

Informing forest carbon inventories under the Paris Agreement using ground-based forest monitoring data

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Key Words:	climate change, database, forest carbon, greenhouse gas inventory, International Panel on Climate Change (IPCC), natural climate solutions, nature based climate solutions

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Summary

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- Humans have been influencing Earth's climate via transformative impacts on forests for
 millenia, and forests are now recognized as critical to climate change mitigation under the
 Paris Agreement. The efficacy of climate change mitigation planning and reporting
 depends on quality data on forest carbon (C) stocks and changes. The Emission Factor
 Database (EFDB) of the International Panel on Climate Change (IPCC) is intended to be
 a definitive source for such data, but needs comprehensive and well-documented data to
 be so.
 - To facilitate submission of forest C estimates from scientific studies to EFDB, we develop and document a process for semi-automated data submission from the Global Forest C database (ForC v4.0), which is the largest compilation of ground-based forest C estimates. We then assess the data currently available through ForC and provide recommendations for improving forest data collection, analysis, and reporting.
 - As of March 2024, ForC contained ~19316 independent records relevant to EFDB, 1068 of which had been submitted and posted to EFDB. These represented 19% of the total EFDB records for forest land. Records were unevenly distributed across variables and geographic regions. 37% of ForC records reviewed could not be submitted because the original publication lacked required information.
 - In the future, ground-based forest C estimates should target gaps in the record, and studies should ensure that they report all information necessary for inclusion in EFDB. Given that climate change is rapidly impacting the world's forests, timely reporting of recent estimates will be critical to accurate forest C inventories.

Keywords:

- 47 climate change, database, forest carbon, greenhouse gas inventory, International Panel on
- 48 Climate Change (IPCC), natural climate solutions, nature based climate solutions

Societal Impact Statement

- 50 Human interactions with forests have shaped Earth's climate for millenia and will continue to do
- so as we target net-zero emission goals. Accurately characterizing these climate impacts requires

making accurate forest carbon data available for forest monitoring and planning. Here we develop a semi-automated process for submitting forest carbon measurements from the largest relevant scientific database to the International Panel on Climate Change's Emission Factor Database, which currently has sparse forest carbon data. Building this bridge from scientific research to international policy is an important step towards managing forests in a net-zero motivated future.



Introduction

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60 Humans have been influencing Earth's climate via ecologically transformative impacts on 61 ecosystems for >12,000 years (Bonan, 2016; Sanderman et al., 2017; Ellis et al., 2021). In recent 62 decades, as both anthropogenic land transformation and climate change have accelerated, this 63 relationship has come into increasing focus (IPCC, 2019a). Deforestation and forest degradation 64 are substantial sources of the greenhouse gas carbon dioxide (CO₂), currently accounting for 65 >10% of anthropogenic emissions (Friedlingstein et al., 2022). At the same time, CO₂ uptake by 66 remaining and regrowing forests, woodlands, and savannas has exceeded releases from 67 deforestation and other severe disturbances, resulting in a net carbon CO₂ sink of ~0.88 - 1.6 Gt 68 C yr⁻¹ (Xu et al., 2021; Harris et al., 2021), that offsets an estimated 10 - 18% of anthropogenic 69 CO₂ emissions from fossil fuels and cement (Xu et al., 2021; Harris et al., 2021). The future of 70 this important CO₂ sink depends primarily upon future trajectories of direct human impacts on 71 forests (i.e., deforestation/ degradation vs. conservation/ restoration, IPCC, 2019a, 2022a), and 72 also upon forest responses to climate change, which are likely to reduce the sink strength 73 (McDowell *et al.*, 2020). 74 Accordingly, forests play a substantial role in international plans for climate change mitigation 75 under the Paris Agreement (UNFCCC, 2015; Grassi et al., 2017). Forest conservation, 76 reforestation, and improved sustainable management all have significant – and relatively cost-77 effective – potential as climate change mitigation options (Roe et al., 2021). Conservation and 78 reforestation have the fourth and fifth largest net emission reduction potentials or all mitigation 79 options (IPCC, 2022b). However, envisioned forest-based climate change mitigation initiatives 80 do not always correspond to actual emission reductions implemented on the ground (e.g., 81 Badgley et al., 2022). Realistic planning and reporting is critically needed to ensure that forest-82 based climate change mitigation initiatives are effective, and this requires solid scientific data 83 and accounting frameworks (Deng et al., 2021; Anderson-Teixeira & Belair, 2022). 84 To this end, the International Panel on Climate Change (IPCC) provides guidance for national 85 greenhouse gas inventories for reporting to the United Nations Framework Convention on 86 Climate Change (UNFCCC) (IPCC, 2006a; IPCC, 2019b). Under this guidance, greenhouse gas 87 emissions to, or withdrawals from, the atmosphere are quantified on an annual basis for all 88 managed land, which includes most of the world's forest land (Ellis et al., 2010; Ogle et al.,

89 2018). The IPCC inventory guidelines include specific instructions for inventories for 90 greenhouse gas (mainly CO₂) exchanges between forest land and the atmosphere (Notes S1, 91 IPCC, 2006b, 2019b). A tiered approach is employed, where the lowest tier (Tier 1) represents 92 the simplest approach and relies on default parameter values – for example, forest carbon (C) 93 stocks values by ecozone and forest age class derived as the average of published estimates 94 (IPCC, 2019b; Rozendaal et al., 2022). Tier 1 values have improved over the years as more data 95 and methods have become available (Requena Suarez et al., 2019; Rozendaal et al., 2022), but 96 there remains room for improvement. For example, following the 2019 release of the latest IPCC 97 guidelines, it was revealed that IPCC's Tier 1 default failed to capture eight-fold variation of C 98 accumulation in regrowth forests within ecozones (Cook-Patton et al., 2020) and that C stocks in 99 mature African tropical montane forests were two-thirds higher than the IPCC Tier 1 values for 100 these forests (Cuni-Sanchez et al., 2021). High variability of forest C cycling within ecozones 101 (e.g., Cook-Patton et al., 2020; Cuni-Sanchez et al., 2021) means that it is crucial for 102 practitioners to have access to locally-specific information, when available. This rapid evolution 103 of scientific information on C cycling in forests is valuable for informing climate change 104 mitigation efforts but requires improved mechanisms for communicating the latest information 105 from scientific researchers to the practitioners who need reliable estimates for greenhouse gas 106 mitigation planning. 107 To improve data accessibility for preparing greenhouse gas estimates, the IPCC created the 108 Emission Factor Database (EFDB; https://www.ipcc-nggip.iges.or.jp/EFDB/main.php), which is 109 intended as a recognized library of emission factors and other parameters that can be used for 110 estimating greenhouse gas emissions and removals. The EFDB can be used both for efforts to 111 tally a nation's intended or accomplished greenhouse gas reductions, or as a basis of comparison 112 for external parties to evaluate these inventories. The EFDB encourages researchers to submit 113 estimates of emission factors or other related parameters (e.g., C stocks, net annual increments, 114 and annual fluxes for various pools, IPCC, 2006a; IPCC, 2019b) from peer-reviewed journal 115 articles or other accepted sources for inclusion in the database. Tens of thousands of relevant 116 forest carbon estimates have been published – and continue to be published at an accelerating 117 rate – but are not readily accessible to the practitioners assembling national greenhouse gas 118 inventories. To contribute to the goal of making forest C parameters available for accounting

119	under IPCC guidelines, forest scientists need an accessible summary of EFDB's requirements
120	and an efficient system for submission of data to the EFDB.
121	Our goal is to facilitate submission of forest C estimates from scientific studies to EFDB. We
122	document the process of submitting data to EFDB from the Global Forest Carbon Database,
123	ForC (https://forc-db.github.io/), which is the largest collection of published estimates of forest C
124	stocks, increments, and annual fluxes (Anderson-Teixeira et al., 2018, 2021; Anderson-Teixeira
125	et al., 2023), including data ingested from individual publications and relevant databases,
126	including the Global Reforestation Opportunity Assessment (GROA) database (Cook-Patton et
127	al., 2020, database doi: 10.5281/zenodo.3983644), and the Global Soil Respiration Database
128	(SRDB-V5, Bond-Lamberty & Thomson, 2010; Jian et al., 2021). We (1) map common
129	scientific forest C estimation methods and definitions to those used by the IPCC; (2) develop a
130	semi-automated process for preparing ForC data for submission to EFDB; and (3) assess the data
131	in ForC relevant to EFDB and records that have been submitted to date. We conclude with
132	recommendations as to how the scientific community can better provide useful data for forest C
133	inventories under the Paris Agreement.
	Materials and Methods
134	Materials and Methods
135	Major steps for submission of data from ForC to EFDB included (1) mapping ForC into EFDB,
136	including aligning ForC terms and concepts with those defined by IPCC guidelines (summarized
137	in Notes S1), (2) revising ForC v3.0 to support semi-automated submissions to EFDB, yielding
138	ForC v4.0 (detailed in Methods S1), and (3) submitting data to EFDB.
139	1. Mapping ForC to EFDB
140	With input from the IPCC's Technical Support Unit and referencing IPCC guidance (IPCC,
141	2003, 2019b; IPCC, 2006a), we determined how EFDB fields should be populated using ForC
142	fields (summarized in Table S2). Fields in EFDB included several fields describing how the
143	record fits within IPCC's framework (source/sink category, greenhouse gas type, C pool,
144	relevant equations), several describing the C estimate itself (variable, value, units, 95%
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145	confidence limits), a few composite fields describing biotic and abiotic conditions (e.g.,

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disturbances, stand age, plot information), and a few describing the source (e.g., type, citation, data provider). Most relevant ForC fields mapped directly into an EFDB field, either as the only contents of that field or as part of a composite record. For some fields, simple conditional logic was used to populate EFDB fields based on ForC records. For example, in cases where original studies did not present 95% confidence intervals (required by IPCC when available) but did present standard error or n and standard deviation, we calculated the 95% confidence intervals. There were two cases in which more complex mapping was required: (1) mapping of C cycle variables and (2) land classification. Carbon cycle variables We identified ForC variables that were relevant to the IPCC methodology and EFDB (Notes S1). These included organic matter or C stocks, net annual increments, influxes (a.k.a. "gross annual increments" by IPCC), and outfluxes for each IPCC-defied C pool (Table 1, Fig. 1). It is important to note that the correspondence of ForC variables to IPCC criteria often depends upon measurement protocols ("important sources of estimate variation" in Table 1). For example, For Crecords of biomass and dead wood vary in the minimum stem diameter censused, such that some records would match the IPCC criteria whereas others would not. Information on minimum diameters censused and other important sources of methodological variation were mapped into EFDB (Table S2). Details on the mapping of ForC variables to EFDB are documented in the GitHub repository associated with this publication (https://github.com/forc-db/IPCC-EFDBintegration).

Table 1. IPCC-defined forest carbon pools with definitions and measurement methods.

pool	definition	important sources of estimate variation	IPCC guidance
aboveground biomass	all biomass of living vegetation	minimum size censused	may exclude understory if minor component
		include non-dicot trees?	yes
		include dead standing?	no
		biomass allometry	Tier 1 defaults draw on a variety of allometric models
belowground biomass	all biomass of live roots	all factors relevant to aboveground biomass	see above
		allometry or assumed ratio of below-to above-ground biomass (R)	can estimate based on R
		minimum root diameter	may exclude fine roots; suggested diameter cutoff of 2 mm for fine roots
dead wood	all non-living woody biomass above a specified diameter, aboveground or belowground	minimum diameter	10 cm default, but may be chosen by country
		include belowground?	yes
litter	all non-living biomass smaller than dead wood but larger than soil organic matter, in various states of decomposition both above or within the mineral or organic soil	maximum diameter (= minimum diameter for deadwood)	10 cm default, but may be chosen by country
		minimum size (= size limit for soil organic matter)	suggested 2 mm
		layers included	entire O horizon: litter (OL), fumic (OF), and humic (OH) layers
		include belowground?	yes
soil organic matter	organic carbon in mineral soils to a specified depth	sampling depth	30 cm default, but may be chosen by country

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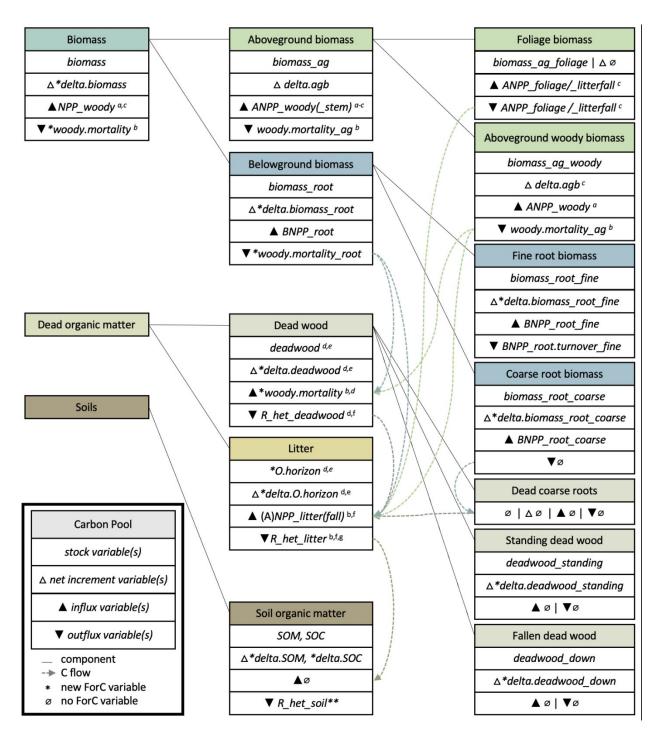


Figure 1. Schematic illustrating the carbon pools defined under IPCC Guidelines for national greenhouse gas inventories; corresponding ForC variables, and relationships among them.

- 172 For each C pool, we show ForC variables corresponding to the stock, net annual increment,
- influx, and outflux. Most, but not all, EFDB-relevant ForC variables are shown here.
- 174 Correspondence of ForC variables to IPCC criteria often depends upon measurement protocols

175	(e.g., minimum stem diameter censused). Additional caveats are as follows: (a,b) branch fall and
176	mortality of stems below the minimum stem diameter censused, which are necessary for a full
177	accounting of dead organic matter production but typically assumed negligible for calculations
178	of biomass change, are excluded by common measurement practice (a) or ForC variable
179	definition (b); (c) assumes that leaf production equals leaf fall, or that changes in foliage
180	biomass are negligble; (d,e) belowground components excluded by common measurement
181	practice (d) or ForC variable definition (e); (f) excludes movement of dead wood into litter
182	through breakage or size reduction; (g) measurements often limited to litter horizon (OL) and
183	may exclude larger branches and stems classified as litter and/or the more decomposed layers of
184	the O horizon. **This variable is techically EFDB-relevant but not selected for submission
185	because their is no corresponding influx variable.
186	Land classification
187	Determination of the IPCC land-use category (i.e., Forest Land, Grassland, Wetlands, Cropland,
188	Settlements, or Other Land; Notes S1) was made based on the dominant life form recorded in
189	ForC. Woody vegetation – including early seral vegetation – was classified as forest, consistent
190	with the IPCC definition that Forest Land includes land expected to succeed to forest. Mixtures
191	of woody vegetation and grasses (i.e., anything from a shrub-encroached grassland to a tree-
192	dominated savanna) were given dual classification of Forest Land and Grassland, indicating that
193	records may be relevant to either category depending on the definition of forest applied (varies
194	by country). Grass- or crop-dominated ecosystems (included in ForC as controls for studies of
195	forest regrowth following agricultural abandonment) were classified as Grassland and Cropland,
196	respectively.
197	Classification into EFDB sub-categories was dependent upon stand age and site history. For
198	Forest Land \geq 20 years old or of unknown (relatively mature) age, or Forest Land \leq 20 years old
199	that was forest prior to a stand-clearing disturbance, the past land-use category was Forest Land,
200	making the sub-category "Forest Land Remaining Forest land". For forests <20 years old with
201	history including cultivation/ tillage or grazing, past land-use categories were Cropland and
202	Grassland, respectively, making land-use subcategories were "Cropland converted to Forest
203	Land" and "Grassland converted to Forest Land", respectively. For forests <20 years old with
204	unspecified previous agricultural use, we assigned the sub-category "Land Converted to Forest

205 land". Forests <20 years old with unknown land use prior to the study date were simply 206 classified as "Forest Land". The same logic was applied for savannas, but including both Forest 207 Land and Grassland as potentially relevant categories. 208 Given the lack of public information needed to determine whether lands are classified as 209 managed (Ogle et al., 2018; Deng et al., 2021), and because the IPCC's definition of managed 210 land is more expansive than is commonly applied in the scientific literature and hence in ForC, 211 we did not include any classification of land management status from ForC in the records 212 submitted to EFDB. However, we did provide auxiliary information that should be useful in 213 making this determination, including geographical location and notable disturbance events. 214 2. Updating ForC 215 Previous versions of ForC (Anderson-Teixeira et al., 2016, 2018, 2021) contained most of the information required by EFDB, and, more broadly, to inform C stock change calculations under 216 217 the Paris Agreement. However, modest changes to the structure and contents of ForC were 218 needed in order to provide all information required by EFDB and to improve ForC's capacity to 219 serve as a repository of valuable information for forest C inventories under IPCC guidelines. To 220 support export of data to EFDB, and to improve the overall quality of the ForC database, we 221 added or modified 18 fields (Table S1), defined 15 new variables, implemented enhanced quality 222 control, manually reviewed >1963 records to obtain additional required information, and added 223 329 new records. Having implemented these changes, which are described in detail in Methods 224 S1, to ForC v3.0 (Anderson-Teixeira et al., 2021), we released a new major version: ForC v4.0. 225 3. Submission of ForC data to EFDB 226 To submit complete, reviewed ForC records into EFDB, we created R scripts to restructure ForC 227 records and populate EFDB's bulk import form. Criteria for data submission were that (1) 228 records had been checked against the original study and determined to be complete and correct, 229 and as originally presented, (2) all relevant records within the publication were included in ForC, 230 (3) the original study presented values in tables or text, as opposed to the values having been 231 digitized from graphs or calculated based on related variables, and (4) the records had not 232 previously been submitted to EFDB. Once converted into EFDB format, the records were 233 reviewed and then sent to the IPCC's Technical Support Unit for submission to EFDB.

235	As of March 01, 2024, ForC (v4.0) contained 32686 independent records (39848 total), 19316 of
236	which were for the 42 variables relevant to EFDB (Fig. 1). We had reviewed or added 2292
237	records for submission to EFDB, 1438 of which met the criteria for submission, and 1068 of
238	which had been reviewed by EFDB's review panel, accepted, and posted (Table S3). The 37%
239	attenuation between records we reviewed and those submitted to EFDB was due to the presence
240	of digitized records and records where a variable's value had been calculated as the sum or
241	difference of related variables rather than presented directly in the text. The discrepancy between
242	the number of records sent and that posted to EFDB is because some records (26%) were deemed
243	not applicable to EFDB by the review panel. As of March 01, 2024, the 1438 ForC records
244	posted in EFDB represented 19% of the total EFDB records for forest land.
245	ForC v4.0 contained records for 29 of the 42 variables (or closely-related variable groups)
246	relevant to EFDB (Fig. 1, Table S3). Records were submitted for 19 variables (or closely-related
247	variable groups), including variables from each C pool (Table S3). The records available in ForC
248	v4.0 – and the subset submitted to EFDB – were very unevenly distributed across variables
249	(Table S3). The majority (82%) of records submitted were for C stocks, including 3% for total
250	biomass, 53% for aboveground biomass, 4% for components of aboveground biomass (wood or
251	foliage), 4% for root biomass, 2% for components of root biomass (coarse or fine roots), 5% for
252	dead wood, 3% for components of dead wood (standing or fallen), 4% for litter (entire O horizon
253	or OL layer component), and 4% for SOM/ SOC. Increment records totaled 9% of records
254	submitted, all but four of which described aboveground biomass. The remaining 9% of records
255	submitted described fluxes, all of which were either inputs or outputs to the aboveground
256	biomass pool, a subset of which also described inputs to the dead wood or litter pool (Fig.
257	1,Table S3).
258	ForC v4.0 records and the subset submitted to EFDB were distributed across all forested
259	continents, biomes, and forest types, albeit very unevenly (Figs. 2-3). Among the records
260	submitted to EFDB, the largest number came from North America, followed by Asia, South
261	America, and Africa (Fig. 3c), with the most represented FAO ecozones being boreal coniferous
262	forest, temperate continental forest, and temperate mountain systems, followed by tropical rain
263	forests and moist deciduous forests (Fig. 3b). By far the most records came from needleleaf

Results

evergreen forests, followed by broadleaf deciduous and broadleaf evergreen (Fig. 3b). The largest number of records came from mature forests (>100 years), followed by young and intermediate-aged stands (Fig. 3d).

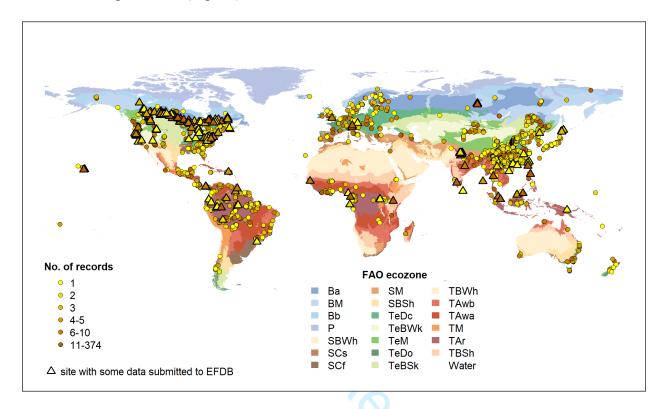


Figure 2. Map of sites in ForC shaded by number of independent records relevant to (circles) and submitted to (triangles) EFDB. Symbols are colored according to the number of records at each site. Underlying map shows FAO ecozones, which are coded as follows: Ba-Boreal coniferous forest, Bb-Boreal tundra woodland, BM-Boreal mountain systems, P-Polar, SBSh-Subtropical steppe, SBWh-Subtropical desert, SCf-Subtropical humid forest, SCs-Subtropical dry forest, SM-Subtropical mountain systems, TAr-Tropical rain forest, TAwa-Tropical moist deciduous forest, TAwb-Tropical dry forest, TBSh-Tropical shrubland, TBWh-Tropical desert, TeBSk-Temperate steppe, TeBWk-Temperate desert, TeDc-Temperate continental forest, TeDo-Temperate oceanic forest, TeM-Temperate mountain systems, TM-Tropical mountain systems.

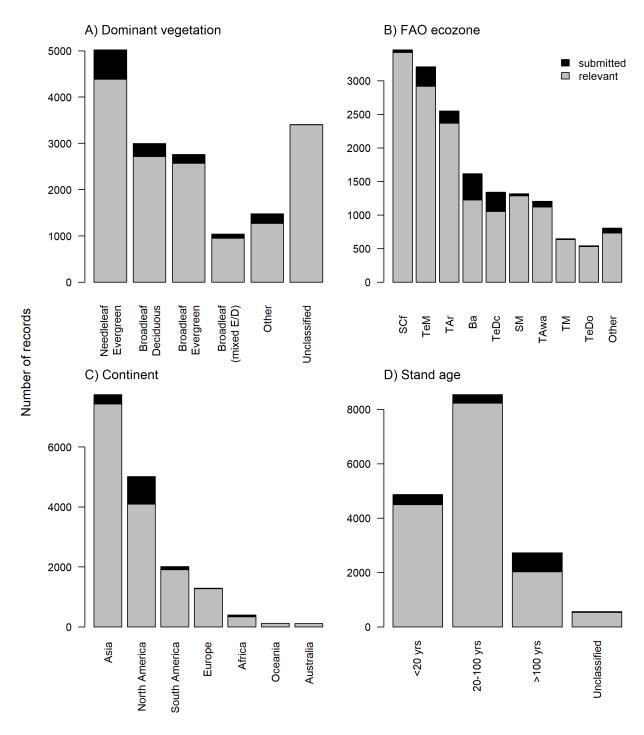


Figure 3. Histograms of number of independent records in ForC relevant to (grey) and submitted to (black) EFDB, organized by (a) dominant vegetation type, (b) FAO ecozone, (c) continent, and (d) stand age. For dominant vegetation (a), 'Other' includes deciduous needleleaf, mixed broadleaf- needleleaf, non-woody vegetation (e.g., early successional), and

282 incompletely classified or mixed forest types. For FAO ecozones (b), codes are as listed in the 283 caption of Figure 2. 284 **Discussion** 285 Here we have developed a framework for submitting records from ForC v4.0, the largest 286 compilation of ground-based forest C estimates, to the EFDB, thus making those data more 287 accessible for reporting CO₂ emissions and removals from forest land consistent with good 288 practice in the IPCC guidelines (IPCC, 2006a; IPCC, 2019b). As of March 01, 2024, 1068 ForC 289 records have been posted to EFDB, which represents just 6% of potentially relevant records in 290 For Cv4.0, but currently comprises 19% of the total EFDB records for forest land. The records 291 present in ForC and submitted to EFDB are very unevenly distributed across variables, regions, 292 and forest types (Figs. 2-3, Table S3), reflecting broader patterns in allocation of research effort 293 and pointing to scientific research and reporting needs. 294 Based on our experience contributing forest C data to EFDB via ForC, we make several 295 recommendations as to how scientists can improve forest C records in EFDB through database 296 work, new data collection and analysis, and reporting. 297 Database needs 298 There is vast potential to expand forest C data in EFDB by completing the process of reviewing 299 and submitting data that are already in ForC (Figs. 2-3). So far, only ~7% of the EFDB-relevant 300 data in ForC have been submitted to EFDB. Although this process requires manual review of 301 records, the submission of new records to EFDB is greatly facilitated by the fact that most 302 pertinent information for each record is already entered in ForC and can be easily prepared for 303 submission to EFDB using the system developed here. Future efforts to review studies for 304 submission should optimize for representation across geographic regions, forest types, and 305 variables, giving priority to those from currently under-represented regions and forest types 306 (Figs. 2-3, Table S3). Other categories of records to prioritize include those from countries 307 relying on existing data for their greenhouse gas inventories (Tier 1 or 2 methodology), the 308 variables most needed by EFDB users, and the most up-to-date records.

309	In addition to the large potential to expand EFDB using records already in ForC, there are
310	extensive EFDB-relevant forest C data that are not currently included in ForC, with more being
311	published on a nearly daily basis. Coverage of particular variables or regions could be vastly
312	improved through systematic review of the literature, although this requires focused and
313	extensive manual effort. Recent efforts have compiled large databases of relevant data from
314	monoculture plantation forests (Bukoski et al., 2022) and mixed species plantation forests
315	(Warner et al., 2022; Feng et al., 2022), and such a compilation is in works for agroforestry
316	(Susan Cook-Patton, unpublished data). Beyond expanding collections of relevant forest C
317	records, such reviews are valuable for assessing the availability of published records and
318	identifying variables and regions that require additional data collection and analysis.
319	Data collection and analysis needs
320	New data collection and analyses are needed to fill notable knowledge gaps. While aboveground
321	biomass stocks in particular have received – and continue to receive – by far the most research
322	attention (Table S3, NISAR, 2018; Quegan et al., 2019; Dubayah et al., 2020; Anderson-
323	Teixeira et al., 2021), production of an accurate global map of forest C stocks remains an
324	ongoing challenge (Araza et al., 2023). Other pools and variables remain poorly quantified and
325	highly uncertain for many parts of the world (Table S3, Tifafi et al., 2018; Anderson-Teixeira et
326	al., 2021), introducing substantive uncertainties into global forest C budgets (Pan et al., 2011;
327	Harris et al., 2021). Furthermore, data distribution is very uneven across forest types and
328	geographical regions (Figs. 2-3). For instance, data on C cycling of tropical forests – particularly
329	in Africa - remains relatively sparse, in large part due to substantial barriers to data collection
330	and distribution (de Lima et al., 2022). More generally, belowground C measurements remain
331	sparse globally. Significant investment in research and researchers focused on ground-based
332	measurement of forest C in such regions will be important to filling knowledge gaps in forest C
333	cycling (de Lima et al., 2022; Labrière et al., 2023; Araza et al., 2023).
334	Several EFDB-relevant variables have not been calculated and presented as frequently as would
335	be possible given existing forest census data and minimal extra research effort. For example,
336	aboveground woody mortality and aboveground biomass increment can be calculated from the
337	same census data as aboveground woody productivity, yet the latter has received far more
338	research attention (Table S3, Anderson-Teixeira et al., 2021). Similarly, live coarse root

339 biomass, total biomass, and changes therein could in theory be estimated in parallel with 340 aboveground biomass, with the greatest barrier being that allometric models for estimating root 341 biomass are not as reliable or easily available as are those for aboveground biomass (Chave et 342 al., 2014; Réjou-Méchain et al., 2017; Gonzalez-Akre et al., 2022). However, while equations 343 for estimating root (and thereby total) biomass require improvement, they do exist for many 344 forest types (Mokany et al., 2006; e.g., Brassard et al., 2011; Chojnacky et al., 2014; Waring & 345 Powers, 2017). We recommend that, when possible, researchers calculate and report all EFDB-346 relevant variables. 347 Data reporting needs 348 We recommend that, in order to make research most useful for estimating C stock changes 349 following IPCC guidelines, researchers calculate and report results according to IPCC good 350 practice (Table 2). Importantly, simple decisions on the presentation of results will determine 351 whether the records meet the current criteria for inclusion in EFDB. For example: (1) presenting 352 data only in a figure makes them ineligible for inclusion in EFDB, whereas presentation in a 353 table or supplementary data file allows inclusion while supporting FAIR goals (Stall et al., 354 2019); (2) direct presentation of all relevant variables allows inclusion, whereas presenting only 355 components of variables of interest (e.g., parsing litter into fine woody debris, OL, OF, and OH 356 layers) or requiring simple mathematical operations to obtain a variable of interest (e.g., delta.agb = ANPP woody - woody.mortality.agb) disqualifies records from inclusion; (3) 357 358 matching IPCC-defined thresholds for defining C pools (Table 1), which may vary by country, 359 can make the data far more relevant estimating forest C stock changes according to IPCC 360 guidelines (e.g., using a 10 cm cutoff between dead wood and litter, presenting soil C to a depth 361 of 30 cm). It should also be emphasized that reporting of 95% confidence intervals (or other 362 metrics of error), when applicable, is highly desirable and makes the data more relevant to IPCC.

Table 2. Recommended best practices for reporting forest C estimates of value to national greenhouse gas inventories under IPCC guidance.

criteria recommendation		rationale
variables to include	When possible, calculate and present all relevant variables that can be readily estimated based on available data.	Estimates of relevant variables are not always calcualted.
forest census methods	Adopt IPCC guidelines (country-specific) for minimum stem size in censues in census and reporting. Ideally, census stem down to the smallest diameter feasible.	IPCC biomass pool definition includes all living vegetation, but understory may be excluded when contribution is minor.
	Census all taxa crontributing signficantly to biomass	IPCC biomass pool definition includes all living vegetation.
dead organic matter sampling	Adopt IPCC recommendations for minimum diameter of deadwood (country-specific, default 10 cm).	Diameter cutoff must be applied consistently by each country.
belowground sampling	Select and report soil sampling increments to include a cutoff at 30 cm depth (or country-specific depth).	Diameter cutoff must be applied consistently by each country.
reporting variables	Present each EFDB- relevant variable individually, as opposed to requiring summation of related variables.	EFDB requires that values in the database be presented in the original article, and cannot accept subsequent calculations.
reporting estimates	Report all relevant values in tables, text, or supplementary tables/ data files, as opposed to in figures only.	EFDB does not accept values digitized from figures.
reporting confidence intervals	Report 95% confidence intervals, standard error, or standard deviation and sample size.	EFDB requires confidence invervals whenever possible.

For those compiling published records (e.g., for meta-analyses), the data set can have added value if all information required by EFDB is extracted from original publications. This includes – but is not limited to – retaining original values as presented without modification or rounding, noting whether data were digitized, recording confidence intervals, and recording all required fields (as indicated in the EFDB's bulk import template). The significant effort required to map a database into EFDB has been accomplished here (Table S2), and we hope that it will prove useful as a model for other efforts.

Conclusions

As human society strives to achieve net-zero greenhouse gas emissions, forest researchers can make their research more useful for forest C inventories under IPCC guidelines by calculating and reporting results in ways that are consistent with methodologies provided in the IPCC guidelines (Tables 1, 2). In addition, substantial investments in research and researchers focused

378 on ground-based measurement of forest C will be required to fill knowledge gaps and thereby 379 increase the accuracy of forest CO₂ inventories for forest lands under the Paris Agreement. This 380 challenge is heightened by the fact that forests are changing rapidly (e.g., McDowell et al., 381 2020), and data collected a decade or more in the past may be increasingly less accurate. This 382 heightens the need for an efficient system of making forest C data accessible for national 383 greenhouse gas inventories. We view the system developed here for submitting ForC data to the 384 IPCC EFDB as one important step towards that goal. 385 Acknowledgements 386 We gratefully acknowledge the substantial contributions of Valentyna Slivinska and Sandro 387 Federici for collaboration on the conception, design, and technical review of this project. Thank 388 you to all researchers who collected and published the data contained in ForC, and to all research 389 assistants and collaborators who have contributed to building the database. Thank you to Avni 390 Malhotra for helpful comments on an earlier draft of this manuscript. Funding for this study was 391 provided by the Smithsonian (Forest Global Earth Observatory, Smithsonian Working Land and 392 Seascapes); a Bezos Earth Fund grant to the Nature Conservancy, with a sub-grant to NZCBI; 393 and the Institute for Global Environmental Strategies. 394 **Author Contribution** KAT and VH conceived and designed the project; VH wrote the scripts for database 395 396 management, data submission to EFDB, and the analyses presented here; MW, TR, and RBM 397 added and reviewed ForC data, BBL and SCP contributed large databases to ForC (EFDB and 398 GROA, respectively); CP provided methodological expertise; KAT, VH, and MW prepared the 399 first draft of the manuscript; all authors reviewed the results and approved the final version of the 400 manuscript. 401 **Data Availability Statement** 402 All code and data are openly available. The ForC database and associated code are available via 403 the ForC repository within the ForC-db organization on GitHub (https://github.com/forc-

- db/ForC), and the version used here (ForC v4.0) is archived in Zenodo (Anderson-Teixeira et al.,
- 405 2023, DOI: 10.5281/zenodo.8020861). The data and code associated with data submission to
- 406 EFDB and preparation of this manuscript are available via the the IPCC-EFDB-integration
- repository within the ForC-db organization on GitHub (https://github.com/forc-db/IPCC-EFDB-
- 408 integration) and archived in Zenodo (DOI: 10.5281/zenodo.8021474).

409 Supplementary Information

- Table S1. Updates to ForC field implemented between releases of v3.0 and v4.0
- 411 Table S2. Mapping of ForC fields to EFDB
- Table S3. Numbers of ForC records and EFDB submissions by variable
- Notes S1. Primer on forest land classification and carbon pools under IPCC guidelines
- 414 Methods S1. Updates to ForC (ForC v4.0)
- 415 References
- 416 Anderson-Teixeira KJ, Belair EP. 2022. Effective forest-based climate change mitigation
- 417 requires our best science. Global Change Biology 28: 1200–1203.
- 418 Anderson-Teixeira K, Herrmann V, Morgan BB, Actions-User, Williams M, Rogers T,
- 419 McGregor I, Hua MWM, Ferson A, Bond-Lamberty B, et al. 2023. Forc-db/ForC: First
- 420 version with EFDB integration.
- 421 Anderson-Teixeira KJ, Herrmann V, Morgan RB, Bond-Lamberty B, Cook-Patton SC,
- 422 Ferson AE, Muller-Landau HC, Wang MMH. 2021. Carbon cycling in mature and regrowth
- forests globally. *Environmental Research Letters* **16**: 053009.
- 424 Anderson-Teixeira KJ, Wang MMH, McGarvey JC, Herrmann V, Tepley AJ, Bond-
- 425 Lamberty B, LeBauer DS. 2018. ForC: A global database of forest carbon stocks and fluxes.
- 426 *Ecology* **99**: 1507–1507.

- 427 Anderson-Teixeira KJ, Wang MMH, McGarvey JC, LeBauer DS. 2016. Carbon dynamics of
- 428 mature and regrowth tropical forests derived from a pantropical database (TropForC-db). *Global*
- 429 *Change Biology* **22**: 1690–1709.
- 430 Araza A, Herold M, de Bruin S, Ciais P, Gibbs DA, Harris N, Santoro M, Wigneron J-P,
- 431 Yang H, Málaga N, et al. 2023. Past decade above-ground biomass change comparisons from
- four multi-temporal global maps. International Journal of Applied Earth Observation and
- 433 *Geoinformation* **118**: 103274.
- 434 Badgley G, Freeman J, Hamman JJ, Haya B, Trugman AT, Anderegg WRL, Cullenward
- 435 **D. 2022.** Systematic over-crediting in California's forest carbon offsets program. *Global Change*
- 436 *Biology* **28**: 1433–1445.
- 437 **Bonan GB. 2016.** Forests, Climate, and Public Policy: A 500-Year Interdisciplinary Odyssey.
- 438 Annual Review of Ecology, Evolution, and Systematics 47: 97–121.
- 439 **Bond-Lamberty B, Thomson A. 2010.** A global database of soil respiration data.
- 440 *Biogeosciences* 7: 1915–1926.
- Brassard BW, Chen HYH, Bergeron Y, Paré D. 2011. Coarse root biomass allometric
- equations for Abies balsamea, Picea mariana, Pinus banksiana, and Populus tremuloides in the
- boreal forest of Ontario, Canada. *Biomass and Bioenergy* **35**: 4189–4196.
- Bukoski JJ, Cook-Patton SC, Melikov C, Ban H, Chen JL, Goldman ED, Harris NL, Potts
- MD. 2022. Rates and drivers of aboveground carbon accumulation in global monoculture
- plantation forests. *Nature Communications* **13**: 1–13.
- Chave J, Réjou-Méchain M, Búrquez A, Chidumayo E, Colgan MS, Delitti WBC, Duque
- 448 A, Eid T, Fearnside PM, Goodman RC, et al. 2014. Improved allometric models to estimate
- the aboveground biomass of tropical trees. *Global Change Biology* **20**: 3177–3190.
- 450 Chojnacky DC, Heath LS, Jenkins JC. 2014. Updated generalized biomass equations for
- North American tree species. *Forestry* **87**: 129–151.

- 452 Cook-Patton SC, Leavitt SM, Gibbs D, Harris NL, Lister K, Anderson-Teixeira KJ, Briggs
- 453 RD, Chazdon RL, Crowther TW, Ellis PW, et al. 2020. Mapping carbon accumulation
- potential from global natural forest regrowth. *Nature* **585**: 545–550.
- Cuni-Sanchez A, Sullivan MJP, Platts PJ, Lewis SL, Marchant R, Imani G, Hubau W,
- 456 Abiem I, Adhikari H, Albrecht T, et al. 2021. High aboveground carbon stock of African
- 457 tropical montane forests. *Nature* **596**: 536–542.
- de Lima RAF, Phillips OL, Duque A, Tello JS, Davies SJ, de Oliveira AA, Muller S,
- 459 Honorio Coronado EN, Vilanova E, Cuni-Sanchez A, et al. 2022. Making forest data fair and
- 460 open. Nature Ecology & Evolution.
- Deng Z, Ciais P, Tzompa-Sosa ZA, Saunois M, Qiu C, Tan C, Sun T, Ke P, Cui Y, Tanaka
- 462 **K**, et al. 2021. Comparing national greenhouse gas budgets reported in UNFCCC inventories
- against atmospheric inversions. Earth System Science Data Discussions: 1–59.
- Dubayah R, Blair JB, Goetz S, Fatoyinbo L, Hansen M, Healey S, Hofton M, Hurtt G,
- 465 Kellner J, Luthcke S, et al. 2020. The Global Ecosystem Dynamics Investigation: High-
- resolution laser ranging of the Earth's forests and topography. Science of Remote Sensing 1:
- 467 100002.
- 468 Ellis EC, Gauthier N, Klein Goldewijk K, Bliege Bird R, Boivin N, Díaz S, Fuller DQ, Gill
- JL, Kaplan JO, Kingston N, et al. 2021. People have shaped most of terrestrial nature for at
- least 12,000 years. Proceedings of the National Academy of Sciences 118: e2023483118.
- 471 Ellis EC, Klein Goldewijk K, Siebert S, Lightman D, Ramankutty N. 2010. Anthropogenic
- 472 transformation of the biomes, 1700 to 2000. Global Ecology and Biogeography 19: 589–606.
- Feng Y, Schmid B, Loreau M, Forrester DI, Fei S, Zhu J, Tang Z, Zhu J, Hong P, Ji C, et
- 474 al. 2022. Multispecies forest plantations outyield monocultures across a broad range of
- 475 conditions. *Science* **376**: 865–868.
- 476 Friedlingstein P, O'Sullivan M, Jones MW, Andrew RM, Gregor L, Hauck J, Le Quéré C,
- 477 Luijkx IT, Olsen A, Peters GP, et al. 2022. Global Carbon Budget 2022. Earth System Science
- 478 Data 14: 4811–4900.

- 479 Gonzalez-Akre E, Piponiot C, Lepore M, Herrmann V, Lutz JA, Baltzer JL, Dick CW,
- 480 Gilbert GS, He F, Heym M, et al. 2022. Allodb: An R package for biomass estimation at
- 481 globally distributed extratropical forest plots. *Methods in Ecology and Evolution* **13**: 330–338.
- 482 Grassi G, House J, Dentener F, Federici S, den Elzen M, Penman J. 2017. The key role of
- 483 forests in meeting climate targets requires science for credible mitigation. *Nature Climate*
- 484 *Change* 7: 220–226.
- Harris NL, Gibbs DA, Baccini A, Birdsey RA, Bruin S de, Farina M, Fatoyinbo L, Hansen
- 486 MC, Herold M, Houghton RA, et al. 2021. Global maps of twenty-first century forest carbon
- 487 fluxes. *Nature Climate Change*: 1–7.
- 488 **IPCC**. **2003**. *Good Practice Guidance for Land Use, Land-Use Change and Forestry* (J Penman,
- 489 M Gytarsky, T Hiraishi, T Krug, D Kruger, R Pipatti, L Buendia, K Miwa, T Ngara, K Tanabe,
- 490 et al., Eds.). Hayama, Japan: Institute for Global Environmental Strategies.
- 491 **IPCC. 2006a.** 2006 IPCC Guidelines for National Greenhouse Gas Inventories, Prepared by the
- National Greenhouse Gas Inventories Programme, Eggleston H.S., Buendia L., Miwa K., Ngara
- 493 T. And Tanabe K. (eds). Japan: IGES.
- 494 **IPCC. 2006b.** Agriculture, Forestry, and Other Land Use. In: Eggleston S, Buendia L, Miwa K,
- Ngara T, Tanabe K, eds. 2006 IPCC Guidelines for National Greenhouse Gas Inventories.
- 496 Hayama, Japan: Institute for Global Environmental Strategies.
- 497 **IPCC. 2019b.** 2019 Refinement to the 2006 IPCC Guidelines for National Greenhouse Gas
- 498 Inventories. In: Calvo Buendia E, Tanabe K, Baasansuren J, Fukuda M, Ngarize S, Osako A,
- 499 Pyrozhenko Y, Shermanau P, Federici S, eds. Switzerland: IPCC.
- 500 **IPCC**. **2019a**. Climate Change and Land: An IPCC special report on climate change,
- desertification, land degradation, sustainable land management, food security, and greenhouse
- 502 gas fluxes in terrestrial ecosystems [P.R. Shukla, J. Skea, E. Calvo Buendia, V. Masson-
- 503 Delmotte, H.- O. Pörtner, D. C. Roberts, P. Zhai, R. Slade, S. Connors, R. Van Diemen, M.
- 504 Ferrat, E. Haughey, S. Luz, S. Neogi, M. Pathak, J. Petzold, J. Portugal Pereira, P. Vyas, E.
- 505 Huntley, K. Kissick, M. Belkacemi, J. Malley, (eds.)].

- 506 **IPCC. 2022a**. Climate Change 2022: Mitigation of Climate Change. Contribution of Working
- Group III to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change
- [P.R. Shukla, J. Skea, R. Slade, A. Al Khourdajie, R. Van Diemen, D. McCollum, M. Pathak, S.
- 509 Some, P. Vyas, R. Fradera, M. Belkacemi, A. Hasija, G. Lisboa, S. Luz, J. Malley, (eds.)].
- 510 Cambridge University Press, Cambridge, UK and New York, NY, USA.
- 511 **IPCC. 2022b.** Summary for Policymakers. Policymakers [P.R. Shukla, J. Skea, A. Reisinger, R.
- 512 Slade, R. Fradera, M. Pathak, A. Al Khourdajie, M. Belkacemi, R. Van Diemen, A. Hasija, G.
- Lisboa, S. Luz, J. Malley, D. McCollum, S. Some, P. Vyas, (eds.)]. In: Climate Change 2022:
- 514 Mitigation of Climate Change. Contribution of Working Group III to the Sixth Assessment
- Report of the Intergovernmental Panel on Climate Change [P.R. Shukla, J. Skea, R. Slade, A. Al
- Khourdajie, R. Van Diemen, D. McCollum, M. Pathak, S. Some, P. Vyas, R. Fradera, M.
- 517 Belkacemi, A. Hasija, G. Lisboa, S. Luz, J. Malley, (eds.)]. Cambridge University Press,
- 518 Cambridge, UK and New York, NY, USA. Doi: 10.1017/9781009157926.001.
- Jian J, Vargas R, Anderson-Teixeira K, Stell E, Herrmann V, Horn M, Kholod N, Manzon
- J, Marchesi R, Paredes D, et al. 2021. A restructured and updated global soil respiration
- database (SRDB-V5). Earth System Science Data 13: 255–267.
- Labrière N, Davies SJ, Disney MI, Duncanson LI, Herold M, Lewis SL, Phillips OL,
- Quegan S, Saatchi SS, Schepaschenko DG, et al. 2023. Toward a forest biomass reference
- measurement system for remote sensing applications. Global Change Biology n/a.
- 525 McDowell NG, Allen CD, Anderson-Teixeira K, Aukema BH, Bond-Lamberty B, Chini L,
- 526 Clark JS, Dietze M, Grossiord C, Hanbury-Brown A, et al. 2020. Pervasive shifts in forest
- 527 dynamics in a changing world. *Science* **368**.
- Mokany K, Raison RJ, Prokushkin AS. 2006. Critical analysis of root: Shoot ratios in
- 529 terrestrial biomes. Global Change Biology 12: 84–96.
- 530 NISAR. 2018. NASA-ISRO SAR (NISAR) Mission Science Users' Handbook. NASA Jet
- 531 Propulsion Laboratory.
- Ogle SM, Domke G, Kurz WA, Rocha MT, Huffman T, Swan A, Smith JE, Woodall C,
- Krug T. 2018. Delineating managed land for reporting national greenhouse gas emissions and

- removals to the United Nations framework convention on climate change. Carbon Balance and
- 535 *Management* **13**: 9.
- Pan Y, Birdsey RA, Fang J, Houghton R, Kauppi PE, Kurz WA, Phillips OL, Shvidenko A,
- Lewis SL, Canadell JG, et al. 2011. A Large and Persistent Carbon Sink in the World's Forests.
- 538 Science **333**: 988–993.
- Ouegan S, Le Toan T, Chave J, Dall J, Exbrayat J-F, Minh DHT, Lomas M, D'Alessandro
- 540 MM, Paillou P, Papathanassiou K, et al. 2019. The European Space Agency BIOMASS
- 541 mission: Measuring forest above-ground biomass from space. Remote Sensing of Environment
- 542 **227**: 44–60.
- Réjou-Méchain M, Tanguy A, Piponiot C, Chave J, Hérault B. 2017. Biomass: An r package
- for estimating above-ground biomass and its uncertainty in tropical forests. *Methods in Ecology*
- 545 and Evolution 8: 1163–1167.
- Requena Suarez D, Rozendaal DMA, Sy VD, Phillips OL, Alvarez-Dávila E, Anderson-
- Teixeira K, Araujo-Murakami A, Arroyo L, Baker TR, Bongers F, et al. 2019. Estimating
- aboveground net biomass change for tropical and subtropical forests: Refinement of IPCC
- default rates using forest plot data. Global Change Biology 25: 3609–3624.
- Roe S, Streck C, Beach R, Busch J, Chapman M, Daioglou V, Deppermann A, Doelman J,
- 551 Emmet-Booth J, Engelmann J, et al. 2021. Land-based measures to mitigate climate change:
- Potential and feasibility by country. *Global Change Biology* **27**: 6025–6058.
- Rozendaal DMA, Suarez DR, Sy VD, Avitabile V, Carter S, Yao CYA, Alvarez-Davila E,
- Anderson-Teixeira K, Araujo-Murakami A, Arroyo L, et al. 2022. Aboveground forest
- biomass varies across continents, ecological zones and successional stages: Refined IPCC default
- values for tropical and subtropical forests. Environmental Research Letters 17: 014047.
- Sanderman J, Hengl T, Fiske GJ. 2017. Soil carbon debt of 12,000 years of human land use.
- 558 Proceedings of the National Academy of Sciences 114: 9575–9580.
- 559 Stall S, Yarmey L, Cutcher-Gershenfeld J, Hanson B, Lehnert K, Nosek B, Parsons M,
- Robinson E, Wyborn L. 2019. Make scientific data FAIR. *Nature* 570: 27–29.

561 Tifafi M, Guenet B, Hatté C. 2018. Large Differences in Global and Regional Total Soil 562 Carbon Stock Estimates Based on SoilGrids, HWSD, and NCSCD: Intercomparison and 563 Evaluation Based on Field Data From USA, England, Wales, and France. Global 564 Biogeochemical Cycles **32**: 42–56. 565 UNFCCC. 2015. Adoption of the Paris Agreement.: 31. 566 Waring BG, Powers JS. 2017. Overlooking what is underground: Root:shoot ratios and coarse 567 root allometric equations for tropical forests. Forest Ecology and Management 385: 10–15. 568 Warner E, Cook-Patton SC, Lewis OT, Brown N, Koricheva J, Eisenhauer N, Ferlian O, 569 Gravel D, Hall JS, Jactel H, et al. 2022. Higher aboveground carbon stocks in mixed-species 570 planted forests than monocultures a meta-analysis.: 2022.01.17.476441. 571 Xu L, Saatchi SS, Yang Y, Yu Y, Pongratz J, Bloom AA, Bowman K, Worden J, Liu J, Yin 572 Y, et al. 2021. Changes in global terrestrial live biomass over the 21st century. Science Advances 573 7: eabe9829.