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Improving estimates of tropical peatland area, carbon storage, and greenhouse gas fluxes — Source link \square

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2 and greenhouse gas fluxes

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46 Abstract

47 Our limited knowledge of the size of the carbon pool and exchange fluxes in forested lowland tropical 48 peatlands represents a major gap in our understanding of the global carbon cycle. Peat deposits in 49 several regions (e.g. the Congo Basin, much of Amazonia) are only just beginning to be mapped and 50 characterised. Here we consider the extent to which methodological improvements and improved 51 coordination between researchers could help to fill this gap. We review the literature on measurement 52 of the key parameters required to calculate carbon pools and fluxes, including peatland area, peat bulk 53 density, carbon concentration, above-ground carbon stocks, litter inputs to the peat, gaseous carbon 54 exchange, and waterborne carbon fluxes. We identify areas where further research and better

55 coordination are particularly needed in order to reduce the uncertainties in estimates of tropical 56 peatland carbon pools and fluxes, thereby facilitating better-informed management of these 57 exceptionally carbon-rich ecosystems.

58 Keywords

59 Peat, greenhouse gases, remote sensing, tropical ecology, carbon cycle.

60 I Introduction

Most peat in the tropics is located in the lowland humid forests of Southeast Asia, Amazonia, Central 61 62 America and equatorial Africa (Figure 1). Page et al. (2011b) estimated the extent of tropical peatlands 63 at 441,000 km² and their carbon (C) pool at 81.7–91.9 Gt C. This carbon pool is susceptible to climate 64 change as well as local human impacts. Some peatlands in inland regions of Kalimantan have lost 65 carbon due to increasing El Niño intensity and changing sea levels during the late Holocene (Dommain 66 et al. 2011, 2014), suggesting that these peatlands may also respond to future (anthropogenic) climatic 67 change. More recently, peatlands throughout Southeast Asia have been degraded by logging and 68 plantation development (Miettinen et al. 2012). Drainage to improve growing conditions for crops 69 such as oil palm and Acacia (for pulpwood) leads to peat subsidence and enhanced CO_2 emissions 70 persisting over decades (Jauhiainen et al. 2012; Hooijer et al. 2012). Accidental burning, due mainly to 71 small-scale land clearance fires getting out of control, can lead to very large peat losses over weeks or 72 months (e.g. in the exceptional El Niño year of 1997, CO₂ emissions from Southeast Asian peatland 73 fires were equivalent to 13–40% of total global fossil fuel emissions: Page et al. 2002). Crucially, whilst 74 a large proportion of lost tropical forest biomass can be recovered within decades in secondary-75 growth forest (Letcher and Chazdon 2009), restoration of peatland carbon stocks to pre-disturbance 76 levels would take thousands of years.

Despite their importance, basic information on carbon storage in tropical peatlands is lacking. This is
not just a problem in the tropics. Reviewing estimates of the size of the boreal peatland carbon pool,
Vasander and Kettunen (2006) showed that they vary by an order of magnitude (from 41.5 Pg C for all

histosols globally: Buringh 1984, to 455 Pg C for boreal and subarctic peatlands: Gorham 1991) due
mainly to substantial variation in estimates of peat dry bulk density (DBD) and thickness. The same
problem applies in the tropics but with the additional difficulty that, unlike in boreal peatlands, the
area of peat is also very poorly known in some regions.

84 Most research in the tropics has focused on Southeast Asian peatlands, and their distribution is comparatively well known, yet large uncertainties still exist in estimates of the amount of carbon 85 86 stored in them. For example, published figures for Indonesia include 37.2 (Wahyunto et al. 2003, 2004, 87 2006), 55 ± 10 (Jaenicke et al. 2008), and 57.4-58.3 Pg C (Page et al. 2011b), based on differing 88 methodologies. New field data can make a substantial difference: Dommain et al. (2014) estimated the 89 peatland carbon pool in western Indonesia (excluding Papua) at 23.1 Pg C, a substantial downward revision of the 33.3 Pg C estimate for the same area by Wahyunto et al. (2003, 2004) due mainly to 90 91 new data on DBD and peat thickness, even though the area of peat estimated by the two studies (which 92 are based on similar GIS datasets) was almost identical (131 500 and 129 700 km² respectively). 93 Peatlands in other tropical regions are less well mapped. For example, although wetlands in the 94 Cuvette Centrale of the Congo Basin, the fourth- or fifth-largest wetland on Earth (Campbell 2005), 95 have been mapped using remote sensing techniques (e.g. Bwangoy et al. 2010; Betbeder et al. 2014), 96 the extent of peat within these wetlands is essentially unknown; published estimates of the size of the 97 peatland carbon pool in the Congo basin are based on very few field data (Joosten et al. 2012). In 98 Amazonia, the existence of c. 40 000 km² of peat in the Pastaza-Marañón basin, Peru, has only recently 99 been confirmed by fieldwork (Lähteenoja et al. 2009a,b, 2012). The uncertainty in this peat carbon 100 pool is very large (0.8–9.5 Gt: Lähteenoja et al. 2012). Limited surveys have been carried out in parts of 101 Brazil (Lähteenoja et al. 2013) and French Guiana (Cubizolle et al. 2013), but remain to be conducted 102 elsewhere in the Amazon basin. A substantial revision of the global estimate of the size of the tropical 103 peatland carbon pool is therefore likely over the next few years. Data on carbon fluxes, especially of 104 the important greenhouse gas (GHG) methane, are likewise extremely scarce, especially outside 105 Southeast Asia.

The high carbon density of tropical peatlands makes them an obvious focus for emissionsmanagement schemes such as the UN REDD+ (United Nations Reducing Emissions from Deforestation and Forest Degradation) programme (Murdiyarso et al. 2010). This will require precise inventories of peatland carbon stocks and fluxes to be carried out for specific regions in order to provide accurate estimates of baseline carbon pools and fluxes against which projected reductions in carbon emissions can be measured, verified, and translated into carbon credits.

112 Therefore two fundamental research priorities are (1) to quantify the amount of carbon stored in 113 tropical peatlands accurately at a range of scales, and (2) to quantify fluxes of carbon to and from the 114 tropical peatland carbon pool. More broadly, a better understanding of the present distribution of 115 tropical peatlands, the processes of peat accumulation and decay, and the development of peatlands 116 over time is relevant to other important research questions, including:

- What determines the distribution of peat in the tropics?
- How will tropical peat stocks change in the future?
- How do tropical peatlands influence biogeography and biodiversity?
- How should tropical peatlands be managed?

In this paper we aim to encourage better and more consistent methodologies for producing carbon inventories and budgets, principally at a regional scale. The value of a coordinated approach is clear: for example, where long term, repeated and systematic carbon inventories of tropical forest biomass have been coordinated by the RAINFOR (Malhi et al. 2002) and AFRITRON networks (among others), many new insights into long- (Baker et al. 2004) and short-term (Lewis et al. 2011) carbon dynamics have emerged, generating over 100 carbon-focused publications by more than 200 collaborators.

The paper is divided into three parts, 1) mapping peat distribution, 2) estimating the size of the carbon pool, and 3) estimating carbon fluxes. In part 1, we focus on emerging remote sensing technologies which, alongside appropriate measurements on the ground, can improve our ability to map peatland extent. In part 2 we discuss the measurement of peat thickness, peat bulk density and carbon concentration. In part 3 we discuss carbon fluxes into and out of peatland ecosystems, specifically, long-term carbon accumulation rates, litter inputs and decomposition, gaseous and waterborne fluxes,
and above-ground carbon stocks. We focus on forested tropical lowland peatlands, although we
recognize that many of the issues we discuss also pertain to other types of peatland in the tropics and
beyond.

136 II Mapping peat distribution

Field mapping of peatlands is a considerable challenge, especially at regional, national and global scales. However, a combination of field measurements and inferences from remote sensing can provide an optimal balance, where realistic programmes of fieldwork can yield map-based products that cover a region comprehensively, are reasonably accurate and reliable, and with robustly quantified uncertainties.

142 Tropical peatlands are often distinct from surrounding terra firme (dry-land) forests in four ways that 143 are observable in satellite and airborne data. Firstly, their vegetation is often low in diversity. In South 144 and Central America, some parts of Africa, and on the island of New Guinea, palms are often more 145 abundant than in upland forests, sometimes forming mono-dominant stands (e.g. Lähteenoja and Page 2011; Wright et al. 2011). However, diversity in some peatland forests can be high (e.g. Sumatran 146 147 swamps: Brady 1997). Secondly, vegetation structure is often (but not always) distinctive, often with 148 more open canopies and low-stature or thin-stemmed trees, or no trees at all (e.g. Anderson 1983; 149 Page et al. 1999; Phillips et al. 1997). Thirdly, their topography can be distinctive. Tropical peatlands 150 typically occupy a specific topographic or geological setting, for example the subsiding Pastaza-151 Marañón basin in Peru or the peats forming along dendritic drainage channels in the Tasek Bera basin, 152 Malaysia (Wüst and Bustin 2004); blanket peats that are indifferent to topography only occur rarely in 153 upland settings (Gallego-Sala and Prentice 2012). Many tropical peatlands are also detectably dome-154 shaped (e.g. Phillips et al. 1997; Jaenicke et al. 2008; Lähteenoja et al. 2009b). Finally, peatland water 155 tables often lie close to, at or above the surface throughout the year (e.g. Lawson et al. 2014). Whilst 156 any one of these four features alone is insufficient to characterise an area of forest as potentially peat-157 forming, the combination of two or more presents a much stronger case (Draper et al. in press).

158 These properties can be mapped using a number of different remote sensing products. Compositional 159 and structural features of peatland vegetation have been distinguished using optical sensors such as 160 Landsat (Phua et al. 2007; Jaenicke et al. 2010; Li et al. 2010; Lähteenoja and Page 2011), SPOT 161 (Systeme Pour l'Observation de la Terre; Lee 2000; Miettinen and Liew 2010), and MODIS (Moderate Imaging Spectrometer; Langner et al. 2007; Wijedasa et al. 2012). Figure 2 presents an example of 162 163 vegetation classification of the Changuinola peat dome in San San Pond Sak, Panama, using multi-scale 164 Landsat Thematic Mapper (TM) image analysis supported by aerial photography and field data to 165 characterise the main vegetation gradient. To date, optical imagery from medium spatial resolution 166 sensors such as the Landsat series (30 m multispectral imagery) has been the primary and most successful tool for mapping peatlands. The new generation of VHR (very high resolution) products, 167 168 such as IKONOS (4 m multispectral) or WorldView-2 (2 m multispectral) imagery, potentially brings 169 new opportunities for detailed and accurate vegetation mapping. One key difficulty in the tropics is the 170 infrequent temporal coverage of these sensors, which makes cloud-free images difficult to obtain. 171 RapidEye (5 m multispectral) products, derived from a constellation of five satellites with 172 consequently more frequent image acquisition, can be a more reliable source of cloud-free imagery 173 (e.g. Franke et al. 2012). Opportunities are likely to grow further as new sensors are launched (e.g. 174 1.24 m resolution WorldView-3) and future missions become operational (e.g. the EU's Sentinel-2 10 175 m resolution twin satellites),

176 Active sensors such as radar and LiDAR (Light Detection And Ranging) which penetrate the canopy 177 can be used to detect the distinctive forest structures (e.g. combination of low canopy, thin stems, high 178 stem density in pole forest) or patterns in structure (e.g. concentric zonation of vegetation 179 communities) that characterize some peatlands. Radar and LiDAR are also able to provide topographic 180 data which can help to distinguish between peatland and *terra firme* forests, but few attempts have 181 been made to identify tropical peatlands using these tools. One useful exception is the use of an orbital 182 LiDAR instrument onboard the ICESAT satellite to measure peat topography and forest biomass at a 183 study site in Kalimantan (Ballhorn et al. 2011; other examples include Hoekman and Vissers 2007; 184 Rakwatin et al. 2009; Jaenicke et al. 2008; Jubanski et al. 2013). A limitation of LiDAR products is that 185 they are generally available as discrete point measurements (orbital sensors) or thin strips of data

(aerial sensors) rather than a full coverage. Other data sources (e.g. L-band radar data and groundbased measurements of forest structure) are usually needed to interpolate between LiDAR
measurements (Mitchard et al. 2012).

The presence of standing water below a forest canopy can produce a distinctive radar backscatter signal, particularly at longer radar wavelengths. L-band radar has been used extensively in tropical contexts to map standing water (de Grandi et al. 1998, 2000; Hess et al. 2003; Hoekman 2007; Bwangoy et al. 2010), to track changes in floodwater extent (Rosenqvist and Birkett 2002; Alsdorf 2003; Jung et al. 2010; Lee et al. 2011; Betbeder et al. 2014), and, using time series of radar data, to distinguish areas that are constantly wet from those that are only seasonally wet (Waldram 2014).

195 Remote sensing therefore already provides effective tools for extrapolating from field measurements 196 to map peatlands over large areas, but there is scope for further methodological research. The use of 197 optical sensors to define peatland extent has been widely and successfully implemented, but the less 198 commonly used active sensors may discriminate between peatland and terra firme more effectively in 199 many circumstances. We recommend that wherever possible, multiple remote sensing products 200 (including both optical and active sensors) should be used in combination. Some of the data types 201 discussed here, such as airborne LiDAR, are too expensive to obtain for many projects. However, 202 suitable products for mapping peatlands based on vegetation structure and composition, topography 203 and inundation at a coarse scale (c. 30 m) are freely available from Landsat, SRTM (Shuttle Radar 204 Topography Mission) and ALOS PALSAR. A combined approach using these products is a highly 205 feasible starting point for future mapping projects.

It remains the case, however, that interpretations of peatland distribution based on remote sensing need to be validated with ground reference data. Ground reference points are often difficult and expensive to install and therefore require careful planning in order to maximise their usefulness. They need to meet many criteria: they should be widely distributed across the study region, encompassing the full range of spatial heterogeneity; they need to span the full range of environments present in the landscape (including anthropogenic ones such as rice fields and oil-palm plantations: Miettinen et al. 2012), not just those associated with peat; and the reference dataset should be sufficiently large both 213 to develop and validate a classification using different subsets of the data. Typically, hundreds of 214 points are required to develop a reliable classification for a particular region, so it would clearly be 215 desirable if researchers visiting new sites could collaborate in generating suitable data. In Table 1 we 216 recommend a set of measurements that are needed for synoptic mapping and ground reference, and 217 which can easily and quickly be collected as part of the basic site description for any kind of research 218 on tropical peatlands. More extensive and specialized measurements (e.g. detailed vegetation 219 composition, water table depth variation) may be necessary for more specific applications of remote 220 sensing, but wherever possible they should be carried out in a way that preserves the compatibility of 221 measurements between studies, in line with Table 1.

222 III Carbon stocks

The quantity of carbon stored by peatlands in a region (M_c , in kg) can be calculated as:

$$224 M_c = AD\rho c (1)$$

where *A* is the area of peatland (m²), *D* represents its mean thickness (m), ρ is the mean DBD (kg m⁻³) and *c* is its carbon concentration (dry mass proportion; Gorham 1991). The area of peatland can be determined by a combination of remote sensing and ground survey, as discussed in Section II. The remaining three variables on the right hand side of equation (1) are each susceptible to considerable uncertainty, which in combination can lead to very large uncertainties in M_c.

230 1 Peat thickness

In carbon inventory research it is usually convenient to treat peat as a separate, particularly carbonrich category of soil, but a long-standing problem in peat research, unlikely to be resolved any time soon, is how to define 'peat'. Most workers define a peat soil as one that contains more than a certain proportion of organic matter, but the critical value varies widely, between about 30 and 65 wt% organic matter (Joosten and Clark 2002:41; Wüst et al. 2003), hindering data synthesis (Page et al. 2011b). Some peat units have a clearly-defined contact with the underlying, less organic material, but others change in composition more gradually. In such cases it can be impossible to judge peat thickness consistently in the field, and unfortunately, core samples are not always taken for laboratory analysis of organic matter content, resulting in unreliable data. Researchers must acknowledge that 'peat' is *de facto* a flexible term, and circumvent definitional issues by collecting objective data on the properties of the material they are studying. One way forward, and our recommendation, is to build on past efforts such as the CARBOPEAT project (http://www.geog.le.ac.uk/carbopeat) to compile the necessary data (core location, sample depth, and sample loss-on-ignition and carbon concentration) to allow reanalysis using a standardized definition of 'peat'.

245 Estimates of peat thickness for a region are usually based on limited numbers of measurements, which 246 may be biased by over-reporting of the thickest deposits. The geometry of peatlands is such that thick 247 peats are often restricted to quite small areas (the centre of domed mires or basin-filling swamps) 248 surrounded by much more extensive areas of shallow peat. Simply taking the mean of a series of 249 measurements from the edge to the centre of a single peatland is likely to result in an overestimate of 250 the mean peat thickness. A much more robust approach is that taken by Dommain et al. (2014), 251 building on existing detailed GIS maps of Indonesian peatlands by Wahyunto et al. (2003, 2004, 2006). 252 By interpolation between field measurements, they defined GIS polygons of small areas of peat of 253 different mean thicknesses. The total volume of peat was then derived by multiplying the area of each polygon by its specific thickness. This provides a useful template for future work, not least because the 254 255 link between field data and estimates of carbon storage is explicit, facilitating revisions as further data 256 emerge. GIS datasets can also be readily shared and assimilated into larger-scale mapping or 257 reanalysis projects, and are thus a very desirable form of research output.

There are few detailed studies of the geometry of individual tropical peatlands, but the exceptionally detailed survey using a dense grid of 194 depth estimates across a 235 ha swamp, the CICRA peatland in southern Peru (Householder et al. 2012), demonstrates that useful insights can be gained that can guide more representative sampling of other peatlands in the region. In this case, comparison of the peatland volume estimate derived from the entire network of points showed that less detailed estimates of the volume of the same peatland based on short transects tended systematically to

overestimate the total peat volume. This 'calibration' was then used to adjust volume estimates basedon transect data only from other sites in the region.

Ground penetrating radar (GPR) could also be used alongside manual coring to determine peat thickness and stratigraphy, and may be especially useful for studying features such as voids which can be important in volume terms in forested peatlands (e.g. Slater and Reeve 2002; Parry et al. 2014). There are, however, considerable practical difficulties associated with deploying relatively bulky GPR over large distances in forested peatlands, though smaller instruments are currently in development.

271 Few attempts have so far been made in the tropics to derive peat thickness by remote sensing, and 272 they are exceptional cases. For example, Jaenicke et al. (2008) integrated Landsat Enhanced Thematic 273 Mapper (ETM+) and SRTM data, a network of 750 field measurements of peat thickness, and a three-274 dimensional peatland development model to estimate the volume of domed peatlands in Kalimantan; 275 the success of this project depended on the rather special properties of the mires in question (regular 276 shape, very thick peats). Ballhorn et al. (2009) also used LiDAR measurements to estimate changes in 277 peat thickness over time due to burning on Indonesian peatlands. A more generally applicable method 278 for measuring peat thickness remotely is perhaps unlikely to emerge but there is scope for further 279 investigation on a site-by-site or region-by-region basis. Peat thickness can sometimes correlate with 280 other properties that are visible by remote sensing. For example, thick peats often occur towards the 281 centre of ombrotrophic peat domes. Field observations suggest that these deep, nutrient-poor peats 282 are frequently (but not always) associated with specialized and structurally distinctive vegetation 283 communities in, for example, Kalimantan (Page et al. 1999), Panama (Sjögersten et al. 2011), Peru 284 (Kelly et al. 2013; Draper et al. in press), and the Republic of Congo (G. Dargie unpublished data). 285 These plant communities are often distinctive in Landsat TM, ALOS PALSAR and other imagery. 286 However, apparent relationships between remote sensing data and peat thickness must be confirmed 287 using empirical data because many factors other than peat thickness may be equally or more 288 important in controlling vegetation composition and structure (Draper et al. in press), as is the case in 289 temperate/boreal peatlands (Wheeler and Proctor 2000).

290 2 Bulk density

291 Measured values of dry bulk density (DBD) for individual peat samples from Indonesia vary by almost 292 an order of magnitude, and the few available data from undisturbed sites in Amazonia vary by a factor 293 of two (all in peats with ash contents <10%; Figure 3). This variability, which arises from factors 294 including the botanical composition of the peat, consolidation of deeper peats, drainage history, and 295 measurement method, is a major source of uncertainty in estimates of the size of the peatland carbon 296 pool because calculations of carbon stocks at individual sites, or even across regions, are frequently 297 based on the mean of a very small number of DBD measurements (e.g. Page et al. 2011b; Householder 298 et al. 2012).

299 Frequently, the number of samples taken may be too small, and/or the spacing between samples may 300 be too large to capture the spatial variability of peat DBD. Within-site lateral variation in DBD has not 301 been explored systematically in tropical contexts and more work is needed to establish whether there 302 are any predictable patterns. Sometimes the stratigraphic profile of DBD is quite consistent at sites 303 within a region (Hooijer et al. 2012), but DBD can also vary systematically between peatland types 304 within a region (e.g. floodplain peatlands, domed peatlands; Shimada et al. 2001). The converse has 305 been shown in boreal peatlands (i.e. DBD varies between regions within the same peatland type; 306 Sheng et al. 2004; Yu 2012; there are insufficient data to know if this also applies in the tropics). Site-307 specific measurements are therefore always desirable, and in general more data are needed to 308 determine whether DBD varies spatially in a predictable way.

A further complication is that tropical peatlands can show considerable stratigraphic variation in DBD (Figure 4) due to fluvial mineral inputs (Lähteenoja et al. 2009b), long-term vegetation succession and related variations in peat structure (Phillips et al. 1997; Roucoux et al. 2013), peat decomposition, post-drainage consolidation (Hooijer et al. 2012), and water- or gas-filled voids. This stratigraphic variation can only be addressed through field measurements and ample down-core sampling. A greater palaeobotanical insight into the origins of variation in DBD in tropical peats would also be a useful line of research.

316 A second potential source of error in DBD estimation is that peat samples of known volume must be 317 recovered, which is difficult to achieve reliably. One method for collecting volumetric samples is to dig 318 a pit into the peat and extract a monolith from the pit wall (Hooijer et al. 2012; Couwenberg and 319 Hooijer 2013), but this may entail continuously pumping water out of the pit which can be impractical, 320 is limited to shallow sections, and, by analogy with what is known of the effects of seasonal changes in 321 water table on peat volume in undisturbed peats (Price 2003), may lead to compaction and over-322 estimation of DBD. Various specialised corer designs have been proposed to improve the collection of 323 volumetric samples from shallow peats (Wright et al. 1984; Givelet et al. 2004; van Asselen and 324 Roosendaal 2009; see De Vleeschouwer et al. 2010 and Glaser et al. 2012 for discussion of the relative 325 merits of different devices). At present, most workers use a Russian-type corer which is suitable for 326 use in shallow and deep peats; as a side-sampling device it offers better control over sample depths 327 than a piston corer and a low risk of sample contamination than a Hiller corer, which is especially 328 important if the peat is to be radiocarbon dated (cf. Glaser et al. 2012), but can be less effective at 329 cutting through woody peats than a corer with a serrated barrel. The volume of the *in situ* peat sample 330 is usually assumed to be identical to the internal volume of the corer. In reality, core recovery is often 331 imperfect, especially in fibrous or woody peat that cannot be cut cleanly, or where the peat is 332 structurally weak and is not retained within the corer, or (in the case of piston corers) it fails to fill the 333 barrel (Wright 1991; Dommain et al. 2011; Lähteenoja et al. 2013). Thus, DBD measurements are 334 probably often subject to large errors stemming from erroneous volume estimations. Russian-type corers in particular may yield systematic underestimates of DBD (Clymo 1983). In general, large-335 336 volume corers (including wide-diameter Russian corers) are to be preferred over smaller devices 337 because they will more likely retrieve a representative sample, but they can be impossible to use in 338 stiff or woody peats and are more logistically problematic compared with a smaller, lighter device. As 339 yet, there have been few systematic comparisons of different methods to assess the extent of the 340 uncertainty in volume measurements (Pitkänen et al. 2011), especially in fibrous and woody peats. 341 More research on this topic could help to quantify and, perhaps, compensate for any differences 342 between datasets that are due to the use of different sampling devices.

A third way in which DBD measurements made by different research groups may vary stems from variation in laboratory methods, for example in the temperature at which the peat samples are dried. Chambers et al. (2011) proposed a protocol for measuring DBD and other basic variables (including drying at 100°C) which we recommend and which, if followed, will minimize this uncertainty.

347 3 Carbon concentration

Two principal approaches are used to estimate carbon concentration in peats. The more accurate and 348 349 direct technique is elemental analysis (Nelson et al. 1996; Chambers et al. 2011). In peats with low ash 350 content the carbon concentration calculated by this method typically varies between about 52 and 351 58% (Figure 5). However, many workers use mass loss-on-ignition (LOI; Heiri et al. 2001) as a cost-352 effective way to estimate organic matter concentration. The LOI at (typically) 450°C is assumed to be 353 attributable to combustion of organic material; the remainder, the 'ash', is typically composed of 354 sedimentary mineral material and biogenic silica. Carbon concentration can then be estimated by 355 assuming that the organic material contains (e.g.) 50 wt% C (Turunen et al. 2002).

356 On the basis of available data, applying the LOI-based approach described by Turunen et al. (2002) 357 could apparently systematically underestimate carbon concentration by c. 8 wt% in tropical peats 358 with very low ash contents (Figure 6). This disparity is principally attributable to the varying 359 abundance of carbon-rich compounds, such as lignin or charcoal, in the organic matter fraction of peat. 360 However, at least at some sites, the paired measurements show a strong linear relationship, albeit with 361 some scatter. There may even be a strong linear relationship between DBD and carbon density (the product of carbon concentration and DBD, measured in e.g. kg C m⁻³), sufficient to support the 362 suggestion that DBD measurements alone may be sufficient for a first-order estimate of carbon 363 364 concentration (Warren et al. 2012; Farmer et al. 2013). However, the strength of this relationship 365 varies and it has not been tested for tropical peats outside a few locations in Southeast Asia.

Therefore, for inventory purposes, and especially when undertaking work for the first time at a new site, in our view it remains important to measure carbon concentration as accurately as possible, i.e. using an elemental analyser following Chambers et al. (2011). LOI remains a useful tool in its own right because it provides a direct measure of the organic matter content of the peat, which is valuable
in understanding the developmental history of a site. As with DBD, systematic studies of within- and
between-site variation in carbon concentration are lacking in tropical contexts.

372 The question of the number of samples to take, both horizontally and vertically, for DBD and carbon 373 concentration in a peatland is inevitably constrained by the resources available for a given project. One 374 approach to designing a sampling strategy is to assess the way in which the variation in different 375 measurements (DBD, carbon concentration, peat thickness) contributes to the overall uncertainty in the estimate of the carbon pool. For example, peat thickness is usually much more variable across a 376 377 region than DBD or carbon concentration, suggesting that a rational use of research effort would be to 378 focus on measuring peat thickness. Resampling techniques can be used to estimate the confidence 379 interval around an estimate of the regional carbon pool for a given region (Manly 2007; Draper et al. in 380 press). Chimner et al. (2014) discussed the relative merits of different core sub-sampling approaches 381 in terms of attempting to encompass stratigraphic variation in DBD and carbon concentration in 382 Canadian peats and found that, in Canada, (a) several different approaches gave similar results and (b) 383 analysis of the DBD and carbon concentration of a single core section, from 25–75 cm depth at each 384 site, gave estimates of the total peat carbon stock that were within 15% of estimates based on exhaustive sampling of entire cores from the same sites, suggesting that even a single (admittedly 385 386 large) sample from each core site may be adequate for inventory purposes. This conclusion must be 387 tested before being applied in other regions where, for example, frequent admixture of clay in deeper 388 peats may give very different results. In general we would recommend a more conservative approach, 389 taking several discrete subsamples throughout the full thickness of the peat (the "intermittent peat 390 sampling method" described by Chimner et al. 2014).

391 **4** Biomass

A widely-used set of standard protocols has been developed for measuring above-ground biomass (AGB) in *terra firme* tropical forests (e.g. Phillips et al. 2009). These protocols are, with modification, applicable in forested peatlands. They should be used wherever possible because using the same protocols on peat and *terra firme* enables biomass, productivity, diversity, and other key vegetation 396 parameters to be compared, which means that peatland vegetation can be understood in the broader 397 context of tropical vegetation as a whole, and can be integrated into regional assessments of AGB 398 across all ecosystems.

Modifications, or rather additions, to *terra firme* protocols for use on peatlands are necessary because peatland vegetation is frequently dominated by plants that are usually regarded as negligible in a standard forest census, for example, thin-trunked trees with diameter at breast height (DBH) < 10 cm (the cut-off used in most AGB inventories), or grasses and sedges. An example of a modified protocol that remains compatible with standard RAINFOR protocols is provided by Roucoux et al. (2013) at Quistococha, Peru: a nested sampling design was used to record small trees with DBH between 2.5 and 10 cm in a series of sub-plots within their main census plot.

406 Another consideration is that peatland forests are often dominated by species (such as trunkless 407 palms) to which standard allometric equations for calculating biomass do not apply (e.g. Chave et al. 408 2005; Feldpausch et al. 2011; Gehring et al. 2011). Outside Southeast Asia, the allometric equations 409 used for AGB calculation of tropical peat swamp forests in recent literature (e.g. Kronseder et al. 2012; 410 Englhart et al. 2013) were originally developed for "moist tropical forests" in general (Chave et al. 411 2005), and the performance of these models in the distinctive pole and palm forests often found on 412 peatlands needs to be tested. Recently, species-specific equations have been developed for the most 413 important peatland palms in South America including Mauritia flexuosa and Mauritiella armata 414 (Goodman et al. 2013). Similar work remains to be carried out on many other species. The long history 415 of economic exploitation of peat swamp forest in Southeast Asia means that allometric equations there 416 are better established (Krisnawati et al. 2012).

New tools for assessing AGB on large spatial scales by remote sensing of canopy structure are rapidly developing (e.g. in Southeast Asian peatlands: Jubanski et al. 2013; Kronseder et al. 2012). Airborne LiDAR (Asner et al. 2013) and satellite-based LiDAR and L-band radar (e.g. Saatchi et al. 2011; Ballhorn et al. 2011; Mitchard et al. 2012; Baccini et al. 2012) have been widely used to estimate AGB, mainly in *terra firme* forests. As with mapping, more ground-reference data are needed to allow these approaches to be applied confidently to peatland forest AGB assessments (Table 1).

423 Below-ground biomass (BGB: living roots, as opposed to the necromass or dead material contributing 424 to peat) has hardly been studied in tropical peats, but existing research shows that root inputs can be 425 more important to peat accumulation than leaf and stem litter, i.e. these are 'replacement peats' 426 (Brady 1997; Chimner and Ewel 2005; Dommain et al. 2011). Measuring BGB is fraught with 427 methodological problems even in terra firme forests, but it would be of great interest to know (in the 428 context of understanding carbon balance and rates of carbon sequestration - see below) how much of 429 the peat at various depths is made up of live or recently dead root material. However, for the purposes 430 of estimating the size of carbon pools, the live and dead components of the peat do not need to be 431 separated explicitly.

432 IV Carbon fluxes

433 In peatlands, carbon enters the peat in the form of litter and leaves it as dissolved organic carbon 434 (DOC), particulate organic carbon (POC), and as the greenhouse gases CH_4 and CO_2 . Comprehensive 435 carbon flux measurements have been made for several well-studied northern peatlands (Roulet et al. 2007; Nilsson et al. 2008; Koehler et al. 2011), but a complete carbon budget has only been attempted 436 437 once in tropical peat swamp forest (Chimner and Ewel 2005). The limited available data suggest that 438 tropical peatland carbon sequestration rates are towards the upper end of the range for peatlands 439 globally (Mitsch et al. 2010; Dommain et al. 2011; Glaser et al. 2012) due to a combination of high net primary production (NPP) and low decomposition rates (Dommain et al. 2011; Sjögersten et al. 2014). 440

441 The annual change in organic carbon for a peatland (ΔC_{org}) can be expressed as follows, following 442 Roulet et al. (2007):

443
$$\Delta C_{org} = NPP - F_{CO_2} - F_{CH_4} - netDOC_{EX} - netPOC_{EX}$$
(2)

444 where ΔC_{org} is equivalent net primary production (NPP) minus F_{CO_2} (the gaseous flux of CO₂), F_{CH_4} 445 (the gaseous flux of CH₄), *netDOC*_{EX} (the waterborne DOC flux), and *netPOC*_{EX} (the waterborne POC 446 flux). 447 None of the quantities on the right hand side of equation 2 are easy to measure in any peatland, but in 448 the tropics, NPP is especially difficult to quantify due to the large size of the plants on forested peatlands. The carbon flux can perhaps more easily be estimated using (i) the observed change in peat 449 450 surface height, caused by accumulation and/or subsidence, relative to a fixed stake (e.g. Nagano et al. 451 2013; Couwenberg and Hooijer 2013); (ii) dating the basal peat to establish the apparent long term 452 rate of carbon accumulation (LORCA: Clymo et al. 1998; Turunen et al. 2002; 'long term' in this context 453 means centuries to millennia) as a direct measure of carbon sequestration over a given period of time 454 (although this can differ substantially from present rates of carbon accumulation: Joosten and Clarke 455 2002:34); or (iii) measured rates of litter production and carbon losses through decomposition 456 (Chimner and Ewel 2005). Our discussion below focuses on the prospects for the direct measurement 457 of litter input and decomposition, greenhouse gas fluxes (chiefly CO₂ and CH₄), and waterborne carbon 458 fluxes, all of which are needed both to quantify the carbon flux in tropical peatlands and in order to 459 develop a fuller mechanistic understanding of the controls on their carbon balance and greenhouse gas 460 fluxes.

461 *1 Litter inputs*

462 The transformation of different types of litter in tropical forested peatlands – roots, non-woody leaf 463 and stem litter, and woody debris ranging from twigs and small branches to whole tree trunks - into 464 peat is poorly understood (Tie and Esterle 1992; Brady 1997; Sulistiyanto 2004; Chimner and Ewel 465 2005; Shimamura and Momose 2005). The quantity of litter inputs and the rate at which each litter 466 type decomposes determine their contribution to the peat carbon pool. Leaves can comprise the bulk 467 of above-ground litterfall in peat swamp forests (Sulistiyanto 2004) but they typically decompose 468 much more rapidly than woody debris and roots and therefore contribute little to the overall 469 accumulation of peat (Chimner and Ewel 2005), except perhaps where leaves accumulate in ponds on 470 the peat surface (Gastaldo and Staub 1999).

Above-ground litterfall can be collected and weighed using nets of e.g. 1×1 m, depending on the size of
the litter, although the logistics of the necessarily frequent sampling (due to high litter decomposition
rates in the traps) can be restrictive. Suitable protocols have been developed by the CTFS Global Forest

474 Carbon Research Initiative (www.ctfs.si.edu) and Harrison (2013); comparable sampling schemes 475 have also been used in lowland tropical rainforest (Chambers et al. 2000; Nepstad et al. 2002). The 476 contribution of large woody debris is harder to measure, and is almost unexplored in tropical 477 peatlands. The frequency and relative importance of branch- and tree-fall events can be estimated by 478 repeated litter surveys along transects (Waddell 2002) or in census plots (Chimner and Ewel 2005; 479 Woodall and Monleon 2008; Baker and Chao 2011), although no standard method has yet been agreed 480 upon (Larjavaara and Muller-Landau 2011); further methodological research is needed.

481 Equally, very few data on root dynamics are available for tropical peatlands. Root growth, especially of 482 fine roots, can be measured using the ingrowth core or bag method (e.g. Symbula and Day 1988; Neill 483 1992; Brady 1997; Finér and Laine 1998; Metcalfe et al. 2008), though this can be problematic: 484 important considerations are the linearity of root growth over the study period and temporal variation 485 in root production, e.g. between dry and wet seasons (Metcalfe et al. 2008), as well as changes in soil 486 structure caused by the removal of roots from the soils during preparation of the ingrowth core, which 487 may affect later root growth. Root mortality (i.e. root litter input to the necromass) is extremely 488 difficult to determine and involves differentiating between live and dead root material (Finér and 489 Laine 1998). Alternative approaches, such as the use of minirhizotrons which allow in situ 490 measurements of root growth and mortality (Iversen et al. 2012), have been used successfully in 491 lowland tropical rainforest (Metcalfe et al. 2007) but not yet in tropical peatlands.

492 2 Litter decomposition

Potential *in situ* litter decomposition rates in tropical peatlands appear higher than in temperate/boreal peatlands (Brady 1997; Chimner and Ewel 2005; Yule and Gomez 2008), presumably due mainly to the year-round higher ambient temperatures, although other factors including litter composition, water table depth, and pH may also be involved (Qualls and Haines 1990; Chimner and Ewel 2005; Yule and Gomez 2008). Few systematic studies have been carried out, especially outside Southeast Asia.

499 Standard techniques for measuring the decomposition rate of fine litter are readily applied in tropical

500 peatlands, in which rapid decomposition allows meaningful results to be obtained over sampling 501 periods as short as two years (Yule and Gomez, 2008; Hoyos 2014). Mesh bags containing known dry 502 weights of litter are firmly anchored to the ground to prevent them being washed away during floods, 503 or are buried in the peat and collected at regular intervals to calculate weight loss (Chimner and Ewel 504 2005; Wright et al. 2013a; Hoyos 2014). Decomposition of large woody debris such as fallen trees or 505 large branches is also presumably important in forested peatlands, but sampling is more difficult, 506 usually requiring repeated surveys over long intervals, and data are lacking. One approach would be to 507 follow (or build on) RAINFOR protocols to establish permanent forest census plots, which include a 508 protocol for measuring coarse woody debris (Baker and Chao 2011). Census plots that are part of a 509 network such as RAINFOR, where the data are relevant to many research questions, are more likely 510 than not to be revisited over many years and hence to generate the necessary long-term datasets.

511 *3 Greenhouse gas fluxes*

To date, most data on GHG fluxes from tropical peatlands have been collected during daylight hours using static sampling chambers placed on the peat surface (Jauhiainen et al. 2005, 2008, 2012; Melling et al. 2005a, b; Sjögersten et al. 2011; Wright et al. 2011, 2013b). Automated sampling has not yet been widely adopted but is becoming more common (Sundari et al. 2012; Hirano et al. 2014). Measured GHG fluxes from peat swamp forests vary greatly both across the tropics (Sjögersten et al. 2014) and within sites (Wright et al. 2013b), often correlating with mean annual water table depth (Couwenberg et al. 2010).

519 Substantial temporal variation (diurnal and seasonal) poses a major challenge to obtaining reliable 520 estimates of GHG emissions. Current data suggest that temporal variation exceeds the spatial variation 521 between forest types (Wright et al. 2013b), and that there is frequently a strong correlation between 522 GHG efflux and temporal variation in water table depth (Jauhiainen et al. 2005; Hirano et al. 2009, 523 2014; Sundari et al. 2012). Long-term data (e.g. over more than a year) obtained at regular (e.g. 524 monthly) intervals are scarce, so the magnitude of intra- and inter-annual variation in fluxes is unclear. Strong diurnal variation in CO₂ and CH₄ efflux (e.g. Hirano et al, 2009; Wright et al. 2013b; Hoyos 525 2014) means that measurements intended for estimating net GHG fluxes need to include 526

527 measurements at 2–4 hourly intervals or better. The effect of weather (e.g. rainfall events) and 528 seasonality (e.g. dry/wet season contrasts) on GHG emissions, and the importance of ebullition in 529 methane emission, also remain to be thoroughly investigated. Automated samplers capable of frequent 530 (sub-hourly) measurements over long periods (many months; e.g. Goodrich et al. 2011) can help to 531 resolve these issues.

Alternatively, the eddy-flux correlation approach can be used to acquire gas flux data with high temporal resolution, spatially integrated over hundreds of metres. Eddy-flux systems are currently under-used in tropical peatlands, with data presently only available from sites in Narathiwat Province, Thailand (Suzuki et al. 1999) and Kalimantan, Indonesia (Hirano et al. 2009, 2012). Additional systems will shortly become operational in Brunei, Sarawak (Malaysia), and Peru.

537 Additional uncertainty in estimates of peat decay rate arise because most CO₂ flux data from tropical 538 peatlands do not separate autotrophic (from roots) and heterotrophic respiration (from decomposing 539 peat) which makes it difficult to use peat surface measurements of GHGs to assess peat decay rates 540 (Page et al. 2011a). In plantations with regularly spaced trees and little ground cover, attempts have 541 been made to distinguish between soil and root respiration by measuring CO₂ release within and 542 between rows of trees (Jauhiainen et al. 2012). Alternative approaches to separating autotrophic and 543 heterotrophic respiration are (i) to compare the CO₂ release from buried mesh collars which restrict in-growth of roots, with collars which allow roots to grow in (Nottingham et al. 2011), or (ii) using a 544 545 trenching approach (Mäkiränta et al. 2008), which involves isolating a patch of ground from root 546 influences by cutting/digging through the roots around the plot (Mäkiränta et al. 2008), although this 547 may affect the peat moisture status in trenched plots, with implications for CO₂ fluxes.

A further uncertainty regarding net GHG emissions is related to the pathway for CH_4 emissions through tree stems, which a study by Pangala et al. (2013) at a peatland in Borneo found to account for a very large proportion (62–87%) of ecosystem CH_4 emissions. The extent to which this pathway is generally important needs to be established by comparable studies at other sites.

552 4 Waterborne carbon fluxes

553 Waterborne carbon fluxes (DOC and POC) from tropical peatlands represent a major source of uncertainty in their overall carbon balance. In temperate peatlands, waterborne carbon typically 554 555 accounts for c. 10% of total carbon export (Limpens et al. 2008). In the few studies in which DOC and 556 POC fluxes from tropical peatlands (all in Southeast Asia) have been measured (Yoshioka et al. 2002; 557 Baum et al. 2007), they have been found to be approximately double those from temperate peatlands 558 (IPCC 2014). Moore et al. (2011) estimated that the peat-covered part of Indonesia alone was 559 responsible for 10% of global fluvial DOC export to the ocean. Waterborne fluxes may be especially 560 significant in degraded peatlands where the forest vegetation has been removed and the peat has 561 destabilised (cf. Moore et al. 2013), and in floodplain peatlands where fluvial erosion can remove large 562 blocks of peat *en masse* during floods. Further quantification of these processes is needed.

563 Total export of waterborne carbon can most easily be estimated for peatlands which have a clear 564 hydrological boundary and discrete outflows, by measuring DOC and POC concentrations in drainage 565 streams regularly during annual or longer periods, along with the total water discharge (Billett et al. 566 2004; Moore et al. 2011, 2013). The achievable temporal resolution of measurements is a critical 567 limitation. Woody tropical peats often have high saturated hydraulic conductivity in their near-surface 568 layers, but below-ground flow is frequently insufficient to shed the large and sporadic inputs from 569 rainfall (Kelly et al. 2013); evapotranspiration and, especially in the wet season, surface runoff play a 570 large role in the hydrological budget. The pathway taken by water as it leaves a peatland affects its 571 DOC and POC load. Water which is shed rapidly through surface runoff may have a low DOC concentration due to its shorter residence time, but equally, especially in degraded peatlands, rapid 572 573 runoff may cause peatland erosion and carry a greater POC load. In such hydrologically dynamic 574 peatlands, discreet pulses of DOC and POC losses may be missed unless monitoring is carried out very 575 frequently. In a recent study, Moore et al. (2013) focused sampling effort on the peak of the dry season 576 and wet season, taking measurements every week during these periods. For the rest of the year, they 577 took samples every fortnight. We recommend this as a minimum sampling resolution for future 578 studies, taking into account the difficulty of regular sampling in many tropical peatlands; where

579 sampling can be undertaken more intensively (for example, using automated samplers) then this 580 should be attempted. Further research on tropical peatland hydrology (e.g. Kelly et al. 2014) leading to 581 more reliable hydrological models would also help in estimating waterborne carbon fluxes.

582 V Conclusions

583 Tropical peatland research with a focus on their role as carbon stores, sinks and sources is becoming 584 an increasingly active field, and an important one in relation to management of the global carbon cycle. 585 In this review we have identified many research needs, including methodological problems, and have 586 suggested some approaches to tackling them. In our view, however, the overarching need is for a more coordinated approach to data collection and sharing. This is necessary to allow us to address the most 587 588 fundamental, large-scale questions about how much peat exists in the tropics, and where it is; and 589 about the role of tropical peatlands in the global carbon cycle, today and in the future. Our main 590 conclusions and recommendations are as follows:

591 1. Tropical peatland research would benefit from a network of sites where basic measurements have 592 been made using identical methodologies. A precedent exists in the well-organised, extensive 593 networks of permanent tropical forest census plots (e.g. RAINFOR and AFRITRON). Where practical, 594 methods should be compatible with those used for peatland research outside the tropics, but perhaps 595 more importantly, they should be compatible with methods used in other tropical ecosystems 596 (particularly terra firme forest), in recognition of the fact that effective management depends 597 primarily on being able to compare the relative costs and benefits of managing peatlands and other 598 ecosystems in the same region for carbon storage and other ecosystem services. Table 1 proposes a set 599 of measurements which are cheap and practical to implement as part of basic site description, and 600 which would help to build a pan-tropical dataset that would put regional and global estimates of 601 tropical peatland carbon stocks and fluxes on a firmer footing.

2. Concerted effort to focus research on particular sites, drawing on both the social and the natural
sciences, has proven successful at one tropical peatland (Allen et al. 2005; Chimner and Ewel, 2005;
Drew et al. 2005). We would like to see the research community continue to build on this collaborative

and interdisciplinary approach by establishing a small number of keystone sites where a rich body of
knowledge can be accumulated over time. This would facilitate the testing of conceptual and numerical
models of peatland processes, and would help to build long-term datasets that can be used to analyse
temporal variability in peatland behaviour.

609 3. As in many other fields, it would be helpful if data were routinely published in full, in tabular form in papers (as supplementary data if necessary) or in appropriate data repositories such as the Carbon 610 611 Dioxide Information Analysis Center (http://cdiac.ornl.gov/) or the UK Environmental Information 612 Data Centre (http://www.ceh.ac.uk/data), in order to facilitate reanalysis and synthesis. Full 613 publication of data is increasingly required by grant funding bodies. The fact that this so rarely 614 happens at present suggests that mechanisms are needed to incentivise sharing of data between 615 researchers. Again, existing networks such as RAINFOR (Malhi et al. 2002) offer precedents to follow, 616 in terms of 'ground rules' that incentivise data sharing by guaranteeing opportunities for co-617 authorship of any publications that result.

4. There is, separately, a need for a community-wide data synthesis project to build a GIS-compatible
database on carbon storage in tropical peatlands (and indeed, peatlands globally) that would facilitate
inter-site comparisons.

5. Throughout this review we have identified research priorities which, if addressed, would improve
our ability to make reliable measurements and to extrapolate from point measurements to regional
and global assessments of peatland carbon stocks and fluxes. These include:

- a. Studying the relationships between peat properties, the overlying vegetation, and their remote
 sensing signatures;
- b. Developing radar/LiDAR techniques for mapping tropical peatlands and measuring AGB;
- 627 c. Investigating the use of multiple remote sensing methods in combination in mapping tropical628 peatlands;

d. Collaboratively developing large ground reference point datasets to support remote sensing;

e. Investigating further the potential for inferring peat thickness by GPR and remote sensing;

- 631 f. Systematically comparing different volumetric peat sampling methods;
- 632 g. Investigating the spatial and stratigraphic variation in peat DBD and carbon concentration;
- h. Investigating the relative importance of different litter inputs (including coarse woody debris
 and roots) to peat formation/C flux;
- 635 i. Improving our understanding of the spatial and, especially, temporal variation in greenhouse636 gas fluxes from peatlands;
- j. Investigating further the importance of vegetation (especially trees) as conduits forgreenhouse gases in tropical peatlands.

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1036 **Table**

Variable	Method(s) and key references
Location	± 10 m precision of good quality consumer-grade handheld
	GPS or GPS/GLONASS units is adequate.
Peat thickness	Measurement by coring or augering, taking care to define
	the base of the 'peat' using reproducible criteria, i.e. taking
	ample samples for carbon concentration and loss-on-

	ignition measurements: e.g. Parry et al. (2014).
Peat carbon concentration	Measured by elemental analysis, including samples from the full range of peat depths: Chambers et al. (2011); Chimner et al. (2014).
Peat dry bulk density	Each carbon concentration measurement should have an associated dry bulk density measurement: see Chambers et al. (2011).
Canopy height	Height of the ten tallest trees within 20 m of the core site, measured using a clinometer and tape measure, or a laser rangefinder: Phillips et al. (2009).
Vegetation composition/structure	Ideally, installation of a permanent 0.5–1 ha vegetation sampling plot following RAINFOR protocols (Malhi et al. 2002), extended where appropriate (e.g. to include small trees, shrubs and herbs where these are important, and coarse woody debris). Where this is impractical, a general description of the vegetation structure and dominant species within 20 m of the core site is sufficient for most remote sensing studies.

Table 1. A suggested protocol for site description which facilitates basic data comparison, and
development/testing of remote sensing techniques for peatland mapping and characterization.

1039 Figure captions

Figure 1. Distribution of the peat carbon pool in the tropics, based on country-scale estimates from
Page et al. (2011b; dotted regions indicate no data). In Australia, the estimate refers to the state of
Queensland only. Examples of lowland peatlands discussed in the text are indicated as follows: SSPS:

San San Pond Sak, Panama; PMF: Pastaza-Marañón Foreland, Peru; CC: Cuvette Centrale, Republic of
 Congo/Democratic Republic of Congo; TB: Tasek Bera, Malaysia; K: Kalimantan, Indonesia.

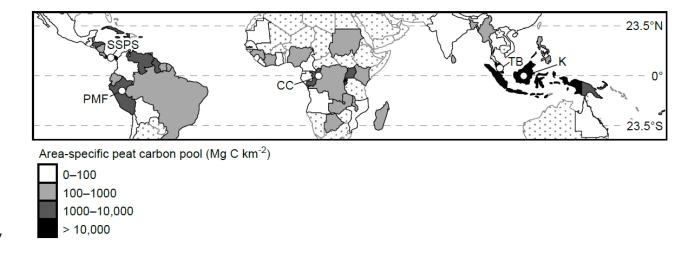
Figure 2. Vegetation classification of the Changuinola peat dome in the San San Pond Sak tropical
peatland, Panama, using Landsat Thematic Mapper imagery, supported by both aerial photographs
and field data as sources of reference.

Figure 3. Dry bulk density (DBD) values from published peat sequences. The box plots show the range of the data (dashed bars) and the lower, middle and upper quartiles (horizontal lines); the width of the bars is proportional to the square root of the size of each dataset (the total number of samples is 90); outliers are shown as circles. Only data from peats with <10% ash are shown. Note that the Sebangau peatland is in Indonesia; all other peatlands are from the Peruvian Amazon. Data sources: Wüst et al. 2002, 2003; Page et al. 2004; Lähteenoja et al. 2009a; Lähteenoja and Page, 2011.</p>

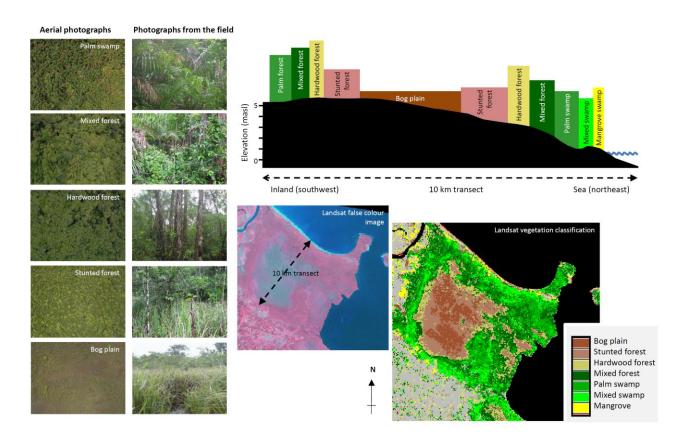
Figure 4. Dry bulk density and ash content from core SA6.5, Kalimantan, Indonesia (Page et al. 2004),
illustrating stratigraphic variation in DBD values.

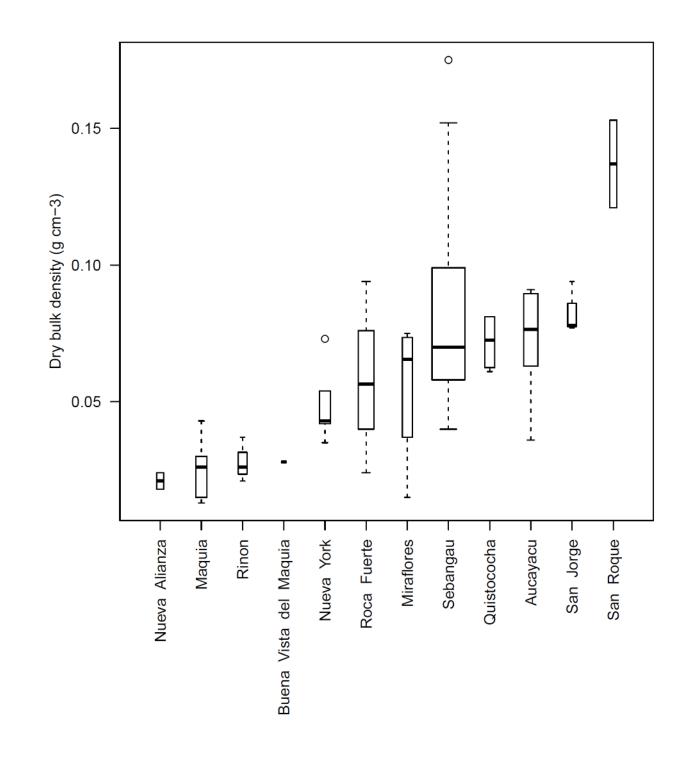
Figure 5. Organic carbon values from published peat sequences (references and symbols as for fig. 2).
Only data from peats with <10% ash are shown. Note that all records are from the Peruvian Amazon,
except Tasek Bera (Malaysia) and Sebangau (Indonesia). Data sources: Wüst et al. 2002, 2003; Page et
al. 2004; Lähteenoja et al. 2009; Lähteenoja and Page, 2011.

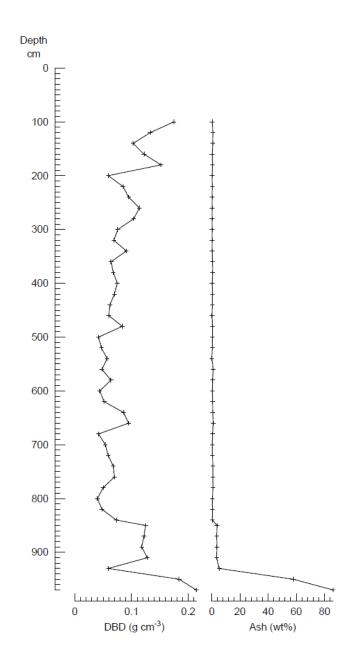
Figure 6. Carbon density measured using an elemental analyser plotted against ash content determined by loss-on-ignition (LOI) for some tropical peats. The straight line indicates the relationship used by Turunen et al. (2002) to estimate carbon content from LOI data. Only data from peats with <10% ash are shown. Data sources: Wüst et al. 2002, 2003; Page et al. 2004; Lähteenoja et al. 2009a; Lähteenoja and Page, 2011.













1078 Fig. 5

