

APPLICATION

flexible: An R package to process ecosystem gas fluxes from closed-loop chambers in an automated and reproducible way

Joseph Gaudard¹  | Richard J. Telford¹  | Julia Chacon-Labela^{2,3}  |
 Hilary Rose Dawson^{4,5}  | Brian J. Enquist⁶  | Joachim P. Töpper⁷  | Jonas Trepel⁸  |
 Vigdis Vandvik¹  | Marta Baumané^{9,10}  | Kristine Birkeli¹  |
 Mukhlis Jamal Musa Holle¹¹  | Jason R. Hupp¹²  | Paul Efren Santos-Andrade¹³  |
 Tin Widayani Satriawan¹⁴  | Aud H. Halbritter¹ 

¹Department of Biological Sciences and Bjerknes Centre for Climate Research, University of Bergen, Bergen, Norway; ²Departamento de Biología, Universidad Autónoma de Madrid, Madrid, Spain; ³Centro de Investigación en Biodiversidad y Cambio Global, CIBC-UAM, Universidad Autónoma de Madrid, Madrid, Spain; ⁴Department of Biology, University of Oregon, Eugene, Oregon, USA; ⁵Research School of Biology, Australian National University, Canberra, Australian Capital Territory, Australia; ⁶Ecology and Evolutionary Biology, University of Arizona, Tucson, Arizona, USA; ⁷Norwegian Institute for Nature Research, Bergen, Norway; ⁸Department of Biology, Center for Ecological Dynamics in a Novel Biosphere (ECONOVO), Aarhus University, Aarhus C, Denmark; ⁹Department of Biology, Faculty of Science, University of Copenhagen, Copenhagen, Denmark; ¹⁰Department of Biology, Faculty of Natural Sciences, Aarhus University, Aarhus C, Denmark; ¹¹Faculty of Biology, Gadjah Mada University, Sleman, Indonesia; ¹²LI-COR Biosciences, Lincoln, Nebraska, USA; ¹³Universidad Nacional de San Antonio Abad del Cusco, Cusco, Peru and ¹⁴Department of Geography, National University of Singapore, Singapore

Correspondence

Joseph Gaudard

Email: joseph.gaudard@uib.no

Funding information

Norges Forskningsråd, Grant/Award Number: 274712, 274831, 287784 and 287801; Norwegian Centre for International Cooperation in Education (SIU), Grant/Award Number: HNP-2015/10037

Handling Editor: Ashish Malik

Abstract

1. Measuring ecosystem gas fluxes is crucial to understanding the dynamics of ecosystem water and energy cycling. A common method to measure ecosystem gas fluxes (CO₂, CH₄, N₂O) and to compare experimental treatments is the use of closed-loop chambers.
2. However, the output data require extensive processing before analysis, making the workflow prone to bias and limiting comparability between studies. While there are various methods to process flux data, they lack reproducibility and automation.
3. The `flexible` R package provides a reproducible way to process raw gas concentration data from closed-loop chambers into an analysis-ready dataset of ecosystem gas fluxes. The processing steps include (1) separating the measurements, (2) fitting a linear, quadratic, or exponential model, (3) assessing the quality of the fit, (4) plotting the fluxes for visual inspection and (5) calculating the fluxes.
4. In addition to reproducibility, `flexible` focuses on homogeneity and automation in data processing, thus reducing the time investment for users while improving comparability across studies.

This is an open access article under the terms of the [Creative Commons Attribution](https://creativecommons.org/licenses/by/4.0/) License, which permits use, distribution and reproduction in any medium, provided the original work is properly cited.

© 2025 The Author(s). *Methods in Ecology and Evolution* published by John Wiley & Sons Ltd on behalf of British Ecological Society.

KEYWORDS

ecosystem gas fluxes, flux chambers, greenhouse gas emissions, R package, reproducible workflow

1 | INTRODUCTION

Measuring ecosystem gas fluxes (hereafter fluxes) quantifies how resources are allocated in ecosystems and provides crucial information to predict future global warming (Bastviken et al., 2022). At the plot scale, fluxes are commonly measured using closed-loop chamber systems (Figure 1; hereafter chambers), allowing comparison across environmental gradients and experimental treatments (Arnone & Obrist, 2003; Maes et al., 2024). The flux calculation (i.e. the amount of gas per time per area) is based on the rate of change of gas concentration inside the chamber. This rate is the derivative of a function $C(t)$ describing the gas concentration over time. However, various methods are used to approximate $C(t)$, and the choice of method can introduce biases and—if not explicitly described—increase the

likelihood of misinterpretations between processed fluxes (Forbrich et al., 2010; Johannesson et al., 2024; Pihlatie et al., 2013).

The way forward is to provide fully reproducible and transparent tools to calculate ecosystem gas fluxes. Such tools and workflows also need to be flexible enough to be compatible across various field setups and gases. This is what the `fluxible` R package, presented in this paper, aims for.

The process of obtaining an analysis-ready flux dataset entails several crucial steps. First, the raw field measurements are trimmed of irrelevant sections in the case of continuous logging or gathered if they were recorded separately, and attributed to metadata. Second, an appropriate model is fit to the data, and fit quality is assessed. Third, the fluxes are calculated from the model output. This step must accommodate different field setups

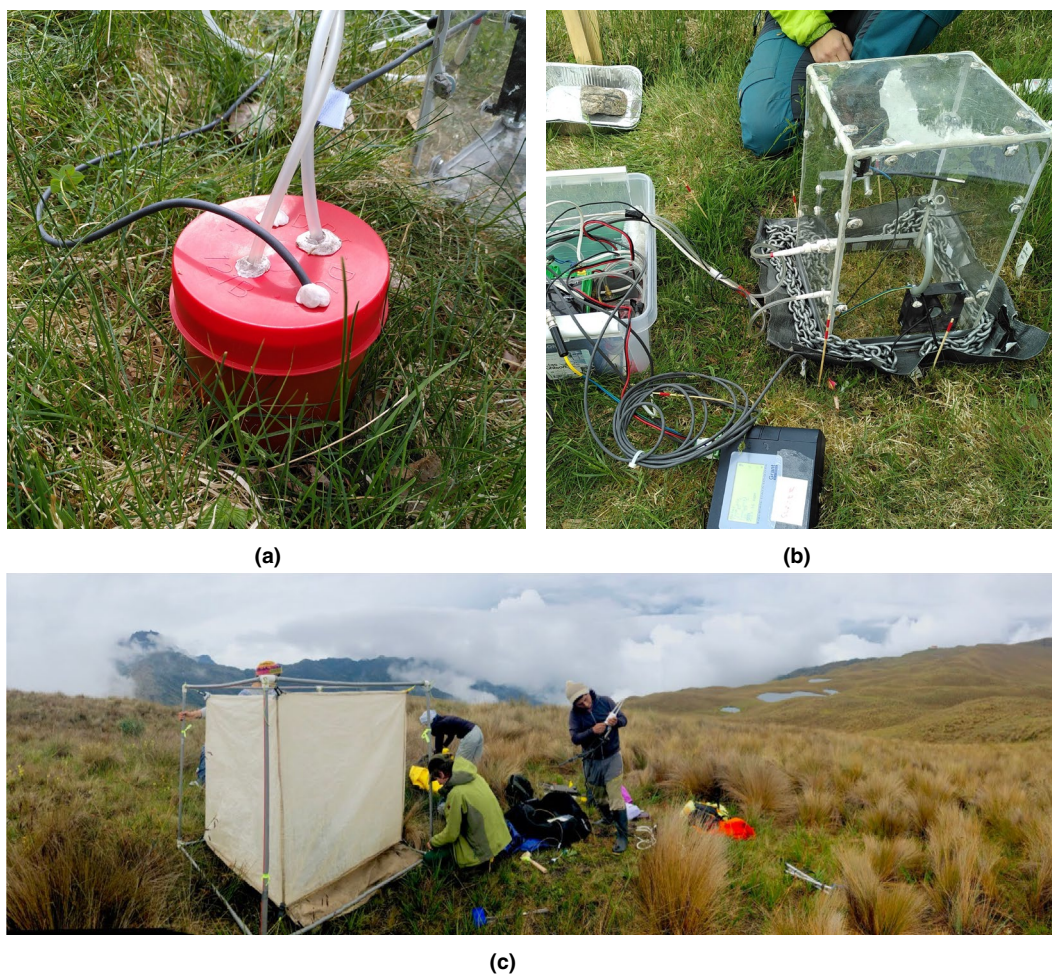


FIGURE 1 Field-based ecosystem gas flux measurements. (a) soil respiration is measured using a PVC pipe chamber. (b) closed-loop plexiglass chamber to measure ecosystem carbon and water fluxes. The chamber has sensors to measure different environmental data, which should be processed simultaneously with the fluxes. (c) Flux tent used for measurements in ecosystems with taller vegetation.

(Figure 1), such as adjusting the volume input for a fixed or variable chamber volume. Indeed, the chamber volume often varies due to microtopography, and this variation should be accounted for in flux calculations.

There are two main pathways to determine $C(t)$. The theoretical approach taken by Kutzbach et al. (2007) includes the relevant flux processes (e.g. photosynthesis, plant respiration, soil respiration) in one equation. An alternative approach is to empirically find a mathematical expression that adequately describes the shape of the fluxes. The latter approach has the advantage of being easier to implement and optimise when fitting a model to real data. The most common is the Hutchinson–Mosier model (HM; Hutchinson & Mosier, 1981). As the HM model cannot fit linear data, implementing it relies on the user deciding if a flux should be fitted with a linear or the HM model (HMR R package; Pedersen, 2024). An improvement was the kappamax method (Hüppi et al., 2018), which automates this step, making the workflow reproducible. As different models tend to under- or overestimate fluxes (Forbrich et al., 2010; Johannesson et al., 2024; Pihlatie et al., 2013), using an exponential and a linear model within the same dataset hinders the comparability of the data. Removing the need to choose between the linear or the HM model, Zhao et al. (2018) introduced a model that can fit both linear and exponential fluxes and takes into account the potential uncertainty around the real start of the measurement.

Here, we introduce the *fluxible* package, using R language (R Core Team, 2024). The package takes raw gas concentration data to an analysis-ready clean dataset of fluxes in five functions (Figure 2). The core principles of *fluxible* are to (1) treat the entire dataset homogeneously, (2) allow full model choice and quality thresholds flexibility and (3) be transparent and reproducible.

2 | IMPLEMENTATION

fluxible provides a user-friendly and fully transparent workflow to calculate fluxes measured with chambers (Figure 2), in which the user decides a priori on model type and quality control. The R package can be installed from CRAN using `install.packages('fluxible')` and an example workflow is available in Supporting Information: Appendix A.

2.1 | Input and output

fluxible requires two inputs: a dataset of the gas concentration over time and a separate metadata file recording the start time of each measurement (Figure 2). The gas concentration dataset is produced by a logger or a gas analyser and contains at minimum a datetime column and a gas concentration column for all the measurements (i.e. a time series of gas concentration that will result in a single flux). The second file must contain a column with the start of each measurement (in datetime format) and identifiers (e.g. the plot IDs) for each measurement. Both inputs can contain other columns,

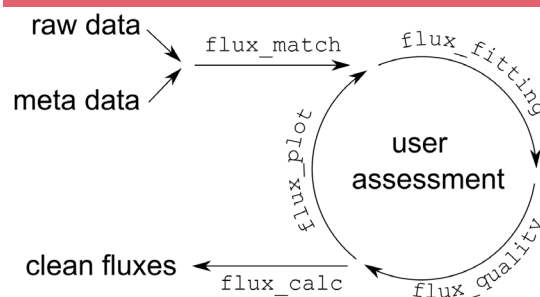


FIGURE 2 The *fluxible* R package processes raw field-measured gas concentration data into analysis-ready fluxes in five steps, including data quality assessment. All functions have the `flux_` prefix. A list of all functions is available in Supporting Information: Appendix F.

which will be kept. Supporting Information: Appendix B provides guidelines for importing the data, both for the continuous logging and 'one file per flux' approaches. As the data processing and quality control are done for the entire dataset, we recommend carefully considering what constitutes one dataset (e.g. splitting different campaigns into separate datasets) before starting the *fluxible* workflow. When measuring different gases simultaneously, we also recommend making a separate dataset for each gas, as the quality checks require different thresholds (see Supporting Information: Appendix C for a multi-gas example).

The final output of the *fluxible* workflow is a dataset containing: clean fluxes (Figure 2), timestamps of the start of the measurements, measurement identifiers provided in the measurements record and any other variables present in either input.

2.2 | Attributing meta-data

The `flux_match` function matches each flux, that is, a series of gas concentration data points over time, with their measurement meta-data based on the time stamp (Figure 2). The `flux_match` function also attributes each flux a unique flux ID (related to the chronological order of the measurements), which is used throughout the entire workflow.

2.3 | Model fitting

The model $C(t)$ is fitted to the data using the `flux_fitting` function (Figure 2), which provides a choice of models (Table 1) and allows the user to prune the start and/or the end of the measurements. The slope (`f_slope` in *fluxible*) used for the flux calculation is the derivative of $C(t)$ at t_0 (`t_zero` in *fluxible*), $C'(t_0)$. The derivative $C'(t)$ is a function of time in the quadratic and exponential models, thus requesting the user to set t_0 (except `exp_zhao18` which includes t_0 as a fitting parameter). We recommend the use of the models `exp_zhao18` (uncertainties around t_0) or `exp_tz` (fixed t_0) because they are more robust (shown in Supporting Information:

TABLE 1 Models available in the `flux_fitting` function and their slope at t_0 . The slope parameter t_0 must be specified by the user in all models, except for the `linear` and `exp_zhao18` models. C_m is the maximum mixing ratio at equilibrium, C_0 is the initial concentration and a and b are fitting parameters. A more detailed explanation is available in [Supporting Information: Appendix D](#).

fit_type	Model C(t)	Slope at t_0 , $C'(t_0)$
linear	$a \times t + b$	a
quadratic	$a \times t^2 + b \times t + c$	$2 \times a \times t_0 + b$
exp_zhao18	$C_m + a \times (t - t_0) + (C_0 - C_m) \times \exp(-b \times (t - t_0))$	$a + b \times (C_m - C_0)$
exp_tz	$C_m + a \times t + (C_0 - C_m) \times \exp(-b \times t)$	$a + b \times (C_m - C_0) \times \exp(-b \times t_0)$
exp_hm	$C_m + (C_0 - C_m) \times \exp(-b \times t)$	$b \times (C_m - C_0) \times \exp(-b \times t_0)$

Quality flag	Meaning	Recommended slope
ok	The model fit is good	As estimated by <code>flux_fitting</code>
discard	A change in gas concentration over time is detected but the fit is poor	NA
zero	No change of gas concentration over time is detected (effectively measuring noise only) and the model fit is poor	0
force_discard	User decided to discard a measurement not flagged as discard	NA
force_ok	User decided to keep a measurement not flagged as ok	As estimated by <code>flux_fitting</code>
force_zero	User decided to replace a measurement by 0	0
start_error	Measurement start is outside of realistic range of ambient concentration	NA
no_data	The measurement does not contain data	NA
force_lm	User decided to use the linear slope for this flux	As estimated by <code>flux_fitting</code> (linear model)
no_slope	Model could not be fit to gas concentration data	NA

TABLE 2 Overview of the quality flags. The full decision scheme for each model is described in [Supporting Information: Appendix E](#).

[Appendix D](#)). Other models are provided for backwards compatibility and user flexibility. The automatic fitting of the `exp_zhao18` model (Zhao et al., 2018) is described in [Supporting Information: Appendix D](#). All models allow positive and negative fluxes, and will produce a negative slope if the concentration decreases over time (ecosystem uptake) and a positive slope if the concentration increases over time (ecosystem loss). Internally, the models use relative time since the measurement start, which accounts for missing data and inconsistent gas concentration measurement frequency.

2.4 | Quality checks and visualisation

The `flux_quality` function ([Figure 2](#)) provides fit diagnostics based on the model, checks that the start of the measurement is within a realistic range (ambient concentration $\pm$ setup error) and flags measurements according to the fit diagnostics. The quality parameters—diagnostic thresholds, ambient concentration and instrument error—are all set by the user and applied consistently to

the entire dataset. Measurement quality (i.e. noise) is assessed by judging the model fit (R^2 or root mean square error; RMSE) and the correlation coefficient between concentration change and time. When the fluxes are low (i.e. small change in gas concentration over time), the fit quality might appear poor because of noise. In that case, the measurement should not be discarded, but instead replaced by 0 (low fluxes with a good fit are kept as such). A measurement with a noticeable change in gas concentration over time and a bad fit because of excessive noise or other disturbances (e.g. leakage) should be discarded. A column with quality flags ([Table 2](#)) and one with recommended slopes are added to the dataframe containing the slopes and metadata. The `flux_flag_count` function produces a table summarising all the flags and measurement counts, which helps to assess and report the overall quality of the dataset (also printed as a side effect of `flux_quality`). Further available quality checks include the g-factor (slope/linear slope) detecting disagreements between the chosen model and linear model, the data ratio (number of data points/length of flux in seconds), detecting measurements with missing data and the

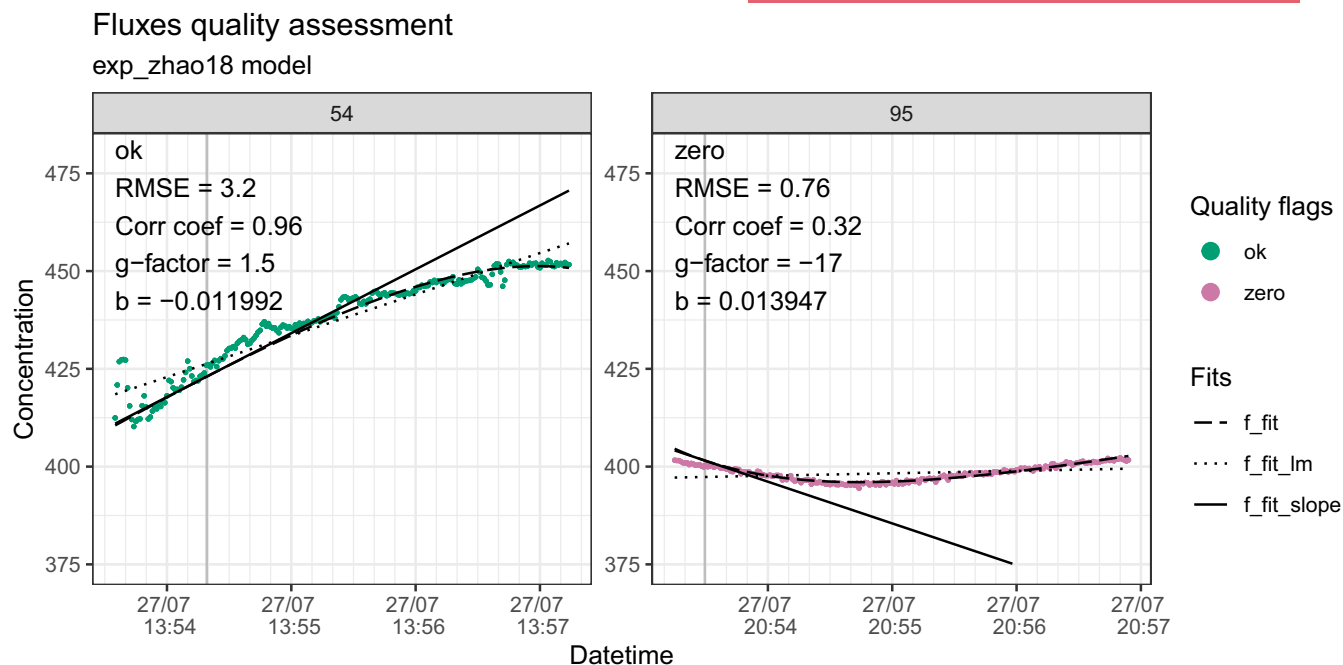


FIGURE 3 Example output of the `flux_plot` function. With quality flags and diagnostics from `flux_quality`, the slope at t_0 (continuous line), the model fit (dashed line), the linear fit (dotted line) and the raw gas concentration (dots). The colours show the quality flags (green for ok, red for discard and purple for zero with default settings) and cuts (same colour as discard). The grey vertical line indicates t_0 (a fitting parameter when using the `exp_zhao18` model, otherwise user defined in `flux_fitting`). The g-factor is calculated as slope/linear slope, and b is the b parameter inside the exponential model. The concentration unit in this example is in ppm, and depends on the data input. The purpose of this visualisation is quality control and to reduce the complexity of these plots the units are not shown. The facet IDs are fluxIDs by default, but can be customised using the `f_facetid` argument.

minimal detectable slope (instrument error range/length of the measurement in seconds).

The `flux_quality` function also offers the possibility to apply the kappamax method (Hüppi et al., 2018) on any of the exponential models. In this method, the slope is replaced by the linear slope when the parameter b in the exponential is greater than the linear slope divided by the instrument error.

The `flux_plot` function (Figure 2) requires the quality diagnostics and thus must be used after `flux_quality`. It produces a plot for each flux with a unique fluxID, including fits, slopes, diagnostics and quality flags provided by `flux_quality` (Figure 3), which can be output as a pdf file. After the visual assessment of the model fits, the user can rerun `flux_fitting` or `flux_quality`, adjusting the fitting window, using a different model or changing the quality thresholds. The user can override the outcome of the quality assessment by providing a vector of fluxIDs to the `force_` arguments in `flux_quality`.

2.5 | Flux calculation

The `flux_calc` function (Figure 2) calculates the flux using Equation (1) where P is the pressure (atm), V the total volume of the setup (L), R the gas constant ($0.082057 \text{ L} \times \text{atm} \times \text{K}^{-1} \times \text{mol}^{-1}$), T the chamber air temperature (K) and A the area of the chamber base (m^2). The slope at t_0 , $C'(t_0)$, can either be the slope estimated

by `flux_fitting` (`f_slope`) or provided by `flux_quality` (`f_slope_corr`).

$$\text{flux} = C'(t_0) \times \frac{P \times V}{R \times T \times A}. \quad (1)$$

The `flux_calc` function can handle mmol mol^{-1} , ppm, ppb or ppt for input of gas concentration and Celsius, Kelvin and Fahrenheit for temperature. The units of the calculated flux is decided by the user and should be in the form $\text{amount surface}^{-1} \text{ time}^{-1}$. Amount can be mol, mmol, μmol , nmol or pmol; surface can be m^2 , dm^2 or cm^2 ; time can be day, hour, minute or second (example in Supporting Information: Appendix A.5). Environmental data that were measured simultaneously as the gas concentration (e.g. soil temperature) can be nested or summarised as mean, sum or median, or kept, assuming that their value is constant within each measurement (e.g. elevation).

3 | VALIDATION

The `fluxible` workflow was validated by comparing a testing dataset of ecosystem gas flux measurements provided by LI-COR (processed by them in SoilFluxPro; SFP; LI-COR Biosciences, Lincoln, NE, USA) with the corresponding raw dataset processed with `fluxible`. A subset of the CH_4 measurements (3925 out of 16,689) required a different fitting window than the rest of the dataset as there was

significant noise at the start of the measurements. Therefore, those data were processed separately before being included in the comparison at the final stage. Strategies to inspect such large datasets are discussed in [Supporting Information: Appendix A](#).

The two methods show a strong correlation ($R > 0.9$) for all gases and instruments ([Figure 4](#)). However, some of the data could not be fitted with the HM model due to the absence of curvature, hence the lower n for the exponential and no exponential fitting for CH_4 ([Figure 4](#) facet a). It was decided here to strictly apply the HM model for comparison purposes, as it is the closest to the SFP exponential workflow, but this strongly supports our recommendation to use the model presented in Zhao et al. (2018) (i.e. `exp_zhao18` or `exp_tz`).

4 | COMPARISON WITH OTHER PACKAGES

We compared `fluxible` with other R packages for ecosystem gas fluxes available on CRAN, based on the criteria of design workflow and availability of functionality ([Table 3](#)). The `fluxible` R package is the only one with a 'one function per step' approach, which is essential to process other environmental data present in the dataset and allows easier outlier investigations. While the packages `HMR` (Pedersen, 2024) and `gasfluxes` (Fuss, 2024) also provide the choice between linear and exponential models, `fluxible` provides different exponential models and a quadratic one in addition to the linear one. The comparison of the quality control

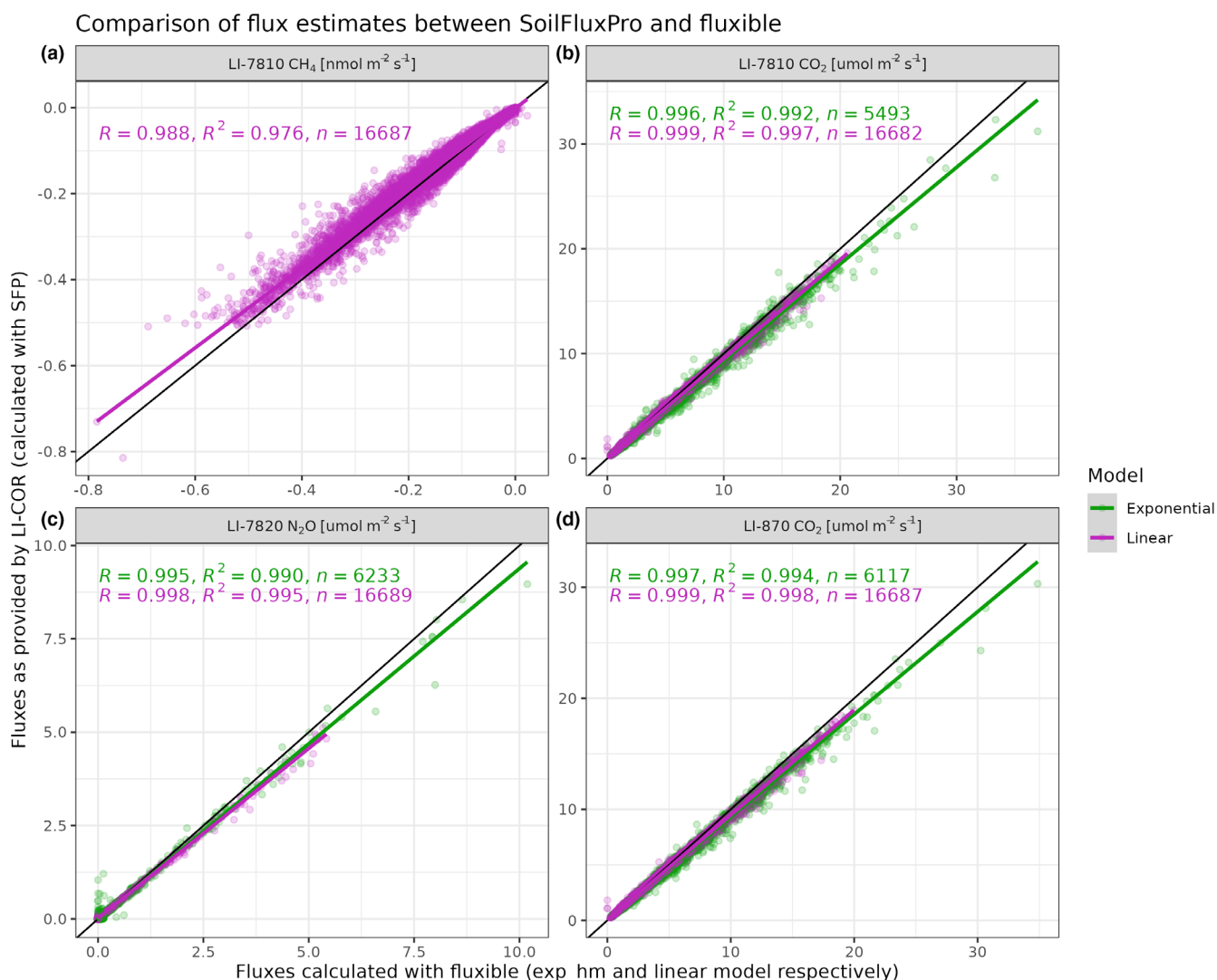


FIGURE 4 Comparison of flux output between SoilFluxPro (SFP; y axis) and the `fluxible` R package (x axis) for CO_2 , CH_4 and N_2O , measured with an LI-8250 soil chamber system using instruments LI-870, LI-7820 and LI-7810, with 8 8200-104 chambers. Continuous black lines show the 1:1 line. Both linear (purple) and exponential (green) models were used. In `fluxible`, the `exp_hm` model was used for the exponential fit as it is the closest to the SFP exponential workflow. The first 20s of each measurement were cut before fitting in `fluxible`, which conservatively matches the deadband used in SoilFluxPro. CH_4 data could not be fitted with the Hutchinson–Mosier (HM) model due to the absence of curvature. Similarly, a large portion of measurements had not enough curvature to be fitted with the HM model and were therefore discarded, hence the smaller n for the exponential model.

TABLE 3 Comparison of *fluxible* with other R packages available on CRAN: *flux* (Jurasinski et al., 2022), *HMR* (Pedersen, 2024) and *gasfluxes* (Fuss, 2024).

Criteria	<i>fluxible</i>	<i>flux</i>	<i>HMR</i>	<i>gasfluxes</i>
One function per step	Yes	No	No	No
Homogeneity of processing	Yes	Yes	No	No
Easy integration in an Open Science pipeline	Yes	No	No	Yes
Input in logger format	<i>licoread</i> R package	No	No	No
Function to match meta data	<i>flux_match</i>	Yes	No	No
Fit models	Linear, quadratic, exponential	Linear	Linear, exponential	Linear, exponential
Reproducible quality flagging of fluxes	<i>flux_quality</i>	Yes	Partly	Yes
Quality summary of the dataset	<i>flux_quality</i> & <i>flux_flag_count</i>	No	No	No
User set quality thresholds	<i>flux_quality</i>	Partly	Partly	Partly
Plots of the measurements	<i>flux_plot</i> (pdf output possible)	As objects	As objects	Yes
Model comparison for each flux	No	No	Yes	Yes
Measurement level variables for flux calculation ^a	Volume, air temperature, pressure, plot area	Volume, air temperature, pressure	Volume	Volume
Transformation of other variables when calculating fluxes	Yes	No	No	No
Additional calculations	<i>flux_diff</i> , <i>flux_lrc</i>	GPP modelisation based on ER and PAR	No	No

^aThe *HMR* and *gasfluxes* packages calculate gas fluxes directly from volumetric concentration, and therefore do not require pressure nor temperature variables. As most gas analysers provide concentration in fractional concentration, the user has to do the transformation themselves beforehand.

functionalities clearly shows that *fluxible* goes further, with a reproducible flagging of the fluxes based on user-set thresholds and a quality summary at the dataset level. Finally, *fluxible* is the only package that processes other environmental data while calculating the fluxes.

5 | CONCLUSION AND FURTHER DEVELOPMENTS

The *fluxible* R package provides a flexible way of processing raw data from flux chambers using a variety of different models and quality checks (see [Supporting Information: Appendix A](#) for a detailed example). The main advantages of *fluxible* are (1) a homogeneous workflow to process ecosystem gas fluxes, (2) a choice of models and flexibility in quality control and (3) a transparent and reproducible workflow that can be integrated in an Open Science pipeline.

Further developments of *fluxible* include (1) a segmentation tool to automatically select the focus window, based on residuals or on environmental variables and (2) extending the workflow to transform the concentration in mole per volume before fitting a model, which would account for temperature changes during the measurements better. Feature requests are always welcome, and assistance through customised import functions is also possible.

AUTHOR CONTRIBUTIONS

All students and collaborators in the courses and projects where the conceptualization of the *fluxible* R package was discussed were offered to participate in the manuscript. Joseph Gaudard, Richard J. Telford, Julia Chacon-Labela, Hilary Rose Dawson, Brian J. Enquist, Joachim P. Töpper, Jonas Trepel, Vigdis Vandvik, Marta Baumane, Kristine Birkeli, Mukhlis Jamal Musa Holle, Paul Efen Santos-Andrade, Tin Widyani Satriawan and Aud H. Halbritter: Conceptualization. Joseph Gaudard, Brian J. Enquist, Joachim P. Töpper, Vigdis Vandvik and Aud H. Halbritter: Funding acquisition. Joseph Gaudard: Project administration. Joseph Gaudard and Richard J. Telford: Software. Richard J. Telford, Joachim P. Töpper, Vigdis Vandvik and Aud H. Halbritter: Supervision. Joseph Gaudard, Julia Chacon-Labela, Hilary Rose Dawson, Brian J. Enquist, Joachim P. Töpper, Jonas Trepel, Vigdis Vandvik and Aud H. Halbritter: Writing—original draft. Joseph Gaudard, Richard J. Telford, Julia Chacon-Labela, Hilary Rose Dawson, Brian J. Enquist, Joachim P. Töpper, Jonas Trepel, Vigdis Vandvik, Marta Baumane, Kristine Birkeli, Mukhlis Jamal Musa Holle, Jason R. Hupp, Tin Widyani Satriawan and Aud H. Halbritter: Writing—review and editing.

ACKNOWLEDGEMENTS

The *fluxible* R package is the result of discussions and methodology conceptualization about ecosystem gas fluxes data processing during the Plant Functional Traits Courses (PFTC) 6 and 7, and the

ThreeD and INCLINE projects. The ThreeD project was funded by the Research Council of Norway KLIMAFORSK project 287801 (2019–2023). The INCLINE project was funded by FRIMEDBIO project 274712 (2018–2022). The PFTC6 course was funded under the Research Council of Norway projects RECITE INTPART project 274831 (2018–2021), ExperTS INTPART project 287784 (2019–2022) and the Norwegian Centre for International Cooperation in Education (SIU) project TraitTrain HNP-2015/10037 (2016–2018). We also would like to thank Vincent Belde who, as an early user of fluxible during his master thesis, came up with valuable improvement suggestions, and Adam Klimes for his help in analytics and R coding.

CONFLICT OF INTEREST STATEMENT

The authors declare no conflict of interest.

PEER REVIEW

The peer review history for this article is available at <https://www.webofscience.com/api/gateway/wos/peer-review/10.1111/2041-210X.70161>.

DATA AVAILABILITY STATEMENT

Data used for the validation (Section 3) and in the data preparation example (Supporting Information: Appendix B) available via <https://doi.org/10.17605/OSF.IO/SRJ24> (Gaudard, 2025a). Data used in the detailed example (Supporting Information: Appendix A) and the workflow with two gases (Supporting Information: Appendix C) are part of fluxible. Code of the package (fluxible v1.3.1) available via <https://doi.org/10.5281/zenodo.17093479> (Gaudard & Telford, 2025). Code for the validation section and figure production available via <https://doi.org/10.5281/zenodo.17093752> (Gaudard, 2025b).

ORCID

Joseph Gaudard  <https://orcid.org/0000-0002-6989-7624>

Richard J. Telford  <https://orcid.org/0000-0001-9826-3076>

Julia Chacon-Labelle  <https://orcid.org/0000-0002-6787-1295>

Hilary Rose Dawson  <https://orcid.org/0000-0003-4613-762X>

Brian J. Enquist  <https://orcid.org/0000-0002-6124-7096>

Joachim P. Töpper  <https://orcid.org/0000-0002-6996-7223>

Jonas Trepel  <https://orcid.org/0009-0004-7358-9204>

Vigdis Vandvik  <https://orcid.org/0000-0003-4651-4798>

Marta Baumane  <https://orcid.org/0000-0003-0184-9252>

Kristine Birkeli  <https://orcid.org/0000-0003-3980-4420>

Mukhlis Jamal Musa Holle  <https://orcid.org/0000-0003-1899-2135>

Jason R. Hupp  <https://orcid.org/0000-0003-4505-5390>

Paul Efren Santos-Andrade  <https://orcid.org/0000-0002-6635-0375>

Tin Widyani Satriawan  <https://orcid.org/0000-0003-4580-9973>

Aud H. Halbritter  <https://orcid.org/0000-0003-2597-6328>

REFERENCES

- Arnone, J. A., & Obrist, D. (2003). A large daylight geodesic dome for quantification of whole-ecosystem CO₂ and water vapour fluxes in arid shrublands. *Journal of Arid Environments*, 55(4), 629–643. [https://doi.org/10.1016/S0140-1963\(02\)00291-4](https://doi.org/10.1016/S0140-1963(02)00291-4)
- Bastviken, D., Wilk, J., Duc, N. T., Gålfalk, M., Karlson, M., Neset, T.-S., Opach, T., Enrich-Prast, A., & Sundgren, I. (2022). Critical method needs in measuring greenhouse gas fluxes. *Environmental Research Letters*, 17(10), 104009. <https://doi.org/10.1088/1748-9326/ac8fa9>
- Forbrich, I., Kutzbach, L., Hormann, A., & Wilmking, M. (2010). A comparison of linear and exponential regression for estimating diffusive CH₄ fluxes by closed-chambers in peatlands. *Soil Biology and Biochemistry*, 42(3), 507–515. <https://doi.org/10.1016/j.soilbio.2009.12.004>
- Fuss, R. (2024). Gasfluxes: Greenhouse gas flux calculation from chamber measurements. <https://doi.org/10.32614/CRAN.package.gasfluxes>
- Gaudard, J. (2025a). Fluxible. <https://doi.org/10.17605/OSF.IO/SRJ24>
- Gaudard, J. (2025b). Jogaudard/fluxible-MEE: Fluxible-mee. Zenodo. <https://doi.org/10.5281/ZENODO.17093752>
- Gaudard, J., & Telford, R. (2025). Plant-Functional-Trait-Course/fluxible: fluxible. Zenodo. <https://doi.org/10.5281/ZENODO.17093479>
- Hüppi, R., Felber, R., Krauss, M., Six, J., Leifeld, J., & Fuß, R. (2018). Restricting the nonlinearity parameter in soil greenhouse gas flux calculation for more reliable flux estimates. *PLoS One*, 13(7), e0200876. <https://doi.org/10.1371/journal.pone.0200876>
- Hutchinson, G. L., & Mosier, A. R. (1981). Improved soil cover method for field measurement of nitrous oxide fluxes. *Soil Science Society of America Journal*, 45(2), 311–316. <https://doi.org/10.2136/sssaj1981.03615995004500020017x>
- Johannesson, C.-F., Nordén, J., Lange, H., Silvennoinen, H., & Larsen, K. S. (2024). Optimizing the closure period for improved accuracy of chamber-based greenhouse gas flux estimates. *Agricultural and Forest Meteorology*, 359, 110289. <https://doi.org/10.1016/j.agrformet.2024.110289>
- Jurasinski, G., Koebisch, F., Guenther, A., & Beetz, S. (2022). Flux: Flux rate calculation from dynamic closed chamber measurements. <https://doi.org/10.32614/CRAN.package.flux>
- Kutzbach, L., Schneider, J., Sachs, T., Giebels, M., Nykänen, H., Shurpali, N. J., Martikainen, P. J., Alm, J., & Wilmking, M. (2007). CO₂ flux determination by closed-chamber methods can be seriously biased by inappropriate application of linear regression. *Biogeosciences*, 4(6), 1005–1025. <https://doi.org/10.5194/bg-4-1005-2007>
- Maes, S. L., Dietrich, J., Midolo, G., Schwioger, S., Kumm, M., Vandvik, V., Aerts, R., Althuizen, I. H. J., Biasi, C., Björk, R. G., Böhrer, H., Carbognani, M., Chiari, G., Christiansen, C. T., Clemmensen, K. E., Cooper, E. J., Cornelissen, J. H. C., Elberling, B., Faubert, P., & Dorrepaal, E. (2024). Environmental drivers of increased ecosystem respiration in a warming tundra. *Nature*, 629, 1–9. <https://doi.org/10.1038/s41586-024-07274-7>
- Pedersen, A. R. (2024). HMR: Flux estimation with static chamber data. <https://doi.org/10.32614/CRAN.package.HMR>
- Pihlatie, M. K., Christiansen, J. R., Aaltonen, H., Korhonen, J. F. J., Nordbo, A., Rasilo, T., Benanti, G., Giebels, M., Helmy, M., Sheehy, J., Jones, S., Juszczak, R., Klefoth, R., Lobo-do-Vale, R., Rosa, A. P., Schreiber, P., Serça, D., Vicca, S., Wolf, B., & Pumpanen, J. (2013). Comparison of static chambers to measure CH₄ emissions from soils. *Agricultural and Forest Meteorology*, 171, 124–136. <https://doi.org/10.1016/j.agrformet.2012.11.008>
- R Core Team. (2024). R: A language and environment for statistical computing. R Foundation for Statistical Computing.

Zhao, P., Hammerle, A., Zeeman, M., & Wohlfahrt, G. (2018). On the calculation of daytime CO₂ fluxes measured by automated closed transparent chambers. *Agricultural and Forest Meteorology*, 263, 267–275. <https://doi.org/10.1016/j.agrfor.2018.08.022>

SUPPORTING INFORMATION

Additional supporting information can be found online in the Supporting Information section at the end of this article.

Data S1. fluxible R package.

How to cite this article: Gaudard, J., Telford, R. J., Chacon-Labelle, J., Dawson, H. R., Enquist, B. J., Töpper, J. P., Trepel, J., Vandvik, V., Baumann, M., Birkeli, K., Holle, M. J. M., Hupp, J. R., Santos-Andrade, P. E., Satriawan, T. W., & Halbritter, A. H. (2025). *fluxible*: An R package to process ecosystem gas fluxes from closed-loop chambers in an automated and reproducible way. *Methods in Ecology and Evolution*, 00, 1–9. <https://doi.org/10.1111/2041-210X.70161>