contents lists available at ScienceDirect

Geoderma

journal homepage: www.elsevier.com/locate/geoderma

Further results on comparison of methods for quantifying soil carbon in tropical peats

ABSTRACT



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ARTICLE INFO

Article history: Received 22 December 2015 Received in revised form 20 January 2016 Accepted 24 January 2016 Available online 6 February 2016

Keywords: Tropical peat Carbon density Carbon content Bulk density

1. Introduction

Tropical peats have the highest terrestrial carbon store per unit area of land (Moore et al., 2013; Page et al., 2011) and estimation of their carbon stocks is essential for both conservation and management practices. Thus, it is useful to have a simple soil carbon estimation or pedotransfer function for tropical peatlands where access to laboratory analysis can be difficult or expensive. In mineral soils, bulk density is known to be affected by organic matter content, and pedotransfer functions have been proposed to estimate bulk density from soil organic matter content as measured by the loss on ignition (LoI) method (Adams, 1973; Jeffrey, 1970). Adams (1973) observed that bulk density decreases exponentially with an increasing organic matter content until a value of 75% where the relationship doesn't hold. Warren et al. (2012) noted that there is no expected relationship between bulk density and carbon content for peat soils as mineral fractions are very low.

Farmer et al. (2014) proposed new equations for predicting carbon content and carbon density for tropical peats, which include:

1. A new factor conversion value for predicting carbon content, C_c from its ash content as measured by the loss on ignition (LoI) method, based on a relationship between LoI and elemental analysis data;

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2. A linear equation for predicting carbon density, C_d (in kg m⁻³) from its bulk density, BD (in g cm $^{-3}$):

We present an alternative equation for estimating carbon density of tropical peatlands. We compiled a dataset of

tropical peatlands with various land uses and found that when carbon content is greater than 0.5 g g^{-1} , there is

no relationship between carbon content and bulk density. Thus, carbon density can be estimated from an average

carbon content (0.5501 ± 0.0225 g g⁻¹) multiplied by the measured bulk density. This simple model is in con-

trast with previous studies, where a linear regression model is fitted to the carbon density and bulk density data. We tested the model to the data and demonstrated its high accuracy and applicability across land uses.

$$C_d = (515.44 \times BD) + 3.01 \tag{1}$$

which is an extension of the equation proposed by Warren et al. (2012):

$$C_d = (468.72 \times BD) + 5.82 \tag{2}$$

The slope of Eqs. (1) and (2) represents an average carbon content (C_c) of tropical peats, i.e. 0.5154 and 0.4687 g g⁻¹, respectively. Both Farmer et al. (2014) and Warren et al. (2012) provided uncertainty estimates of the regression models. Eqs. (1) and (2) aimed to circumvent the need for a more expensive laboratory analysis of C content, as BD can be determined relatively easy using a standard oven.

Warren et al. (2012) derived their model from peat forests data and stated that Eq. (2) is valid for well-developed peats having carbon content greater than 0.4 g g^{-1} (or 40%). Farmer et al. (2014) showed that the accuracy of Warren's Eq. (2) equation decreased with increasing bulk density and they suggested that the equation was applicable for bulk density values between 0.05 to 0.16 g cm⁻³. Meanwhile, Farmer's Eq. (1) was developed for peat soils affected by mechanical compaction (e.g. under oil palm plantations).

In this communication, we would like to propose an alternative simple relationship for predicting C_d directly from C_c which considers the uncertainty of the data. In addition, we will define the applicable range of bulk density for this relationship based on a new data set of tropical peatlands with diverse land use.





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2. Dataset

We compiled a dataset of carbon content and bulk density from various tropical peatlands in Indonesia, Malaysia and Peru with different land uses (i.e., primary and secondary forest, natural timber production, planted forest *Acacia crassicarpa* and oil palm) published in the literature and from our past studies. The details of the location, the number of observations, land use, observed depth ranges, and methods used for determining carbon content are listed in Table 1. Carbon content from our studies used the loss on ignition (LoI) method (Agus et al., 2011; Farmer et al., 2014), while data from the literature used either LoI or elemental analysis technique. Bulk density was measured using the gravimetric method.

3. Results and discussion

In total, our compiled dataset has 577 observations with bulk density values between 0.013 and 0.572 g cm⁻³, and carbon content ranges between 0.11 and 0.62 g g⁻¹ (Table 2). Fig. 1 shows the plot of bulk density, BD vs. carbon content, C_c with their histogram. The plot shows that when carbon content values are above 0.5 g g⁻¹ (N = 483 or 84% of the

data), there is no relationship between C_c and BD. And within this data range ($C_c > 0.5$ g g⁻¹), 95% of the data have BD values less than 0.25 g cm⁻³.

We then separated out the data where BD is less than 0.25 g cm⁻³ and C_c is greater than 0.5 g g⁻¹. In this subset of data, we calculated an average value of C_c of 0.5501 g g⁻¹ (or 55.01%) with a standard deviation of 0.0225 g g⁻¹ (or 2.25%), and within these values, C_c values are constant with varying BD values (Fig. 2a).

Since C_d is a dot product of C_c and BD ($C_d = C_c \times BD$), and there is no relationship between C_c and BD, we can estimate C_c directly from its average value ($\overline{C_c}$ in g g⁻¹):

$$C_d = \bar{C_c} \times BD = 0.5501 \pm 0.0225 \times BD$$
 (.(3))

where the units of C_d and BD should both either in g cm⁻³ (Mg m⁻³) or kg m⁻³. This method is in contrast with the regression approach of Warren et al. (2012) and Farmer et al. (2014) where they fitted a linear regression $C_d = a + b \times BD$ to the data relating BD to C_d . We believe that the regression approach has a couple of shortcomings: firstly, this regression presents co-dependent variables ($C_c \times BD$ vs. BD), and thus the estimate of the slope depends on both C_c and BD and there could

Table 1

Data of tropical peatlands' carbon content and bulk density obtained from the literature and authors' studies.

Location	Number of observations	Depth ranges (cm)	Land use	Method of carbon measurement	Source
Bengkalis, Riau, Indonesia	6	5-30	2nd rotation of planted forest	CHNS autoanalyzer	Sumawinata et al. (2014)
Central Kalimantan, Indonesia	16		Acacia crassicarpa Virgin or secondary forested including riverine, terrace, basin, marginal, floodplain, coastal	CHN elemental analyzer	Shimada et al. (2001). Average value
Demeni, Central Amazon, Brazil	3	30-140	Forest	CN analyzer	Lähteenoja et al. (2013)
Zalalá, Central Amazon, Brazil	4	30-170	Semi-open		
Daracuá, Central Amazon, Brazil	3	30-210	Forest		
Calibuqui, Central Amazon, Brazil	2	30–70	Open		
Jambi, Indonesia	6	0–65	2nd rotation of planted forest Acacia crassicarpa	CHNS autoanalyzer	Sumawinata et al. (2014)
Katingan, Central Kalimantan, Indonesia	12	20	Secondary forest	Loss on ignition	Boehm and Frank (2008)
Kubu Raya, West Kalimantan, Indonesia	131	5–75	Secondary forest and planted forest Acacia crassicarpa	Loss on ignition, conversion factor $= 1.724$	Our data collection
Miraflores-Amazon Peru	5	30-290	Forest	CN analyzer	Lähteenoja and Page (2011)
Nueva York-Amazon Peru	5	30-290	Forest		
Aucayacu	15	30-730	Forest		
Roca Fuerte	11	30-520	Center: forested,		
Nueva Alianza	4	30-190	High canopy; edge: floodplain forest		
Maquía	7	30-390	Open, scattered M. flexuosa		
San Roque	3	30-540	Open, scattered M. flexuosa		
Buena Vista del Maquía	2	30-140	<i>M. flexuosa</i> palm swamp		
Ogan Komering Ilir, South Sumatra, Indonesia	6	5–10	<i>M. flexuosa</i> palm swamp Secondary forest and planted forest <i>Acacia crassicarpa</i>	Loss on ignition, conversion factor $= 1.724$	Our data collection
Ogan Komering Ilir, South Sumatra, Indonesia	7	10-70	Planted forest Acacia crassicarpa	CHNS autoanalyzer	Sumawinata et al. (2014)
Ogan Komering Ilir, South Sumatra, Indonesia	16		Oil palm	Loss on ignition, conversion factor $= 1.724$	Prayitno et al. (2013)
Pontianak, West Kalimantan, Indonesia	36		Secondary forest	Loss on ignition, conversion factor $= 1.724$	Our data collection
Padang Sugihan, South Sumatra	2	0-40	Primary forest	Elemental analyzer	Brady (1997)
Sugihan East, South Sumatra	2	0-40	Primary forest	-	• • •
Padang Island, Riau, Indonesia	6	0-40	Primary forest		
Quistococha, Peruvian lowland Amazonia, Peru	5	70-400	Forest	Leco Analyzer	Lähteenoja et al. (2009)
Rinón	4	90-380	Open peatland		
San Jorge	6	70–570	<i>M. flexuosa</i> palm swamp and forested		
Rasau Jaya, Kubu Raya, West Kalimantan, Indonesia	11	0-700			Dariah et al. (2011)
Rokan Hilir, Riau, Indonesia	20		Tropical rainforest (natural timber production)	Walkley and Black	Yuono (2009)
Sarawak, Malaysia	3		Forest, sago, oil palm		Melling et al. (2005) Average value
Sebangau, Central Kalimantan, Indonesia	57	3-960		Leco Analyzer	Page et al. (2004)
Siak, Riau, Indonesia	131	10-300	Secondary forest	Loss on ignition	Our data collection
Total	577				

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