

1 **Decomposability of soil organic matter over time: The Soil**
2 **Incubation Database (SIDb, version 1.0) and guidance for**
3 **incubation procedures**

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31 **Abstract**

32 The magnitude of carbon (C) loss to the atmosphere via microbial decomposition is a function of
33 the amount of C stored in soils, the quality of the organic matter, and physical, chemical and
34 biological factors that comprise the environment for decomposition. The decomposability of C is
35 commonly assessed by laboratory soil incubation studies that measure greenhouse gases
36 mineralized from soils under controlled conditions. Here, we introduce the Soil Incubation
37 Database (SIDb) version 1.0, a compilation of time series data from incubations, structured into a
38 new, publicly available, open access database of C flux (carbon dioxide, CO₂, or methane, CH₄).
39 In addition, the SIDb project also provides a platform for the development of tools for reading
40 and analysis of incubation data as well as documentation for future use and development. In
41 addition to introducing SIDb, we provide reporting guidance for database entry and the required
42 variables that incubation studies need at minimum to be included in SIDb. A key application of
43 this synthesis effort is to better characterize soil C processes in Earth system models, which will
44 in turn reduce our uncertainty in predicting the response of soil C decomposition to a changing
45 climate. We demonstrate a framework to fit curves to a number of incubation studies from
46 diverse ecosystems, depths, and organic matter content using a built-in model development
47 module that integrates SIDb with the existing SoilR package to estimate soil C pools from time
48 series data. The database will help bridge the gap between point location measurements, which
49 are commonly used in incubation studies, and global remote-sensed data or data products derived
50 from models aimed at assessing global-scale rates of decomposition and C turnover. The SIDb
51 version 1.0, is archived and publicly available at [DOI: 10.5281/zenodo.3871263](https://doi.org/10.5281/zenodo.3871263) (Sierra et al.,
52 2020) and the database is managed under a version-controlled system and centrally stored in
53 GitHub (<https://github.com/SoilBGC-Datashare/sidb>).

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62 **1 Introduction**

63 Temperature, soil moisture, soil type, plant-microbe interactions, microbial community
64 compositions, physical protection of organic matter (e.g., sorption on minerals and aggregation)
65 and physical disconnection of microbes/enzymes and their substrates all control microbial
66 decomposition processes and fluxes of greenhouse gases to the atmosphere (Conant et al., 2011;
67 Schmidt et al., 2011). The relative importance of all these factors in controlling decomposition
68 processes is poorly quantified but is important to understand as warming temperatures shift rates
69 of microbial processes, potentially increasing releases of soil-stored carbon (C) to the
70 atmosphere (Davidson and Janssens, 2006).

71 Research synthesis (e.g. meta-analysis) has become an increasingly important tool in
72 science to overcome site-specific results, identify universal patterns across ecosystems and at
73 global scales, and to assess what is known and what needs further research (Gurevitch et al.,
74 2018; Gurevitch and Hedges, 1999; Hillebrand and Gurevitch, 2013; Osenberg et al., 1999).
75 Numerous reviews, syntheses, and meta-analyses have been performed using laboratory
76 incubation studies (e.g. Conant et al., 2011; Hamdi et al., 2013; Schädel et al., 2014, 2016; Treat
77 et al., 2015) to answer questions about the relative decomposability or stability of soil organic
78 matter, the temperature response of soil respiration, and the ratio of CO₂:CH₄ production in
79 anaerobic incubations. New experiments are continuously contributing to the growing body of
80 soil incubation literature. While individual soil incubation studies are performed to answer
81 specific research questions that may not require measuring a large variety of variables, the more
82 details that are provided and the more comprehensive the meta-data are, the greater the utility of
83 an individual study beyond its original use (Hillebrand and Gurevitch, 2013). Metadata help to
84 characterize these data sets, enable identification of data through relevant criteria, and provide
85 the information needed for data archiving (Hillebrand and Gurevitch, 2013; Jiang et al., 2015)
86 making individual incubation studies as useful as possible.

87 Here, we report on the development and compilation of a subset of available incubation
88 data into a new, publicly available Soil Incubation Database (SIDb). In addition to introducing
89 SIDb, we provide clear reporting guidance for database entry and the required variables that
90 incubation studies need at minimum to be included in SIDb. Further, we provide guidance and
91 associated recommendations to help inform best practices for conducting consistent, comparable

92 soil incubation studies while retaining the adaptability required for individual research groups
93 and projects.

94 A key application of this synthesis effort is to better characterize soil C processes in Earth
95 system models, which will in turn reduce our uncertainty in predicting the response of soil C
96 decomposition to a changing climate. Soil C decomposition is traditionally represented by a
97 simple first-order decay function (Jenkinson, 1990) in C cycle models assuming one or more
98 conceptual C pools (Davidson and Janssens, 2006; Parton et al., 1987; Trumbore, 1997) with fast
99 and slower rates of C turnover. The models are described by several parameters such as the
100 decay rate of each pool, as well as the transfer rates among pools. These parameters can be
101 utilized to predict the evolution of CO₂ one would observe in an incubation over time. Incubation
102 time series data could therefore be used to constrain the parameters of these models by solving
103 the corresponding inverse problem.

104 We demonstrate a framework to fit such curves to a number of incubation studies from
105 diverse ecosystems, depths, and organic matter content using a built-in model development
106 module that integrates SIDb with the existing SoilR package (Sierra et al., 2012) to estimate soil
107 C pools from time series data. This allows users to test different model structures against their
108 data, representing a benefit of contributing data to SIDb. We hope the database will help bridge
109 the gap between localized measurements, which are commonly used in incubation studies, and
110 global remote-sensed data or data products derived from models aimed at assessing global-scale
111 rates of decomposition and C turnover (Carvalhais et al., 2014; Koven et al., 2017). This work
112 also complements other compilations of soil C related datasets such as the International Soil
113 Carbon Network (<https://iscn.fluxdata.org/>), the open source Continuous Soil Respiration
114 database, COSORE, (<https://github.com/bpbond/cosore>), the Global Database of Soil Respiration
115 Data, Version 4.0 (Bond-Lamberty and Thomson, 2018), and the International Soil Radiocarbon
116 Database (ISRaD, soilradiocarbon.org; Lawrence et al., 2020).

117

118 **2 Laboratory incubations as a tool to assess soil C decomposability**

119 Laboratory soil incubation studies are a commonly used method to estimate the decomposability
120 of soil organic matter by measuring greenhouse gas release as C is mineralized from soils under
121 controlled conditions. Results from incubation studies can inform global models about C pool
122 sizes and rates of soil organic matter processing (mostly derived from long-term incubations) and

123 sensitivities of process rates with respect to changes in abiotic factors such as soil temperature,
124 moisture, pH, etc. Incubation durations may vary from less than one day to up to many years.
125 Short-term incubations (a few days to a few months) provide information on how much C is
126 readily decomposable and may be closer to the initial conditions experienced within the soil
127 profile. Long-term incubations (months to years) may diverge from the conditions found within
128 the profile, but can give insights into the potential decomposability of slower cycling C (e.g.
129 Schädel et al., 2014). At the beginning of laboratory incubations, respiration of fast cycling C
130 dominates total C respired, but it declines rapidly, whereas slow cycling C accounts for most of
131 the C being respired after the fast C pool is mostly depleted (Figure 1). In this respect, laboratory
132 incubations serve as a method to biologically partition soil C into different kinetic pools using
133 the microbes themselves as the main partitioning agent. The time series produced is often well
134 approximated by a sum of exponential functions, which are the solution of systems of first-order
135 linear differential equations with constant coefficients (Metzler and Sierra, 2018). Fitting data
136 from incubations to these types of functions has been done for individual site-level studies (e.g.
137 Schädel et al., 2013, 2014; Sierra et al., 2017).

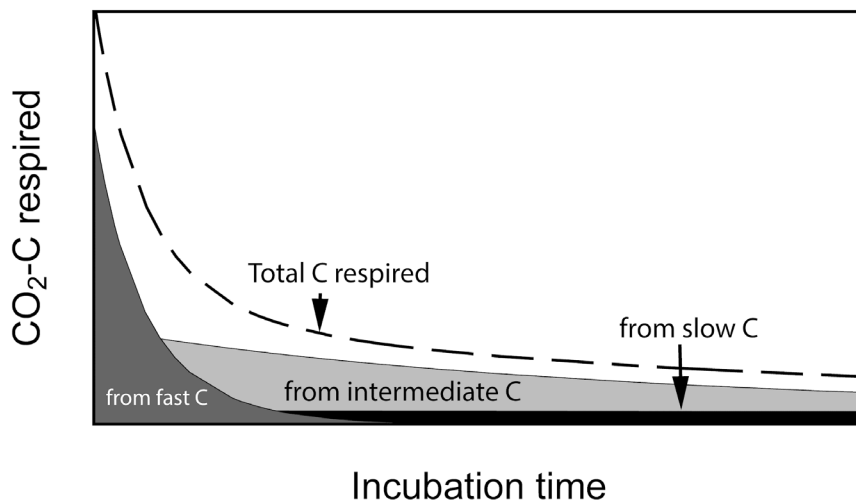


Figure 1: Conceptual figure of C respiration during aerobic soil incubations. Total CO₂-C flux is composed of contributions from different C pools which changes over time. Fast cycling C dominates total CO₂-C flux at the beginning of the incubation and is later replaced by slower cycling C pools.

138
139 Like all methods, incubations have their advantages and disadvantages. Many laboratory
140 methods exist for splitting soil C into pools of various purported stabilities (e.g. density

141 fractionation (Sollins et al., 2006), sequential extraction (Heckman et al., 2018), and thermal
142 analysis (Barré et al., 2016)), but incubations are the only biological assay for testing soil C
143 stability, an ultimately biological process. Carbon stability is a measure of how resistant and
144 inaccessible organic molecules are to microbial decay.

145 Another distinct advantage of incubations is the high level of control they allow, as
146 compared to field methods. For example, incubations that test the temperature sensitivity of C
147 flux (e.g. Bracho et al., 2016; Conant et al., 2008) offer a greater level of control compared to
148 field measurements in several ways. First, in situ soil respiration is a mixture of both
149 heterotrophic microbial respiration and autotrophic root respiration; soil incubations isolate the
150 heterotrophic flux. Second, in situ temperatures change daily and seasonally thereby
151 confounding any direct effects of temperature with the phenology of C inputs such as root
152 exudates and litter fall. At many locations, such as those under Mediterranean climate regimes,
153 temperature is highly correlated with soil moisture so that the effects of one are impossible to
154 disentangle from the other (Sierra et al., 2015; Subke and Bahn, 2010). With incubations,
155 temperature and moisture effects can be tested both in isolation and with interactions.
156 Incubations are a tractable and accessible method that can be run with minimal equipment (scale,
157 gas-tight jars that seal, and a CO₂ analyzer). Much of the utility of incubations lies in their
158 simplicity. Lastly, as described above, the time series data collected by most incubations can be
159 connected to soil C models (Sierra et al., 2012, 2014).

160 The main shortcoming of incubations is their isolation from the soil ecosystem.
161 Incubations lack new inputs, which could otherwise prime the decomposition of the existing soil
162 C pool (Huo et al., 2017). However, the lack of inputs simplifies the system and allows a focus
163 on decay processes. Substrates can be added to incubations to measure the decomposability of
164 specific compounds or materials (particularly if they are isotopically labeled), or to measure the
165 priming effect under experimentally controlled conditions, a common extension of incubation
166 methods (e.g. Finley et al., 2018; Pegoraro et al., 2019). Additionally, the microbial community
167 in incubations may not reflect in situ communities. For example, constant environmental
168 conditions in incubations may reduce the available niches and potentially result in a decline of
169 microbial diversity—an effect that has yet to be tested. The lack of inputs can also induce
170 changes in the microbial community as more oligotrophic microbes are favored over time.
171 Lastly, soils used in incubations are always disturbed to varying degrees during removal from the

172 field and often further in the laboratory: during sieving or root-picking procedures, or through re-
173 wetting prior to the start of the incubation. For example, at the time of publication, half of the
174 studies in our database reported sieving prior to incubation, while a third do not report pre-
175 incubation procedures. This disturbance may increase the susceptibility of occluded soil C to
176 decay via disruption of aggregates, potentially overestimating the amount of C released during
177 incubations relative to field conditions (Salomé et al., 2010). In general, the experimental control
178 of incubations allows for most of these criticisms to be explicitly tested and accounted for as
179 needed, and overall, the advantages of incubations far outweigh their drawbacks when the goal is
180 understanding C pool structure, C stability and C sensitivity to drivers such as temperature and
181 moisture.

182

183 **3 The Soil Incubation Database (SIDb)**

184 The Soil Incubation Database (SIDb) version 1.0 is an open source software project that provides
185 open access to data and is a platform for the development of tools for reading and analysis of
186 data as well as documentation for future use and development. The data is freely available at
187 DOI: 10.5281/zenodo.3871263 (Sierra et al., 2020) and the database is managed under a version-
188 controlled system and centrally stored in GitHub (<https://github.com/SoilBGC-Datashare/sidb>).

189

190 **3.1 The repository**

191 The structure of the SIDb project contains three main folders: *data*, *docs*, and *Rpkg* which
192 provide access to the database, the website (<https://soilbgc-datashare.github.io/sidb/>), and the R
193 package. The tree structure of the essential repository components is as follows:

194

```
195 SIDb project
196 Readme.md
197 LICENSE.md
198 travis.yml
199 |-- data
200     |-- entry1
201         |-- initConditions.csv
202         |-- metadata.yaml
203         |-- timeSeries.csv
204 |-- docs
205     |-- _config.yml
206     |-- index.html
```

```

207     |-- _layouts
208     |-- _includes
209     |-- assets
210     |-- css
211 |--tests
212     |--testthat
213         |--test_dataStructure.R
214     |--data_test.R
215     |--pkg_test.sh
216 |-- Rpkg
217     |-- DESCRIPTION
218     |-- NAMESPACE
219     |-- R
220     |--data
221     |--inst
222     |-- man
223     |--vignettes
224

```

225 3.2 The database

226 The open-source approach to SIDb allows data access, manipulation, analysis and contribution to
227 be accomplished without proprietary software. The soil incubation data is stored in the *data*
228 folder. Each entry in the database consists of a folder containing three files and has the name
229 convention ‘*AuthornameYEAR*’ (optionally with journal name abbreviation appended) and the
230 suffix ‘a’ or ‘b’ if multiple entries for one author and year exist. 1) The *metadata.yaml* file
231 contains the following required sections: citation and curator information, basic site information
232 (*siteInfo*), experimental set-up of incubation (*incubationInfo*), and the metadata for the variable
233 in the time series data (*variables*). The structure of the metadata file allows for flexible inclusion
234 of many types of experimental and incubation designs. 2) The *initConditions.csv* file includes
235 site, treatment, and initial soil characteristics (C content, texture conditions, etc.; Table 1). 3) The
236 *timeSeries.csv* file contains measurements made over the course of the incubation. Column
237 headers in the *timeSeries.csv* file are required to match the values entered for variable names in
238 the variables section of the *metadata.yaml* file (e.g. V1:name, V2:name, etc.). The *Readme.md*
239 file in the data folder provides a detailed explanation of how to add entries to the data folder.
240 Note that for entries to be ingested in SIDb they must pass certain QA/QC tests (described in
241 detail in section 3.2.4 in the R package).

242

243 3.2.1 The metadata file

244 The metadata file is a simple text file that includes all relevant information about the incubation
245 study. The *yaml* format is both human and machine readable. YAML (YAML Ain't Markup
246 Language) files are text files that utilize indent hierarchy to store information in iterable and
247 query-able format. Thus, data stored under main headings may contain subcategories and arrays
248 of information. In an array, each line is started with a hyphen, followed by a space, then the data.
249 A heading of any level must end with a colon, followed by a new line return. The *metadata.yaml*
250 file contains four sections. The first section consists of bibliographical data about the database
251 entry, including DOI and contact information (Fig. 2). The second section, *siteInfo*, includes
252 geographic data, land cover, vegetation, and soil data (Fig. 3). The third section, *incubationInfo*,
253 provides data on laboratory experimental setup and sample treatment (Fig. 4). The fourth section,
254 *variables*, contains metadata for the individual columns of the *timeseries.csv* file (Fig. 5).
255

```
citationKey: # Unique identifier in the format: LastnameYearJOURNAL
doi: # DOI of the publication where data is published
entryAuthor: # First and last name of the person who enters the data in this file
entryCreationDate: # Date when the data is entered in this file. Format: YYYY-MM-DD
contactName: # First and last name of contact person
contactEmail: # Email of the contact person
entryNote: # Any notes or comments related to this entry.
study: # Overall study description
```

Figure 2: Bibliographic data needed for each database entry

256 One advantage of the *yaml* format is the ease with which specific types of data can be grouped in
257 a hierarchical array. For example, in Figure 3 *site* is a subfield of *siteInfo*, and latitude is a
258 subfield of coordinates. More subfields can be added to the *siteInfo* subfield as necessary,
259 however, adding a secondary subfield beneath existing subfields should be avoided in SIDb as
260 consistent data structure is required for data aggregation. For example, in the *siteInfo* section, the
261 variables *coordinates*, *country*, *MAT*, *MAP*, *landCover*, *vegNotes* and *soilTaxonomy* all need to
262 be equal to the length of the site array Fig. 3.

263 In Fig. 4, the *incubationInfo* field has a subfield with a description on how the
264 incubations were carried out. This is important information for documenting the experimental
265 conditions under which the incubations were conducted. However, specific treatments and
266 experimental conditions (temperature, moisture, etc.) should be explicitly entered under the
267 appropriate corresponding subfields (Fig. 4).
268

```

siteInfo:
  site: # Names of individual sites,
        # if one site, keep on this line, if multiple, use array format
        # These fields should be arrays of equal length to site array
  coordinates:
    latitude: # Latitude in decimal units
              #(check for negative that denotes southern hemisphere)
    longitude: # Longitude in decimal units
              #(check for negative that denotes west)
  country: # Name of country where site is
  MAT: # Mean annual temperature in degrees Celsius
  MAP: # Mean annual precipitation in mm
  elevation: # Elevation of study site in meters above sea level
  landCover: # Land cover of the site. Valid fields are:
              # bare, cultivated, forest, rangeland/grassland,
              # shrubland, urban, wetland, tundra
  vegNote: # Additional details about land cover such as
            # species or functional type composition
  soilTaxonomy:
    soilOrder: # Soil order according to the classification system described below
    soilFamily: # Soil family description (e.g., Eutric of Eutric Cambisol)
    soilSeries: # Soil series according to the classification system described below
    classificationSystem: # Name of classification system used.
                          # Valid fields are: USDA, FAO, and WRB
  permafrost:
    permafrostExist: # Yes or blank if no (if yes, permafrost must exist at the site)
    activeLayer: # Depth of the active layer in meters

```

Figure 3: Site information for each database entry

269 The last fields that must be filled in are in the *variables* section (Fig. 5). This section consists of,
270 in sequential order, subsections containing the metadata that correspond to the respiration time
271 series observations (columns) of the *timeSeries.csv* file. The number of variables (V1-Vn) must
272 therefore correspond to the number of columns in the *timeSeries.csv* file. The first column in the
273 *timeSeries* file must be a vector of time (in days or other consistent unit), and thus the first
274 variable name (V1:name) in the variables section must also be “time”. Experimental and
275 incubation treatments listed in the *incubationInfo* section must be specified under each variable
276 (V2, V3, etc.). Note that if a treatment has only one level it will be reported in the *incubationInfo*
277 section and does not need to be repeated in the *variables* section. For example, if all incubations
278 were conducted at the same temperature, the incubation temperature would be reported under the
279 *temperature* subheading in the *incubationInfo* section and the information will be automatically
280 propagated to all of the variables (example of Crow2019a in the database). However, if a
281 treatment has multiple levels, e.g. an incubation study utilizing three temperatures, the
282 *temperature* subheading under *incubationInfo* would be left blank, and the temperature level
283 would need be specified for each variable in the *variables* section in a subheading called
284 “temperature” (example of Bracho2018SBB in the database).

285

```

incubationInfo:
  incDesc: # Short description of the incubation setup and main treatments
           # These fields should all be one dimensional arrays.
           # Values for experimental variables with multiple treatment levels
           # should be entered in the variables section, and left blank here
  depthInfo: # Soil depth in cm. If only one depth listed instead of range,
             # enter as top and bottom, 0 is organic/mineral interface.
             # If organic layer, enter 0 as top and bottom.
             # If multiple depths, leave blank and specify in variables section
    top:
    bottom:
    midDepth:
  surfaceAtm: # blank if zero is organic/mineral interface,
             # yes if surface is atmospheric interface
  horizon: # soil horizon designation
  temperature: # Temperature at which incubations were performed in Celsius.
              # If temperature is an experimental treatment with multiple levels,
              # leave blank and specify in variables section
  moisture: # Use moisture as a template for any additional treatments performed,
            # i.e. report treatmentName, value, and units (if applicable)
    value: # Overall moisture at which incubations were performed.
          # If moisture is an experimental treatment with multiple levels
          # leave blank and specify in variables section
    units: # Valid fields are: percentGWC, percentFieldCapacity,
          # percentWaterFilledPoreSpace
  anaerobic: # Yes if headspace flushed with N2 or He, blank if aerobic
  gasMeasured: # Blank if CO2, other valid entries are:
              # CH4, N2O, 13CO2, 14CO2, 13CH4, etc.
              # Leave blank if multiple gases measured and specify in variables section
  replicates:
    value: # Number of replicates per treatment
    type: # Valid fields are: field or lab
  incubationTime: # length of incubation in days
  preincubationTime: # Pre-incubation time in days
  samplePreparation:
    intactCore: # yes or no
    sieving: # no, or mesh size in mm
    rootPicking: # yes or no
    rockPicking: # yes or no
  gasAnalyzer: # Gas analysis equipment for measurements

```

Figure 4: Incubation information for each database entry

286 3.2.2 Data entries

287 The *timeSeries.csv* file for each entry in the database contains the time series of incubation data
 288 in comma-separated format. The first column of the data file must contain the times at which gas
 289 measurements were taken. Subsequent columns must contain the respiration measurements. The
 290 format of the data is irrelevant (e.g. units) as long as the relevant information to identify each
 291 respiration column is described in the variables field of the metadata file.

292

```

variables: # These describe the columns of your timeSeries.csv file
  V1: # column 1
    name: # Name of first variable in the accompanying csv data file.
          # First variable should be time
    units: # Units of first variable in accompanying file. Usually "d" for days
  V2: # column 2
    name: # Name of second variable in accompanying file
    varDesc: # Description of the variable
    site: # Site where the incubated sample was collected
    experimentalTreatment: # 'experimentalTreatment' here is a place holder.
                           # Replace this word by any of the listed variables
                           # in incubationInfo above (temperature, moisture, etc.)
                           # and type value or level after colon
    gasMeasured: # Blank if CO2, Other valid fields are:
                 # CH4, N2O, 13CO2, 14CO2, 13CH4, etc
    units: # Units in which this variable is provided if not a factor
    statistic: # Leave blank if mean values.
              # Other valid fields include: SD, SE, and none (if a single rep)
    primaryVariableName: # Links variable with associated timeseries data
                        # collected on the same sample e.g. SD data or 13C-CO2 data
                        # associated with mean CO2 data

```

Figure 5: Information for each variable

293

294 3.2.3 The website

295 Documentation of the project, which includes the database and the R package, is presented on
 296 the project's website (<https://soilbgc-datashare.github.io/sidb/>). The website is publicly served by
 297 *GitHub Pages*. Every time new changes are pushed to the SIDb repository, the website is rebuilt
 298 and served automatically by GitHub.

299

300 3.2.4 The R package

301 Data in SIDb are stored in a format that can be read in any programming language. We provide
 302 an R package to allow users to compile or read the database into R and a platform to facilitate
 303 future analyses. To install the package, open R and run:

304

```

305 install.packages("devtools")
306 devtools::install_github('SoilBGC-Datashare/sidb/Rpkg/', build_vignettes=TRUE)

```

307

308 Once the R package 'sidb' is installed and loaded, a browser-based html version of the available
 309 vignettes can be accessed using:

310

```

311 browseVignettes('sidb')

```

312
313 There are currently two vignettes available: ‘sidbQueryReportPlot’ and ‘Fitting data to models’.
314 The first vignette describes a simple workflow for querying, generating reports, and plotting data
315 with SIDb. The second vignette demonstrates the model fitting functions built into the R package
316 ‘sidb’.

317 In the sidb R package two main functions are provided: *loadEntries.R* and *readEntry.R*. As
318 their names suggest, *loadEntries.R* collects all metadata and data from all entries and produces
319 an ‘R list’ with the entire database. The function *readEntry.R* reads individual entries from the
320 database and also produces an ‘R list’. The package also provides a function that “flattens” and
321 coerces the database list object into a simpler data structure for easier querying (*flatterSIDb.R*),
322 as well as stand-alone functions to query the entire database in its native list format for specific
323 variables. For instance, the function *coordinates, R* extracts all latitudes and longitudes for each
324 entry in the database. Similarly, other functions are provided to extract C and nitrogen (N)
325 content, or the incubation duration of each entry.

326 Quality control is provided for code testing and data validation. A brief overview is given
327 here and more details can be found in the *Readme.md* file located in the directory ‘sidb/tests’
328 within the SIDb GitHub repository. Code testing can be done both locally and remotely. For
329 local testing we have written a shell script that runs R CMD check on the package directory
330 (github: sidb/tests/pkg_test.sh). For remote testing, we use Travis Continuous Integration to run
331 R CMD check on the Rpkg directory of the SIDb GitHub repository. This ensures that any
332 modifications to the functions or other aspects of the ‘sidb’ R package are tested every time a
333 new commit is made in the repository, and that we will be notified of any errors, warnings, or
334 issues.

335 For data validation, raw SIDb data (entry files that live outside the R package in the
336 ‘data’ directory) can be tested for conformity to SIDb standards using the file ‘data_test.R’
337 (github: sidb/tests/data_test.R). This R script runs all tests in the subdirectory ‘testthat’. Tests can
338 be run from the command line or directly inside R using the R package *devtools*. Contributors of
339 new data or code must run these tests before contributing to SIDb and no pull requests will be
340 accepted if any of the tests fail.

341

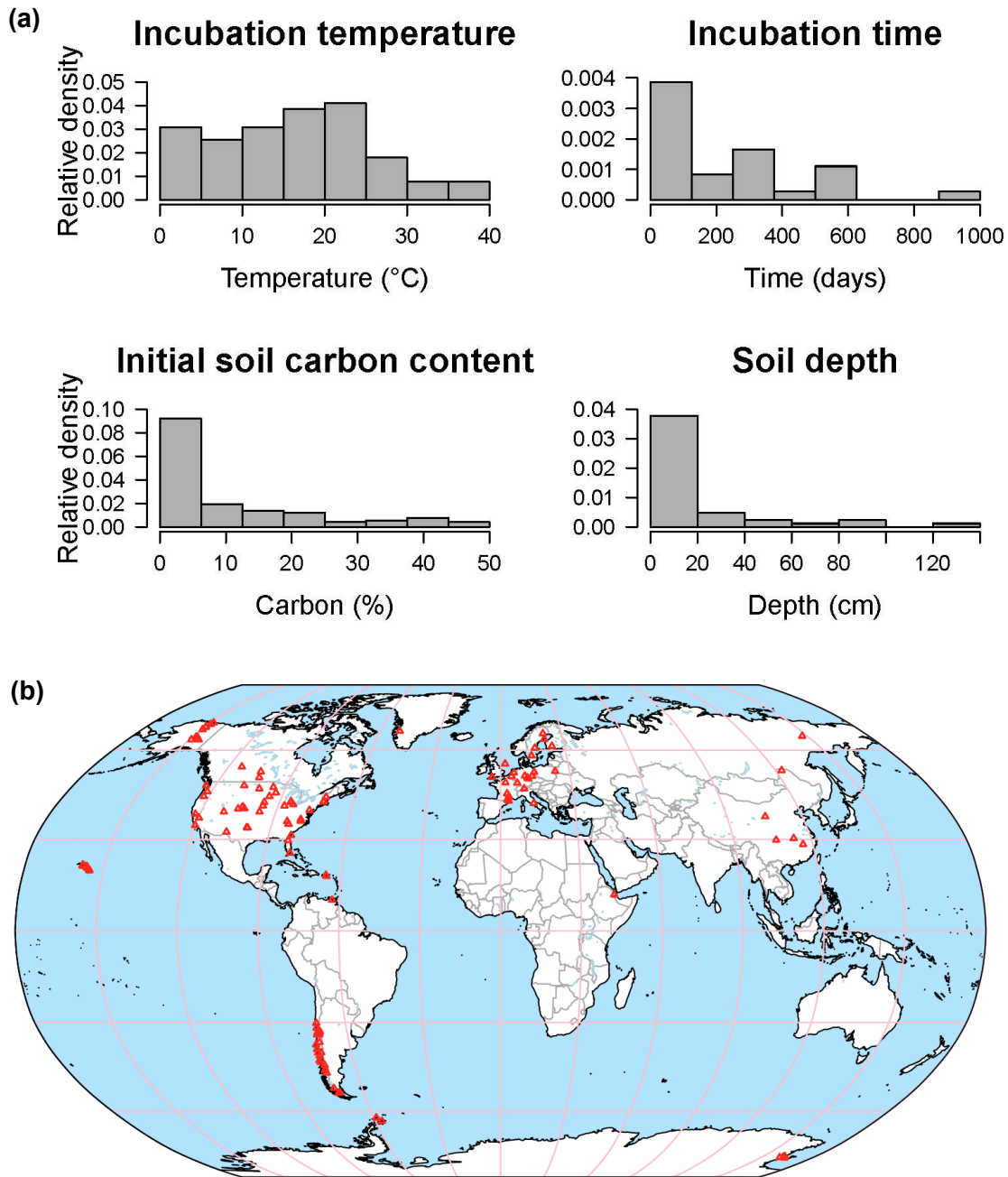


Fig. 6 Data distribution histograms of incubation temperature, time, initial soil C content, and soil depth for available incubation data in SIDb 1.0 (a). Map of currently available incubation studies (b).

342 **3.3 Summary statistics in SIDb version 1.0**

343 The database is a work in progress: currently SIDb includes 31 studies with 684 time series,
 344 representing a total number of 42,545 datapoints (Fig: 6). Most entries contain multiple time
 345 series of CO₂ fluxes. Incubations reported in SIDb were performed under temperatures ranging

346 from 0 to 40 °C with the majority of incubations under normal laboratory temperature (20-25 °C)
347 (Fig. 6a). Soil temperature is the most frequently reported laboratory treatment, while soil
348 moisture is less frequently reported despite the fact that it is also a key factor in incubation
349 studies. The omission of soil moisture data may be related to inconsistencies in reporting
350 conventions, a topic that is discussed further in section 4.3. All soils listed in our database
351 included surface soil samples, however some studies considered soil depth as a treatment and
352 report incubation data from soil layers as deep as 1.2 m (Fig. 6a).

353 Important geographic and ecological gaps exist in SIDb version 1.0. Coverage is highest in
354 temperate, followed by arctic regions, with only a few studies in tropical areas while the
355 continents of Africa and Australia are barely represented (Fig. 6b). Incubation data from the
356 tropics are currently poorly represented in SIDb despite their vulnerability and the importance of
357 tropical regions to global C cycling, and therefore should be a priority for both future ingestion
358 into SIDb and further study. For most ecosystems, there are still many incubation studies to be
359 included into SIDb in the future. Additionally, recent work (Fontaine et al., 2007; Hicks Pries et
360 al., 2018; Mathieu et al., 2015) has highlighted the importance of understanding deep soil
361 processes and potential changes due to global warming. In fact, warming effects on respiration
362 have been observed at depths as great as 1m (Hicks Pries et al., 2017). Incubations of deep soils
363 thus represent a major gap in SIDb, which is reflective of the lack of deep soil incubation studies
364 more broadly, and present a large potential for future study. It was not our intention with SIDb to
365 produce a comprehensive database. Instead, we want to introduce SIDb's structure, tools, and the
366 current capacity of the database to the broader scientific community, with the potential to expand.
367

368 **4 Required and suggested data reporting for inclusion into SIDb**

369 While consistent methods across studies facilitate meta-analysis, incubation studies must remain
370 adaptable to each research question, available resources, and soil properties. Nonetheless, in
371 developing SIDb and the entry template, the most critical required components of incubations for
372 making comparisons across studies emerged. On the basis of these observations, we have
373 generated a list of variables, including information about the sites, soils, and the set-up of the
374 incubation itself, that we require in order for a study to be ingested in SIDb (Table 1). Here, we
375 discuss the issues associated with these critical variables and make suggestions for other useful
376 variables to report that, while not required, will increase the interpretability of results and allow

377 for broader inclusion into syntheses and meta-analyses (Table 1). In the supplemental material,
378 we also offer a limited discussion of methodologies and measurements such as incubation setup,
379 sample preparation, additional variables to measure, and special considerations for radiocarbon
380 incubations.

381

382 **4.1 Site information**

383 Site characteristics provide a context for the inherent conditions of the soils. General site
384 characteristics, such as latitude and longitude, mean annual temperature and mean annual
385 precipitation are important in drawing out the similarities or differences between studies.
386 Descriptions of the ecosystem and the aboveground vegetation give information on litter input
387 and chemistry, which can be a direct link to organic matter quality. Additionally, providing
388 information on the soil order and taxonomy helps to put findings into context with other studies
389 (Schimel and Chadwick, 2013).

390

391 **4.2 Soil characteristics**

392 There are ultimately two essential soil variables that must be reported for incubation studies, and
393 a myriad of suggested variables that facilitate comparisons among and explorations of potential
394 drivers. The first essential soil variable is depth, which is a major organizing factor of many soil
395 characteristics. No matter whether an individual incubation study measured soil from a single
396 depth increment or multiple depth increments, either the depth increment (top, bottom, and
397 middle) or the horizon must be reported. Ideally, both depth and horizon should be reported as
398 samples can be taken from a generic depth or from a mixture of horizons (when sampled to a
399 certain depth). All subsequent soil characteristics should then be reported for each depth
400 increment or horizon incubated and provided in the *initConditions.csv file*.

401 When reporting the sampling depth, it is necessary to report whether depth is in relation
402 to the soil surface, which can be defined as the top of the mineral soil or the top of the organic
403 horizon depending on the system, or within a specific soil horizon. Additionally, specifics of the
404 geography and topography of the sampling locations, such as permafrost zone, active layer
405 thickness, or water table depth in permafrost and peatlands are crucial to report.

406 The second required soil variable is either the initial C (reported in mg C gdw⁻¹ or %) or
407 organic matter (which can be converted to C), which is essential for facilitating comparisons

408 across studies and for normalizing rates of C losses during incubations. Other common and
409 useful variables to measure are initial N (reported in mg C or N gdw⁻¹ or %), bulk density in g
410 cm⁻³, soil texture, and pH.

411 Most soil characteristics, as listed in Table 1, can be measured at the beginning of an
412 incubation on a subsample of the soil being incubated, while others like pH, redox, or microbial
413 biomass may be best measured multiple times during the course of an incubation (see
414 Supplement for more details). For anaerobic incubations, we strongly recommend measuring
415 redox potential because it may not be sufficient to assume that anoxic conditions (e.g. soils
416 inundated with water and headspace filled with N₂ or He) will result in the production of CH₄
417 during the incubation as there can be a considerable lag period before CH₄ production occurs
418 (Knoblauch et al., 2018; Treat et al., 2015).

419

420 **4.3 Incubation information**

421 Details of incubation studies should be reported as they enhance the value of a primary study, but
422 also, critically, they determine whether or not they can be included in a synthesis or meta-
423 analysis. Thus, most of the information about how an incubation and its treatments are carried
424 out are required variables in SIDb. Incubation duration, temperature, and soil moisture are
425 among the most important details to provide because they directly affect microbial activity and
426 therefore C flux rates (Table 1). For temperature and soil moisture, it needs to be clarified
427 whether temperature and moisture were controlled at a single value or whether there were
428 multiple temperature or moisture treatment levels. For temperature, details on how incubation
429 temperature was achieved should be provided (e.g. water bath, freezer, or controlled environment
430 chamber). For moisture, it should be specified whether the soils were all brought to the same
431 moisture content or left at field conditions. For below-freezing incubation temperatures, unfrozen
432 soil water can also be quantified, if possible, as temperature responses of CO₂ production at
433 subzero temperatures are influenced by water availability (Öquist et al., 2009). Moisture
434 treatments range from fully aerobic (either drier than or at field capacity) to fully anoxic
435 (headspace of jar flushed with N₂ or helium) to fluctuating moisture conditions. In aerobic
436 incubations, soils are often freely drained and deionized water is added over the course of the
437 incubation to maintain constant moisture content. However, caution should be paid in order to
438 maintain constant moisture through the incubation and not allow soils to dry out as drying and

439 rewetting of soils can affect C mineralization rates and microbial activity (Birch, 1958; Rey et
440 al., 2005; Unger et al., 2010). In addition, adjustments to soil moisture are ideally made at least
441 24-48h prior to making measurements to minimize confounding effects of water addition (Rey et
442 al., 2005). For anaerobic incubations it may not be necessary to add water during the course of
443 the incubation as incubation vessels typically remain closed, but caution should be taken if water
444 is added as it often contains dissolved oxygen. Other critical parameters to report about the
445 incubation from the synthesis perspective include whether replicates are field (i.e., spatially
446 different soil cores) or analytical replicates, whether soil samples were homogenized (e.g. by soil
447 sieving), or whether roots were removed prior to incubation (see Supplement for more
448 information). Lastly, the duration of a pre-incubation should be reported if carried out.

449

450 **4.3.1 Flux measurements**

451 Incubation data are most commonly published as C flux rates or cumulative C release over time
452 for the whole incubation period. SIDb is designed around incubation studies that report
453 respiration rates and cumulative release over time (*timeSeries.csv*), and time series data is
454 required for inclusion in SIDb. Reporting only one average flux value, one maximum production
455 value, or one single cumulative C release value for the whole incubation period may be useful
456 for comparison of treatments within a study, but omits key information about changes in C
457 dynamics over time and precludes our ability to model dynamics of different C pools. If changes
458 in C dynamics over time are not of interest for a specific study, time series data should be
459 provided in supplementary material or in a data repository such as SIDb. Flux rates can be
460 provided on a per gram dry soil or per gram soil C basis, as $\text{mg CO}_2\text{-C g dry weight}^{-1} \text{ d}^{-1}$ or mg
461 $\text{CO}_2\text{-C g}^{-1} \text{ soil C day}^{-1}$. These units can be easily converted to one another using the required
462 initial C data (Table 1). Providing flux rates on a wet-weight soil basis or per volume of soil
463 slurry is discouraged, as SIDb does not support this format and it precludes comparisons to other
464 studies. If units of dry weight are not available, then soil moisture content and bulk density need
465 to be reported so that data can be converted to standard units. Reporting C release on a per gram
466 C basis captures information about C decomposability and reveals information about the relative
467 C release from a given soil that is independent of its C quantity; this is particularly useful for
468 comparisons among soils, sites and incubation studies (Schädel et al., 2014).

469

470 **5 Case study: Fitting time series data to pool models in SIDb version 1.0**

471 Our incubation database can be easily integrated with other R packages for further analyses. For
472 instance, it is possible to integrate soil C pool modeling from the SoilR package (Sierra et al.,
473 2012) with parameter optimization from the FME package (Soetaert and Petzoldt, 2010). We
474 illustrate this functionality with a simple example. The entry Crow2019a in the database contains

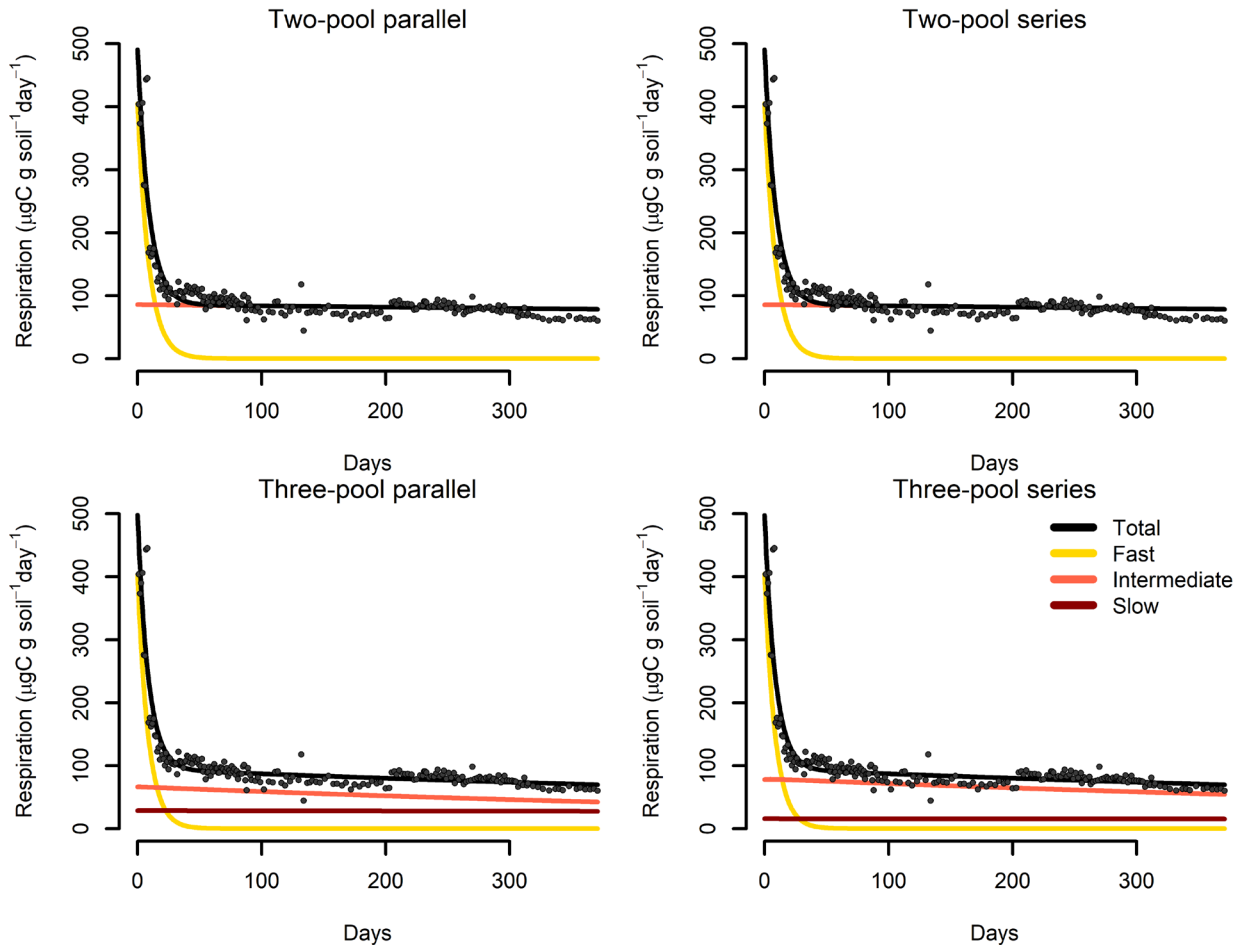


Figure 7: Results from a parameter optimization procedure to soil incubation data from a native tropical forest of Hawaii. The parallel model structures do not consider transfers of C among pools, while the series model structures transfer C sequentially from fast to slow cycling pools. In all cases, the models fitted the data relatively well (Table 2), and identified the relative contribution of the different pools to the overall respiration flux.

475 a large number of long-term incubations (371 days). From those incubations, we selected data
476 from a native forest in Hawaii and fitted a set of first order models with two or three pools.
477 Following the procedure described in Sierra et al. (2015), we optimized two- and three-pool

478 models with parallel, series, and feedback connections among them (Fig. 7). Depending on the
479 question asked different criteria can be considered to select the best model (e.g., Akaike
480 information criterion or number of parameters, Table 2) and it is beyond this manuscript to
481 identify the best model, we simply show the basics of an example using SIDb.

482

483 **6 SIDb connections to other databases**

484 There are two approaches to database building, which can be characterized by tradeoffs between
485 the scope and quantity of data, the ease of data analysis, and the simplicity of data entry. SIDb
486 has a narrow scope (i.e. incubation time series), allowing for the flexibility to incorporate studies
487 with different variable types and experimental designs, while the data itself is highly structured
488 in order to facilitate data analysis. Other soil databases, such as the International Radiocarbon
489 Database (ISRaD, Lawrence et al., 2020) or the International Soil Carbon Network (ISCN,
490 <https://iscn.fluxdata.org/>) have the advantage of a much larger quantity of data and a much
491 broader scope. However, maintenance and data ingestion with these larger databases becomes
492 much more challenging and requires either, a) relaxing control of data structure, units of
493 variables, and direct data oversight, such as the case with the International Soil Carbon Network,
494 or b) in the case of the International Radiocarbon Database, increasing the complexity of the data
495 structure while enforcing strict variable control, e.g. allowable names, factor levels for
496 categorical data, and numerical limits for quantitative data. Owing to the broader scope,
497 maintaining these larger databases inevitably requires additional time and effort.

498 However a database is structured, establishing a common set of required measurements,
499 metadata, and site-level data provides transparency that helps both to identify and to reduce
500 systematic bias. The statistical power provided by the wealth of data points in a database such as
501 SIDb is only useful as long as any potential systematic bias is identified. For example, all studies
502 in SIDb report data at the variable level with respect to a time variable, as well as provide
503 information about the experimental design, where the samples were collected from, who
504 performed the study and how to access the original data. Additionally, providing data such as
505 geographic coordinates, land cover, MAT, MAP, soil taxonomy, and soil C content enables
506 leveraging of databases that may have a different scope but contain potentially useful supporting
507 data. For example, respiration time series data from SIDb could be compared to ¹⁴C content of
508 bulk soil or respired ¹⁴CO₂ from ISRaD (Lawrence et al., 2020) by stratifying both databases

509 along common variables, or a query could be made using geographic coordinates, DOI, or other
510 variables.

511

512 **7 Data availability and user guidelines**

513 Version 1.0 of SIDb is publicly available at [DOI: 10.5281/zenodo.3871263](https://doi.org/10.5281/zenodo.3871263) (Sierra et al., 2020)

514 Documentation of the project and the R package are presented on the project's website

515 (<https://soilbgc-datashare.github.io/sidb/>).

516 The database is open for reuse and the usage license follows the MIT license

517 (<https://opensource.org/licenses/MIT>). When using the database or R package, users should cite

518 this definition publication and consider citing individual studies (publication or dataset).

519

520 **8 Conclusion**

521 Currently, SIDb is a compilation of a wide range of incubation studies with built in capacities to
522 summarize the database and conduct model comparisons for fitting curves to time series data.

523 There is great potential benefit for the soil C community through identification and ingestion of
524 new datasets into SIDb. Every incubation study is planned and performed to answer a specific
525 question; however, when analyzed in aggregate, syntheses of incubation studies can help answer
526 fundamental questions about soil C pools, their stability, and vulnerability to global change.

527 Furthermore, setting up incubation studies involves several decision points, such as whether to
528 sieve or preincubate the soil, whose consequences have not yet been tested systematically, but
529 which may be able to be tested using SIDb.

530 A comprehensive collection of existing laboratory incubation data will be invaluable for
531 the synthesis of spatial, methodological, and functional trends, as well as for identifying key gaps
532 in our current knowledge. Individual researchers are encouraged to add individual study results
533 to the database thereby helping fill gaps in our broader understanding of soil C cycling in the
534 process. A key goal for the next stages of development in SIDb will focus on expanding the
535 geographical and ecological coverage of the entries.

536 SIDb is specifically designed to host incubation data with time series of respiration rates
537 to facilitate synthesis studies. We encourage researchers to archive their data in the format
538 presented here, but caution that this database is not a long-term archive. SIDb not only collects
539 data in a structured format, it also provides tools for data analysis and reporting through an R

540 package and a website. Soil incubations are a commonly used technique for answering many
541 different kinds of research questions, and here we provide recommendations on best practices, as
542 well as a common data infrastructure for reporting. We expect the size of this database to grow in
543 the future as it can be used as a standard repository for time series soil incubation data following
544 open-source standards.

545

546 **Author contribution**

547 C.A.S. designed the database; C.A.S., C.S., J.B.M., M.A.R., S.E.C., A.P., C.H.P., S.S., and
548 A.M.H. built and populated the database while J.B.M. provided technical database support. C.S.,
549 J.E., C.T. developed the first version of incubation recommendations and C.S. wrote up the
550 initial draft of the SIDb manuscript. All authors contributed to the writing.

551

552 **Competing interests.** The authors declare that they have no conflict of interest.

553

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732 **Table 1** Required (R) and suggested (S) variables to report and measure prior to or during time
 733 series soil incubations.

Variable	Unit	Time of measurement	Required/ Suggested	Notes
Site information				
Latitude/Longitude	(decimal) degrees	A ¹	R	
Mean annual temperature	°C year ⁻¹	A	R	
Mean annual precipitation	mm year ⁻¹	A	R	
Ecosystem/vegetation		A	R	descriptive
Soil taxonomy		A	R	USDA, FAO, WRB
Soil characteristics				
Horizon		A	S	Either horizon or depth in cm is required
Soil Depth		A	R	Include top, mid, and bottom of each increment incubated
Initial C	mg C gdw ⁻¹ or %	A	R	Initial C preferred, but organic matter allowed
Soil organic matter	mg C gdw ⁻¹ or %	A	R	Required if initial C not reported
Initial N	mg C gdw ⁻¹ or %	A	S	
Bulk density	g cm ⁻³	A	S	
pH		A, B ²	S	
Soil redox potential (Eh)	mV	A, B	S	One measurement (end) or continuous. Most critical for anaerobic soils
Horizon texture	% clay, silt, sand	A,	S	
Horizon soil porosity	% (m ³ m ⁻³ x 100)	A	S	
Microbial biomass	mg C gdw ⁻¹	A, B	S	Or as mg N gdw ⁻¹
δ ¹³ C	‰	A, B	S	Carbon isotope composition
Incubation information				
Incubation duration	days	A	R	
Incubation temperature	°C	A, B	R	Report multiple times if not consistent
Incubation moisture	%	A, B	R	Gravimetric water content, field capacity
Temperature control method		A	S	Descriptive; e.g. room temperature, water bath, environmental chamber

Variable	Unit	Time of measurement	Required/Suggested	Notes
Moisture control method		A	S	Descriptive; e.g. field conditions, added water to get to a target water content, how often checked moisture content, etc
Aerobic/Anaerobic		A	R	Anaerobic if headspace flushed with N ₂ or He
Treatments		A	R	Descriptive; if quantitative include units
Replicates		A	R	Field or analytical replicates
Sample preparation		A	R	e.g. intact core, sieving, homogenization, roots removed
Pre-incubation duration	days	A	S	
Flux time series	mg CO ₂ -C gdw ⁻¹ day ⁻¹	A, B	R	
Gas analysis		A	R	Description of equipment used

734 ¹A: report once

735 ²B: can be reported multiple times during incubation

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Table 2 Summary statistics from the parameter optimization procedure using the database entry Crow2019a, a 371 day long incubation with soil from native forest in Hawaii.

Model structure	Number of optimized parameters	Sum of squared residuals	Mean of squared residuals	AIC ¹
Two-pool parallel	3	113685.2	554.5	-6.64
Two-pool series	4	113685.2	554.6	-4.64
Two-pool feedback	5	113685.2	554.6	-2.64
Three-pool parallel	5	109584.4	534.6	-2.56
Three-pool series	7	109583.4	534.6	1.44

¹ Akaike information criterion