

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

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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 August 2024, ForC contained ~19286 records potentially 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 millennia 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), a 62 relationship that has come into increasing focus in recent decades as anthropogenic land 63 transformation and climate change have accelerated (IPCC, 2019a). Forests play a substantial 64 role in international plans for climate change mitigation under the Paris Agreement (UNFCCC, 65 2015; Grassi et al., 2017). Forest conservation, reforestation, and improved sustainable 66 management all have significant – and relatively cost-effective – potential as climate change 67 mitigation options (Roe et al., 2021). Conservation and reforestation have the fourth and fifth 68 largest net emission reduction potentials or all mitigation options (IPCC, 2022). However, 69 envisioned forest-based climate change mitigation initiatives do not always correspond to actual 70 emission reductions implemented on the ground (e.g., Badgley et al., 2022). Realistic planning 71 and reporting is critically needed to ensure that forest-based climate change mitigation initiatives 72 are effective, and this requires solid scientific data and accounting frameworks (Deng et al., 73 2021; Anderson-Teixeira & Belair, 2022). 74 To this end, the International Panel on Climate Change (IPCC) provides guidance for national 75 greenhouse gas inventories for reporting to the United Nations Framework Convention on 76 Climate Change (UNFCCC) (IPCC, 2006a; IPCC, 2019b). Under this guidance, greenhouse gas 77 emissions to, or withdrawals from, the atmosphere are quantified on an annual basis for all 78 managed land, which includes most of the world's forest land (Ellis et al., 2010; Ogle et al., 79 2018). The IPCC inventory guidelines include specific instructions for inventories for 80 greenhouse gas (mainly CO<sub>2</sub>) exchanges between forest land and the atmosphere (Notes S1, 81 IPCC, 2006b, 2019b). A tiered approach is employed, where the lowest tier (Tier 1) represents 82 the simplest approach and relies on default parameter values – for example, forest carbon (C) 83 stocks values by ecozone and forest age class derived as the average of published estimates 84 (IPCC, 2019b; Rozendaal et al., 2022). Tier 1 values have improved over the years as more data 85 and methods have become available (Requena Suarez et al., 2019; Rozendaal et al., 2022), but 86 there remains room for improvement. For example, following the 2019 release of the latest IPCC 87 guidelines, it was revealed that IPCC's Tier 1 default failed to capture eight-fold variation of C 88 accumulation in regrowth forests within ecozones (Cook-Patton et al., 2020) and that C stocks in

89 mature African tropical montane forests were two-thirds higher than the IPCC Tier 1 values for 90 these forests (Cuni-Sanchez et al., 2021). High variability of forest C cycling within ecozones 91 (e.g., Cook-Patton et al., 2020; Cuni-Sanchez et al., 2021) means that it is crucial for 92 practitioners to have access to locally-specific information, when available. This rapid evolution 93 of scientific information on C cycling in forests is valuable for informing climate change 94 mitigation efforts but requires improved mechanisms for communicating the latest information 95 from scientific researchers to the practitioners who need reliable estimates for greenhouse gas 96 mitigation planning. 97 To improve data accessibility for preparing greenhouse gas estimates, the IPCC created the 98 Emission Factor Database (EFDB; https://www.ipcc-nggip.iges.or.jp/EFDB/main.php), which is 99 intended as a recognized library of emission factors and other parameters that can be used for 100 estimating greenhouse gas emissions and removals. The EFDB can be used both for efforts to 101 tally a nation's intended or accomplished greenhouse gas reductions, or as a basis of comparison 102 for external parties to evaluate these inventories. The EFDB encourages researchers to submit 103 estimates of emission factors or other related parameters (e.g., C stocks, net annual increments, 104 and annual fluxes for various pools, IPCC, 2006a; IPCC, 2019b) from peer-reviewed journal 105 articles or other accepted sources for inclusion in the database. Tens of thousands of relevant 106 forest carbon estimates have been published – and continue to be published at an accelerating 107 rate – but are not readily accessible to the practitioners assembling national greenhouse gas 108 inventories. To contribute to the goal of making forest C parameters available for accounting 109 under IPCC guidelines, forest scientists need an accessible summary of EFDB's requirements 110 and an efficient system for submission of data to the EFDB. 111 Our goal is to facilitate submission of forest C estimates from scientific studies to EFDB. We 112 document the process of submitting data to EFDB from the Global Forest Carbon Database, 113 ForC (https://forc-db.github.io/), which is the largest collection of published estimates of forest C 114 stocks, increments, and annual fluxes (Anderson-Teixeira et al., 2018, 2021; Anderson-Teixeira 115 et al., 2023), including data ingested from individual publications and relevant databases, 116 including the Global Reforestation Opportunity Assessment (GROA) database (Cook-Patton et 117 al., 2020, database doi: 10.5281/zenodo.3983644), and the Global Soil Respiration Database 118 (SRDB-V5, Bond-Lamberty & Thomson, 2010; Jian et al., 2021). We (1) map common

119 scientific forest C estimation methods and definitions to those used by the IPCC; (2) develop a 120 semi-automated process for preparing ForC data for submission to EFDB; and (3) assess the data 121 in ForC potentially relevant to EFDB and records that have been submitted to date. We conclude 122 with recommendations as to how the scientific community can better provide useful data for 123 forest C inventories under the Paris Agreement. 124 **Materials and Methods** 125 Major steps for submission of data from ForC to EFDB included (1) mapping ForC into EFDB, 126 including aligning ForC terms and concepts with those defined by IPCC guidelines (summarized 127 in Notes S1), (2) revising For V3.0 to support semi-automated submissions to EFDB, yielding 128 For Cv4.0 (detailed in Methods S1), and (3) submitting data to EFDB. 129 1. Mapping ForC to EFDB 130 With input from the EFDB Technical Support Unit and referencing IPCC guidance (IPCC, 2003, 131 2019b; IPCC, 2006a), we determined how EFDB fields should be populated using ForC fields 132 (summarized in Table S2). Fields in EFDB included several fields describing how the record fits 133 within IPCC's framework (source/sink category, greenhouse gas type, C pool, relevant 134 equations), several describing the C estimate itself (variable, value, units, 95% confidence 135 limits), a few composite fields describing biotic and abiotic conditions (e.g., vegetation type, 136 minimum diameter included, site location, climate, edaphic properties, notable disturbances, 137 stand age, plot information), and a few describing the source (e.g., type, citation, data provider). 138 Most relevant ForC fields mapped directly into an EFDB field, either as the only contents of that 139 field or as part of a composite record. For some fields, simple conditional logic was used to 140 populate EFDB fields based on ForC records. For example, in cases where original studies did 141 not present 95% confidence intervals (required by IPCC when available) but did present standard 142 error or n and standard deviation, we calculated the 95% confidence intervals. There were two 143 cases in which more complex mapping was required: (1) mapping of C cycle variables and (2)

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land classification.

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Carbon cycle variables

Working in consultation with the EFDB Technical Support Unit and referencing IPCC guidance (IPCC, 2003, 2019b; IPCC, 2006a), we identified ForC variables that were relevant to the IPCC methodology and EFDB (Fig. 1, 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-defined 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, ForC records 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-EFDB-integration). To the second se

### 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

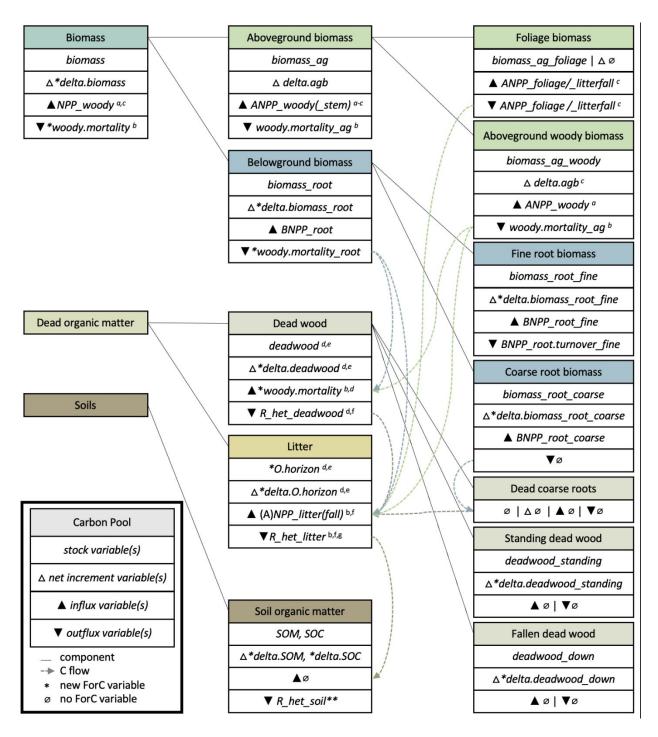


Figure 1. Schematic illustrating the carbon pools defined under IPCC Guidelines for national greenhouse gas inventories; corresponding ForC variables, and relationships among them.

For each C pool, we show ForC variables corresponding to the stock, net annual increment, influx, and outflux. Most, but not all, of the 40 EFDB-relevant ForC variables are shown here.

Correspondence of ForC variables to IPCC criteria often depends upon measurement protocols

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

196 land". Forests <20 years old with unknown land use prior to the study date were simply 197 classified as "Forest Land". The same logic was applied for savannas, but including both Forest 198 Land and Grassland as potentially relevant categories. 199 Given the lack of public information needed to determine whether lands are classified as 200 managed (Ogle et al., 2018; Deng et al., 2021), and because the IPCC's definition of managed 201 land is more expansive than is commonly applied in the scientific literature and hence in ForC, 202 we did not include any classification of land management status from ForC in the records 203 submitted to EFDB. However, we did provide auxiliary information that should be useful in 204 making this determination, including geographical location and notable disturbance events. 205 2. Updating ForC 206 Previous versions of ForC (Anderson-Teixeira et al., 2016, 2018, 2021) contained most of the 207 information required by EFDB, and, more broadly, to inform C stock change calculations under 208 the Paris Agreement. However, modest changes to the structure and contents of ForC were 209 needed in order to provide all information required by EFDB and to improve ForC's capacity to 210 serve as a repository of valuable information for forest C inventories under IPCC guidelines. To 211 support export of data to EFDB, and to improve the overall quality of the ForC database, we 212 added or modified 18 fields (Table S1), defined 15 new variables, implemented enhanced quality 213 control, manually reviewed >1968 records to obtain additional required information, and added 214 329 new records. Having implemented these changes, which are described in detail in Methods 215 S1, to ForC v3.0 (Anderson-Teixeira et al., 2021), we released a new major version: ForC v4.0. 216 3. Submission of ForC data to EFDB 217 To submit complete, reviewed ForC records to EFDB, we created R scripts to restructure ForC 218 records and populate EFDB's bulk import form (publicly available; see Data Availability 219 **Statement**). Criteria for data submission were that (1) records had been checked against the 220 original study and determined to be complete and correct, and as originally presented, (2) all 221 relevant records within the publication were included in ForC, (3) the original study presented 222 values in tables or text, as opposed to the values having been digitized from graphs or calculated 223 based on related variables, and (4) the records had not previously been submitted to EFDB. 224 Review of potentially relevant records was prioritized as described in Methods S1. Once

225 converted into EFDB format, the records were reviewed and then sent to the IPCC's Technical 226 Support Unit for their review and submission to EFDB. Submissions were sent in six batches 227 between April 2021 and June 2022, and feedback on each batch from the EFDB team was used 228 to improve our system for subsequent submissions. 229 **Results** 230 As of August 2024, we had submitted to EFBD only a fraction ( $\sim$ 7%) of potentially relevant 231 For Crecords (i.e., records for EFDB-relevant variables), but these already comprised a 232 substantial portion (19%) of EFDB records for forest land. Specifically, of the 39848 records in 233 ForC (v4.0), ~19286 were independent records of EFDB-relevant variables (Fig. 1, Table S3). 234 We reviewed 1968 records for submission to EFDB, of which 43% were inappropriate, mostly 235 because values were digitized from graphs or calculated from related variables rather than 236 directly presented. We also added 329 new records, mostly because EFDB required that all 237 relevant records in a publication be submitted together. Of 1443 submitted records, EFDB's 238 review panel accepted and posted 1068 but deemed 26% not applicable. Reasons included – but 239 were not limited to – non-applicability to the IPCC methodology of the variable submitted (e.g., 240 net ecosystem CO<sub>2</sub> exchange, litter - OL layer; subsequently excluded from lists of relevant 241 variables and counts of potentially relevant records), inadequate information on the quality of 242 data and thus on its replicability (e.g., confidence interval/uncertainty), and imprecisely 243 described disturbance histories (e.g., "moderately"/ "severely" burned). The iterative process of 244 review by EFDB's Technical Support Unit and review panel significantly improved our process, 245 though it did not achieve a 100% acceptance rate. 246 For Cv4.0 records, including the subset submitted to EFDB, were unevenly distributed across all 247 forested continents, biomes, and forest types (Figs. 2-4). Relative to forested area, submitted data 248 under-represented the tropics and over-represented temperate regions, reflecting the composition 249 of ForC (Figs. 2, 4b). Submitted data were concentrated in climates that are relatively common 250 over Earth's land area and covered all forested climates with substantial land area except the 251 colder boreal regions (boreal tundra woodland; Figs. 2, 3). Among the records submitted to 252 EFDB, the largest number came from North America, followed by Asia, South America, and 253 Africa, with strongly disproportionate representation of North America and under-representation

of South America, Africa, Europe, Australia, and Oceania (Fig. 4c). In terms of FAO ecozones, boreal coniferous forest, temperate continental forest, and temperate mountain systems had the most records and were the best represented relative to their global area, whereas other ecozones were poorly represented (Fig. 4b). By far the most records and best representation relative to global area were for needleleaf evergreen forests, followed by broadleaf deciduous and broadleaf evergreen (Fig. 4b). The largest number of submitted records came from mature forests (>100 years), but young (<20 year) stands were best represented relative to their global area, while 20-100 year stands were under-represented (Fig. 4d). For C contained the records needed to provide more balanced geographical and forest type representation via selective data submission (Figs. 2-4).

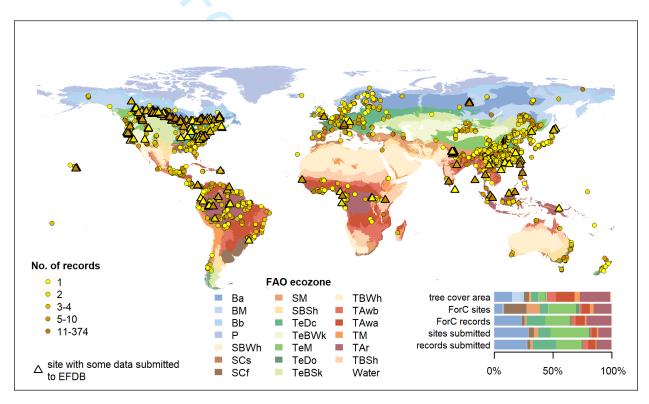


Figure 2. Map of sites in ForC shaded by number of independent records potentially relevant (circles) and submitted (triangles) to EFDB. Underlying map shows FAO ecozones, and symbols are colored according to the number of records at each site. Inset stacked bar chart shows proportional representation by FAO ecozone for tree cover area (from SYNMAP, Jung et al, 2006), potentially relevant ForC records, and submitted records. FAO ecozones (obtained from FAO's GEOnetwork: http://www.fao.org:80/geonetwork) 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.

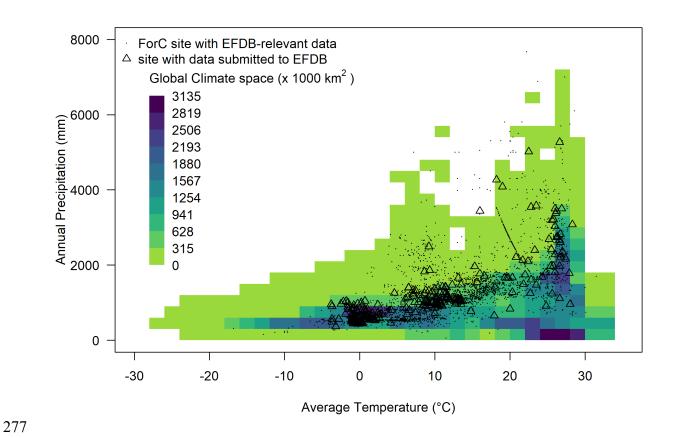


Figure 3. Distribution of ForC records potentially relevant (dots) and submitted (triangles) to EFDB within the global climate space of mean annual temperature (MAT) and mean annual precipitation (MAP). Climate data are from the CRU TS3.10 dataset (0.5° resolution, 1990-2014, Harris et al, 2014). Background colors indicate the global land area with each MAT-MAP combination.

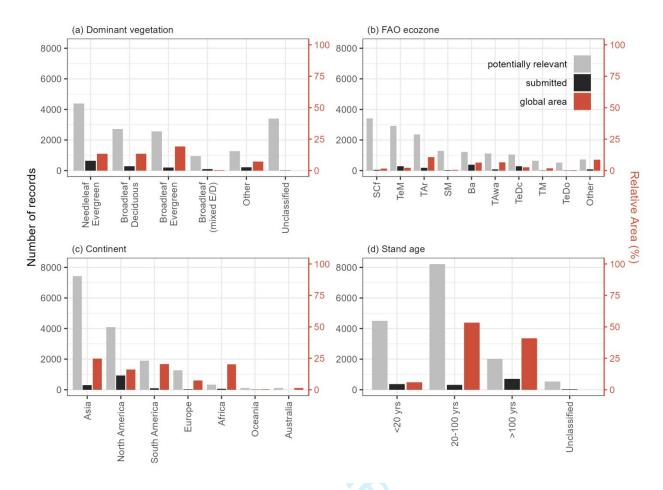


Figure 4. Histograms of number of independent records in ForC potentially relevant (grey) and submitted (black) to EFDB (both left axis), along with the relative global area of each type (red, right axis), organized by (a) dominant vegetation type, (b) FAO ecozone, (c) continent, and (d) stand age. For dominant vegetation (a), 'Other' includes deciduous needleleaf, mixed broadleaf- needleleaf, non-woody vegetation (e.g., early successional), and incompletely classified or mixed forest types. Global coverage of each was obtained from SYNMAP (Jung et al, 2006). For FAO ecozones (b), codes are as listed in the caption of Figure 2. The relative area of forests by age class was obtained from the global forest age database of Besnard et al. (2021, 2024).

In terms of variables, ForC v4.0 contained records for 28 of the 40 variables (or closely-related variable groups) relevant to EFDB (Fig. 1, Table S3). The records submitted to EFDB (18 variables) were very unevenly distributed across variables, mirroring the composition of ForC (Table S3). The majority (81%) of records submitted were for C stocks, including 3% for total

297 biomass, 53% for aboveground biomass, 4% for components of aboveground biomass (wood or 298 foliage), 6% for root biomass or its components, 8% for dead wood or its components, 3% for 299 litter (O horizon), and 4% for soils. Increment records (mostly for aboveground biomass) totaled 300 9% of records submitted. The remaining 9% of records submitted described fluxes, all of which 301 were either inputs or outputs to the aboveground biomass pool, with some also describing inputs 302 to dead wood or litter pools (Fig. 1, Table S3). Many of the EFDB-relevant variables remain 303 poorly represented in ForC (Table S3). 304 **Discussion** 305 We developed a framework for submitting data from ForC v4.0, the largest compilation of 306 ground-based forest C estimates, to the EFDB, making the data more accessible for reporting 307 CO<sub>2</sub> emissions and removals from forest land in line with IPCC guidelines (IPCC, 2006a; IPCC, 308 2019b). Although the process of preparing ForC data for EFDB submission requires manual 309 review of records, most of the pertinent information is already entered in ForC, and complete 310 records can be automatically converted to EFDB's format using the system developed here. As 311 of August 2024, 1068 For Crecords have been posted to EFDB, which comprises 19% of the 312 total EFDB records for forest land. 313 Based on our experience contributing forest C data to EFDB via ForC, we make several 314 recommendations as to how scientists can improve forest C records in EFDB through database 315 work, new data collection and analysis, and improved reporting. 316 Potential for forest C databases to contribute to EFDB 317 There is vast potential to expand forest C data in EFDB by reviewing and submitting records 318 from ForC (Figs. 2-4). So far, only ~7% of the potentially relevant records in ForC have been 319 submitted to EFDB. These are unevenly distributed across regions, forest types, and variables 320 (Figs. 2-4, Table S3), reflecting the uneven composition of ForC (Anderson-Teixeira et al., 321 2021) and the fact that our effort was focused on developing an accurate and efficient data 322 submission system rather than optimizing the composition of submitted data. Future efforts to 323 submit ForC records to EFDB should optimize for representation across geographic regions, 324 forest types, and variables, giving priority to those that are currently under-represented (Figs. 2-

325	4, Table S3). Other categories of records to prioritize include those from countries relying on
326	existing data for their greenhouse gas inventories (Tier 1 or 2 methodology) and more recent
327	records. Coverage of many variables or regions could be vastly improved through systematic
328	review of the literature, although this requires focused and extensive manual effort. Recent
329	efforts have compiled large databases of relevant data, compatible with ForC, from monoculture
330	plantation forests (Bukoski et al., 2022) and mixed species plantation forests (Warner et al.,
331	2022; Feng et al., 2022), and such a compilation is in works for agroforestry (Susan Cook-
332	Patton, unpublished data).
333	While submitting records to EFDB from ForC or other databases is by far the most efficient
334	approach to expanding forest C data in EFDB, it does require time-consuming database work
335	(e.g., review of records based on original publications, adding missing information). As the IPCC
336	is not able to pay for data, such work will require support from research and government
337	institutions, funding agencies, and non-governmental organizations.
338	Forest C data collection and analysis needs
339	While further database work could vastly improve EFDB's coverage of forest C data, new data
340	collection and analyses will be essential to providing accurate global data on forest C stocks and
341	changes therein. Even for for aboveground biomass stocks, which have received by far the most
342	research attention (Table S3, NISAR, 2018; Quegan et al., 2019; Dubayah et al., 2020;
343	Anderson-Teixeira et al., 2021), production of an accurate global map of forest C stocks remains
344	an ongoing challenge (Araza et al., 2023). Other variables – particularly those characterizing
345	belowground C pools – remain poorly quantified and highly uncertain for many parts of the
346	world (Table S3, Tifafi et al., 2018; Anderson-Teixeira et al., 2021), introducing substantive
347	uncertainties into global forest C budgets (Pan et al., 2011; Harris et al., 2021). Furthermore,
348	data distribution is very uneven across forest types and geographical regions (Figs. 2-4); for
349	instance, data on C cycling of tropical forests – particularly in Africa – remains relatively sparse,
350	in large part due to substantial barriers to data collection and distribution (de Lima et al., 2022).
351	Significant investment in research and researchers focused on ground-based measurement of
352	forest C in such regions will be important to filling knowledge gaps in forest C cycling (de Lima
353	et al., 2022; Labrière et al., 2023; Araza et al., 2023).

354	Several EFDB-relevant variables have not been calculated and presented as frequently as would
355	be possible given existing forest census data and minimal extra research effort. For example,
356	aboveground woody mortality and aboveground biomass increment can be calculated from the
357	same census data as aboveground woody productivity, yet the latter has received far more
358	research attention (Table S3, Anderson-Teixeira et al., 2021). Similarly, live coarse root
359	biomass, total biomass, and changes therein could in theory be estimated in parallel with
360	aboveground biomass, with the greatest barrier being that allometric models for estimating root
361	biomass are not as reliable or easily available as are those for aboveground biomass (Chave et
362	al., 2014; Réjou-Méchain et al., 2017; Gonzalez-Akre et al., 2022). However, while equations
363	for estimating root (and thereby total) biomass require improvement, they do exist for many
364	forest types (Mokany et al., 2006; e.g., Brassard et al., 2011; Chojnacky et al., 2014; Waring &
365	Powers, 2017). We recommend that, when possible and relevant to their own study, researchers
366	calculate and report all EFDB-relevant variables – a goal that could be facilitated by
367	development of relevant analytical tools (sensu Chave et al., 2014; Réjou-Méchain et al., 2017;
368	Gonzalez-Akre et al., 2022) and encouraged by scientific journals.
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369 370 371 372 373 374 375 376 377 378 379	Recommendations of good practices for reporting forest C stocks and stock changes  We recommend that, in order to make research most useful for estimating C stock changes following IPCC guidelines, researchers calculate and report results according to IPCC good practice (Table 2). Importantly, simple decisions on the presentation of results will determine whether the records meet criteria for inclusion in EFDB. For example: (1) presenting data only in a figure makes them ineligible for inclusion in EFDB, whereas presentation in a table or supplementary data file allows inclusion while supporting FAIR goals (Stall <i>et al.</i> , 2019); (2) direct presentation of all relevant variables allows inclusion, whereas presenting only components of variables of interest (e.g., parsing litter into fine woody debris, OL, OF, and OH 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)

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of 30 cm). It should also be emphasized that reporting of 95% confidence intervals (or other metrics of error), when applicable, is highly desirable and makes the data more relevant to IPCC. Researchers compiling published records (e.g., for meta-analyses) can increase the value of their data set by ensuring that 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). Ily da.
Igencies. Improved reporting – and potentially data submission to EFDB – could be encouraged by journal editors, reviewers, and funding agencies.

# 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 calculated.
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.

#### **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 on ground-based measurement of forest C will be required to fill knowledge gaps and thereby increase the accuracy of C inventories for forest lands under the Paris Agreement. This challenge is heightened by the fact that forests are changing rapidly (e.g., McDowell *et al.*, 2020), making data collected a decade or more in the past increasingly unreliable. This heightens the need for an efficient system of making forest C data accessible for national greenhouse gas inventories. We view the system developed here for submitting ForC data to the IPCC EFDB as one important step towards that goal.

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415	Author Contribution
416	KAT and VH conceived and designed the project; VH wrote the scripts for database
417	management, data submission to EFDB, and the analyses presented here; MW, TR, and RBM
418	added and reviewed ForC data, BBL and SCP contributed large databases to ForC (EFDB and
419	GROA, respectively); CP provided methodological expertise; KAT, VH, and MW prepared the
420	first draft of the manuscript; all authors reviewed the results and approved the final version of the
421	manuscript.
122	Data Availability Statement
123	All code and data are openly available. The ForC database and associated code are available via
124	the ForC repository within the ForC-db organization on GitHub (https://github.com/forc-
125	db/ForC), and the version used here (ForC v4.0) is archived in Zenodo (Anderson-Teixeira et al.,
426	2023, DOI: 10.5281/zenodo.8020861). The data and code associated with data submission to
127	EFDB and preparation of this manuscript are available via the IPCC-EFDB-integration
428	repository within the ForC-db organization on GitHub (https://github.com/forc-db/IPCC-EFDB-
129	integration) and archived in Zenodo (DOI: 10.5281/zenodo.8021474).
430	Supplementary Information
431	Table S1. Updates to ForC field implemented between releases of v3.0 and v4.0

- 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
- 435 Methods S1. Updates to ForC (ForC v4.0)
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