Dear Gerd Folberth,

We wish to thank our two reviewers for their time and considered comments on our manuscript. Below we lay out their comments in full and include our replies in bold font.

Anonymous Referee #1

This manuscript describes the CLASSIC v1.0 model, which is a combination of the CLASS and CTEM models. In particular CLASSIC represents a move to open-source development, with additional features to facilitate the use of the model by a diverse community in a range of settings. The paper also gives an overview of the performance of the model at a number of sites.

Overall the paper was well written and clear, and I consider it would be suitable for publication after minor corrections.

This will be a relatively short review as I didn’t find myself with many questions - essentially the paper appears to be a reasonable summary of essentially ‘technical’ developments to the modelling system. Many of my comments below are suggestions, which of course the authors might wish to argue against.

Thank you for your comments. We are glad to hear of your generally positive view of our work.

The only major thing that I think is missing is some discussion about governance - how will the community be run and governed so that the core model develops in a way that satisfies all or a majority of users? How will the trunk code be managed? How will the potentially conflicting needs of different user groups be managed? For example, I assume there are key stakeholders, such as the climate and earth system modellers who need a reliable, highly-optimised code, but equally there might be people who want to use CLASSIC to study a particular process that is perhaps only important in a small geographical area and/or to a few people but which requires a bigger change to the code base. That might be a rather extreme contrast, but my point is that I would like to know about the proposed process by which the main model trunk is managed - who is in control?

This is an interesting question. While we see some value in spelling out the model governance early, we feel the role of this paper is to focus on the technical and scientific transition of CLASS-CTEM to a community model format. As well, ideally, the model governance structure will be developed in partnership with the CLASSIC community. All of the questions listed above are likely to be ones that we will need to answer over time. While CLASSIC’s primary use, historically, has been the CanESM family of models, it also has significant applications in other regional climate models, hydrologic modelling, and site-level studies. It would disingenuous to claim to adopt a community approach and then decide a priori which model purpose is most important. That said, internal to ECCC, the model application will be tightly linked to the development of CanESM and its influence on CanESM performance will be closely scrutinized. In the end, decisions around model parameterization adoption are likely to be relatively simple. As we have been developing an extensive benchmarking apparatus for CLASSIC (AMBER), model development decisions will be tightly linked to model benchmarking results. Changes will be adopted if they lead to demonstrable improvements in model performance. This approach also appears to be the experience with CLM development (D. Lawrence, pers. comm. 2020).

Slightly less important, but recurring and so I will mention it here: On P5 and elsewhere it wasn’t clear to me whether CLASSIC supports the use of CLASS without CTEM, i.e. can one prescribe vegetation characteristics (such as height and LAI) rather than having these simulated by CTEM? I suspect this configuration is possible, but it is unclear. Such a configuration arguably makes some evaluations (such as against FLUXNET sites) more meaningful - e.g. looking at the turbulent fluxes is less interesting if CTEM has decided to produce vegetation that is very different from reality at a site (which also relates to the difficulties discussed in spinning up long-timescale processes with only a few years of meteorological data). And if this functionality does exist in CLASSIC, why not employ it in (at least some of) the FLUXNET-based evaluation described in Section 5? (I appreciate that prescribing vegetation characteristics is not a panacea for all of the difficulties outlines in Sec.5.)
Yes indeed, CLASSIC still retains the ability to turn off the biogeochemical cycling (CTEM). We have now added an explicit mention of the ability to turn off the biogeochemistry to Section 2.1, "While CLASSIC includes a biogeochemical component (CTEM; described below), it is possible to turn it off and run CLASSIC with specified vegetation structural attributes such as tree height, plant area index, etc, which is desirable in some situations." We didn't use this capability in this study as we are primarily interested in benchmarking how the model performs as a whole, i.e. including both the physics and biogeochemistry. The ability of the model to spinup the C pools (without the issue of limited years of meteorology) is investigated in a companion paper that will look at CLASSIC's global scale performance.

Specific comments

P2 L15 Benefits of OSS: I wonder if the creation of a community of collaborating modellers might be considered another benefit of the OSS/community approach (though arguably it might come under your first point ‘affordability’).

We feel that is a benefit from the community aspect, and not necessarily an OSS benefit. This links into our answer to the comment below.

Also on L10 you appear to equate OSS development with being a community model, but these could be considered separate - OSS means the code is accessible to all, but a community model might imply more than that in terms of there being a coordinated approach across groups to develop and use the model. I think it is this second sense that you intend, consistent with the aims of the other models you cite (JULES, CLM, etc.). Perhaps you could expand upon some of these aspects.

Yes, we didn’t wish to imply OSS and a community model are one and the same. Any model that has put their code in a public repository could argue that they are OSS. To be a community model though, requires a stronger connection between collaborating groups as well as a sense of ownership over the model and its direction. To better convey that sentiment we have reword the beginning of the paragraph from:

"OSS development in land surface modelling presents several benefits to its participants including: 1) affordability, creating a new land surface scheme is a massive undertaking; 2) transparency, as the code is open to full scrutiny; 3) flexibility, the models are designed to be both used in their present configuration and also extended to answer new science questions; and 4) perpetuity, many users across diverse institutions help protect the code against loss, deletion, or obsolescence. However…"

to

"A community approach to land surface modelling presents several benefits to its participants … new science questions; 4) perpetuity, many users across diverse institutions help protect the code against loss, deletion, or obsolescence; and 5) a collaborative community invested in developing, applying, and improving a common resource - the model. However…”

I'm not convinced that L16 'flexibility of configuration' is a benefit from OSS. Flexibility is a desirable characteristic of the model, but by itself OSS does not give this (but flexibility can become more affordable if OSS provides a bigger community of developers). I think you should remove this from the benefits section - or at least rephrase/recast the idea.

The revision of that paragraph (above) to highlight the community aspects may now be more convincing. We agree any OSS project has little requirement for flexibility of configuration. However, as mentioned by the referee, as a community develops around a model, users will wish to adapt the model to be able to handle different applications. This then increases the models flexibility of configurations.

P1 L29 key features: You later describe how CLASSIC provides a means to check that results are reproducible (checksums etc.), which could just about fall under your 8th point here (benchmarking). To me this is an important aspect in its own right and might deserve explicit mention here.

Thanks for the suggestion. We have amended, "8) extensive benchmarking" to, "8) extensive benchmarking of model state (via checksums) and performance"
P3 L30 'standard offline model setup' - what is this and where can a user find it (and other configurations)? Later you mention that various runs can be set up and performed using the information provided as part of the package/container, but in general I was unclear as to where (if at all) standard configurations can be obtained. Version controlled code and the container environment are important, but obviously the model set up is also critical. At some point in the manuscript I would like to see some (or clearer) coverage of where standard configurations can be acquired. On P3 it might suffice to point to this information, e.g. "see Section X".

If the steps in Appendix A are followed, the model source code, all model inputs and configuration files, and the container can be used to produce all results presented here. In these files is the standard model configuration. We agree that model setup is vital for correct model performance (Easterbrook’s configurability as mentioned in the Introduction) which is why we have provided all model scripts needed to exactly reproduce our work. Additionally, each user will have 31 examples of how the model setup was set up for the FLUXNET sites used in our benchmarking suite. To make this more clear we have added the following text to Section 2.1, “the standard model setup is provided for each of the 31 FLUXNET sites; please see Appendix A”

P9 L7-18: The level of detail and discussion about checksums is unnecessary. The basic principles are well-established and can (most likely) be cited. Most or all of these lines can be removed.

We agree there is nothing earthshaking here. However we anticipate this paper being read by grad students new to our model, and indeed to modelling in general. As a result, since it is relatively short, we would like to retain it for that future audience

P9 Sec 3.2.3: I had a quick look at some of the material on Zenodo and it is possible that I didn’t find the correct material, but I am wondering if there is any collated scientific documentation for CLASSIC. (In fact I’m thinking of science equations rather than the more code-specific documentation that is the focus of this section. I suspect the answer is ‘no’, as is the case for many models. Section 2 outlines the science and provides various references that can be followed up, but for many models it takes a keen student and days of work to assemble the definitive description of the science in the latest version of the model. As far as I can see the self-documenting code is very useful but does not address the more detailed scientific documentation. Are there any plans for further documentation? As I said, many models face similar challenges, so this is not a particular criticism of CLASSIC alone!)

Actually all model documentation can be found in the code and does indeed contain the scientific basis, i.e. equations, for much of the model subroutines. The documentation within the CLASSIC source code can be found in documentation/html/index.html. It is also available from the CLASSIC webpage (e.g. the competition parameterization: https://cccma.gitlab.io/classic/namespacecompetitionscheme.html)

P10 L25-end: The examples of how to select domains are unnecessary detail in such an overview paper; it is sufficient to know that the capability exists. I suggest removing this.

Agreed, these have been removed.

P11 L7-8: I don’t understand this. It reads as if FORTRAN is used to convert namelist files into netCDF. Perhaps the ASCII files listed in the namelists are covered into netCDF? If so, rephrase.

It is correct as written. A Fortran namelist file is used to specify all model initialization values, then a Fortran program reads the namelist and produces a netCDF file that CLASSIC can ingest for model initialization. The conversion script could have been written in another language, but, for simplicity and code reuse, we used Fortran. Most strictly, a namelist file is an ASCII file that follows a formatting convention which Fortran can easily parse. The legacy ASCII files we mentioned were both free and fixed format files that did not correspond to the namelist formatting rules and hence were considerably more difficult to work with.

P13 L22: I’m only moderately familiar with ILAMB but I thought it mainly (or solely) used colours to show rankings of results (rather than whether thresholds are passed). Also, even more pedantically, I think ILAMB and other tools can be very useful when assessing and presenting how different versions or configurations of a single model compare - in which case your objection to its use for a single model (L23) would not hold.
There we reference Figure 1 from Collier et al., 2018 which uses ‘stoplight’ colours to denote different thresholds of model performance. We agree that benchmarking tools such as iLAMB and AMBER are very useful for single model comparisons. However, our specific objection in the line mentioned is the use of colour schemes that convey the existence of important thresholds in the score values, as would be implied by the use of ‘stoplight’ colours.

P6 L13: Being able to use the same code in standalone and coupled applications is, of course, highly advantageous. I know other models also struggle to ensure that a single code base can be used for both (with the main difficulties often being at the higher-level interfaces to parent structures). To me this is an important consideration that could (possibly) be brought up earlier in the manuscript (perhaps as an important aspiration for CLASSIC, even if in fact it is still in the future). As it is, some readers might be slightly confused by the earlier claims that the models are used in coupled applications, yet at this late stage they are told the codes differ (even if in fact the differences are largely ‘technical’ rather than ‘scientific’).

Yes, it is indeed not simple to have complete coherence between the offline and coupled model codes. To be clear we have added the following, “While future releases of CanESM will include CLASSIC, CanESM5 contains older model versions (CLASS v.3.6.2 and CTEM v. 1.1).” to Section 6.

Minor comments

P3 L10 ‘second generation’ - I never like these terms as they are so vague, but it’s just about OK as you quote it! I can’t recall who first coined the term (Sellers, Shuttleworth?) but perhaps you could consider a citation (or explanation) to back it up.

True, it can be confusing. Here we add the definition provided by the creator of CLASS, “The Canadian Land Surface Scheme (CLASS) was initiated in 1987 to produce a ‘second generation’ land surface scheme, characterized by more soil moisture and thermal layers with a separate treatment of the vegetation canopy (Verseghy, 2000), for inclusion in the Canadian general circulation model (GCM) and has been under continual development since.”

P5 L14 "nine PFTs that map directly onto...": Is it that the 9 PFTs of CTEM are collapsed onto the 4 PFTs used by CLASS, so that two or more CTEM PFTs are represented by one CLASS PFT? This could be made clearer. I assume there is some motivating evidence that supports that this is a reasonable approach that saves CPU time.

Yes, that is the case. The reasoning was that the physics does not need to know if, for e.g. the broadleaf tree has an evergreen or deciduous habit (its LAI/PAI would convey that information indirectly). While there is a small computational cost savings, this feature comes largely from the model development history.

P5 L28 fire: I realise Fig.1 is just a schematic, but possibly some representation of fire (and land use) could be included.

The figure has been revised to include fire and land use change.

P8 I found the order of some sections in this area was slightly unexpected. Looking back, perhaps it was mainly that I thought the containerisation section might come later, after code design etc., as it is essentially a way to bundle everything that has been created/described previously. Coding standards and documentation I would also push up the list. The move to Git/Gitlab I would describe before other details such as how to add new PFTs. Similarly, P9 L20 mentions the move away from fixed-form FORTRAN… but mentally I have already checked outputs using checksums in the previous section. To me anything about the code repository and coding standards comes before checksums. But really this is personal preference, and your order is fine!

Yes, arguably there could be many different ways to organize these sections. Since the referee allows this to fall under personal preference, we would prefer to retain the original structure.

P10 L20: As I read this, point runs can produce gridded (2-D) outputs. Is that really the case?

No, it isn’t. To correct this we have amended this sentence from, "To run the model at a point location, the model input and output files can be either, also site-level with point-scale information, or regional (two-dimensional fields)." to, "To run the
model at a point location, the model input files can be either, also site-level with point-scale information, or regional (two-dimensional fields)."

P11 L22 and others: I'd prefer "3 or 6" rather than "3/6" which I keep reading as a fraction. I assume CLASSIC is fairly flexible in terms of timestep lengths and data intervals, but this section (and maybe others) does read a bit as if only certain timesteps are allowed. e.g. L21 says inputs are typically 3 or 6 hourly (which is fine), but later in the paragraph we are talking about "3/6" hourly data and a 30 minute timestep - i.e. it becomes specific. Consider clarifying that the model is indeed more flexible than this.

Yes, the model is indeed more flexible. While we generally use the offline model on a 30 minute physics timestep with 3 or 6 hourly meteorology, any combination where the meteorology is some multiple of the physics timestep could work, such as a 15 minute physics timestep with 2 hour meteorology. We have rephrased this section to make this more apparent, "For global offline simulations, reanalysis meteorological variables are typically available on either a 3 hour or 6 hour timestep, but it could be any multiple of the physics timestep. To convert the reanalysis meteorological data to the offline CLASSIC physics timestep, CLASSIC disaggregates the coarse temporal resolution meteorology on the fly. Surface pressure, specific humidity, wind speed and surface temperature are linearly interpolated. Long-wave radiation is uniformly distributed across the reanalysis time period. Shortwave radiation is distributed diurnally using the day of year and a grid cell's latitude with the maximum value occurring at solar noon. The total reanalysis time period precipitation amount determines the number of wet half-hours in each period following (Arora, 1997). In a conservative manner, the total time period's precipitation amount is then randomly spread across the wet physics timestep periods. If CLASSIC is being run with observed meteorology, such as from an eddy covariance tower, on a 15 - 30 minute timestep, the meteorology is used without modification."

P12 Sec4.2: Consider clarifying at this stage that AMBER is included in the container (as is later made clear).

We have changed "All AMBER outputs from site-level evaluation of CLASSIC v. 1.0 are included in the model benchmarking archive (Table 2)." at the end of Section 4.2 to "All AMBER outputs from site-level evaluation of CLASSIC v. 1.0 are in the model benchmarking archive (Table 2) with AMBER itself included in the CLASSIC software container."

**Typos etc.**

For all typos listed, we have either adopted the change or defer to the GMD copy editor for whether the change is needed.

P3 L31: add a comma after 61m.

P10 L20: 'also' seems to be misplaced here. Remove?

P11 L4: remove 'which'

P12 L12: remove 'and'

P13 L21 (and at least one other location): change "don't" to "do not"

P14 L1 'as well': I'd prefer 'additionally'.

P15 L11: change 'Both' to 'both'

P18 L24: Misplaced comma (and I'd remove the first comma on L25 too).

P39 L38: I haven't pored over all the references, but by chance I noticed a slightly dubious entry here: '0, null'.

**Anonymous Referee #2**
Moving CLASS-CTEM to an open source community model is an important initiative. This paper is a well-written discussion – perhaps a little unbalanced in the level of technical detail between different sections and with a few typos to pick up in copy editing, but those are not major problems. I judge that is acceptable for publication essentially as it stands, but I have some minor suggestions and questions:

Thank you for your review

- soil freezing and fire are mentioned in the text but not represented in Figure 1.

We have revised the figure to include them

- can separate reference heights be provided for temperature and wind speed? They often differ, both in site measurements and global datasets - total precipitation is required as an input, but what flexibility is there in specifying how the model will divide it into snow and rain?

Yes in both cases. We have now added the following sentence to the Section 2.1 (Model physics: CLASS), "Differing measurement heights for the observed temperature and wind speed are accounted for. Since the forcing data typically does not separate precipitation into its rainfall and snowfall components, three options are available in the CLASSIC modelling framework, 1) a strict threshold of 0 $^\circ$C below which all precipitation is considered snowfall, 2) a linear gradual transition of rainfall into snowfall as temperature decreases from 2 $^\circ$C to 0 $^\circ$C, and 3) a non-linear transition from rainfall to snowfall as temperature decreases from 6 $^\circ$C to 0 $^\circ$C. Here, we use option 1."

- the single model snow layer is a historical feature of CLASS, but is there a scientific justification for maintaining it in a model with many more ground layers?

We have retained the single snow layer in our physics scheme as it continues to perform well. While outside the scope of this manuscript, early results from SnowMIP have indicated CLASS (CLASSIC’s physics model predecessor) is amongst the top performing models at numerous field sites studied (P. Bartlett pers. comm. 2020).

- other models have used FLUXNET data for benchmarking. Could a short discussion of CLASSIC performance in the context of other studies be added

While other models have used FLUXNET to benchmark their model, we don’t feel it is straight-forward to compare across the studies. Other studies may have tuned their model performance to the sites run, run only a small handful of sites, or been interested in only one aspect of model performance for a particular study making comparison difficult or impractical. To acknowledge the extensive use of FLUXNET in model evaluation, we have added some example studies that have used FLUXNET data to Section 4.1, "FLUXNET data has been used extensively to improve and evaluate land surface models hydrology, energy, and carbon fluxes as well as model processes (e.g., Blyth et al., 2010, Stockli et al. 2008, Melaas et al., 2013)"
CLASSIC v1.0: the open-source community successor to the Canadian Land Surface Scheme (CLASS) and the Canadian Terrestrial Ecosystem Model (CTEM) - Part 1: Model framework and site-level performance

Joe R. Melton¹, Vivek K. Arora², Eduard Wisernig-Cojoc¹, Christian Seiler¹, Matthew Fortier¹, Ed Chan³, and Lina Teckentrup⁴

¹Climate Research Division, Environment and Climate Change Canada, Victoria, B.C., Canada
²Canadian Centre for Climate Modelling and Analysis, Environment and Climate Change Canada, Victoria, B.C., Canada
³Climate Research Division, Environment and Climate Change Canada, Toronto, Ont., Canada
⁴Max Planck Institute for Meteorology, 20146 Hamburg, Germany

Correspondence: Joe Melton (joe.melton@canada.ca)

Abstract.
Recent reports by the Global Carbon Project highlight large uncertainties around land surface processes such as land use change, strength of CO₂ fertilization, nutrient limitation and supply, and response to variability in climate. Process-based land surface models are well-suited to address these complex and emerging global change problems, but will require extensive development and evaluation. The coupled Canadian Land Surface Scheme and Canadian Terrestrial Ecosystem Model (CLASS-CTEM) framework has been under continuous development by Environment and Climate Change Canada since 1987. As the open-source model of code development has revolutionized the software industry, scientific software is experiencing a similar evolution. Given the scale of the challenge facing land surface modellers, and the benefits of open-source, or community model, development, we have transitioned CLASS-CTEM from an internally developed model to an open-source community model, which we call the Canadian Land Surface Scheme including Biogeochemical Cycles (CLASSIC) v. 1.0. CLASSIC contains many technical features specifically designed to encourage community use including software containerization for serial and parallel simulations, extensive benchmarking software and data (Automated Model Benchmarking : AMBER), self-documenting code, community standard formats for model inputs and outputs, amongst others. Here we evaluate and benchmark CLASSIC against 31 FLUXNET sites where the model has been tailored to the site-level conditions and driven with observed meteorology. Future versions of CLASSIC will be developed using AMBER and these initial benchmark results to evaluate model performance over time. CLASSIC remains under active development and the code, site-level benchmarking data, software container and AMBER are freely available for community use.
1 Introduction

Open collaboration has revolutionized software development leading to a proliferation of open source software (OSS) projects. Notable successes include the internet browser Mozilla Firefox, office suite LibreOffice, the GNU/Linux operating system and its derivative operating system for mobile devices, Android. OSS also has a large impact in scientific computing. A Google Scholar search for ‘science open source software’ reveals 4.7 million hits (accessed June 21th, 2019) indicating a high level of activity. In the field of land surface modelling there are several well-known large-scale community (OSS) models including the Joint UK Land Environment Simulator (JULES; Best et al., 2011; Clark et al., 2011), the Community Atmosphere Biosphere Land Exchange model (CABLE; Haverd et al., 2018), the Community Land Model (CLM; Lawrence et al., 2019), and the Noah Multi-Parameterization land surface model (Noah-MP; Niu et al., 2011). As recently stated by WIRED Magazine, ‘Open source isn’t counterculture anymore. It’s the establishment’ (Finley et al., 2019).

OSS development in a community approach land surface modelling presents several benefits to its participants including: 1) affordability, creating a new land surface scheme is a massive undertaking; 2) transparency, as the code is open to full scrutiny; 3) flexibility, the models are designed to be both used in their present configuration and also extended to answer new science questions; and 4) perpetuity, many users across diverse institutions help protect the code against loss, deletion, or obsolescence; and 5) a collaborative community invested in developing, applying, and improving a common resource - the model. However, it is not clear that open code will necessarily lead to more open science. Easterbrook (2014) outlines several barriers to sharing of code in his commentary piece on the utility of open code for more open science. While he argues that open code should lead to better quality code, as he sees it, the main barriers to code sharing include portability (ability to run the code on different platforms), configurability (model setup to perform a simulation), entrenchment (historical reasons behind code development decisions), model-data blur (processing of model inputs impact upon model outputs), and provenance (reproduction of a model result).

Here we present the Canadian Land Surface Scheme including Biogeochemical Cycles (CLASSIC; v 1.0), the successor to the coupled model framework of the Canadian Land Surface Scheme (CLASS; Verseghy, 2017) and the Canadian Terrestrial Ecosystem Model (CTEM; Melton and Arora, 2016). In developing the model framework for CLASSIC, careful attention has been paid to exploit the benefits of OSS, while minimizing the risks outlined by Easterbrook (2014). The CLASSIC model framework includes several key features designed to encourage collaboration and community use including: 1) self-documenting code, 2) version control allowing source code management, distributed non-linear workflows, issue tracking, and wiki functionality, 3) native support for Network Common Data Format (netCDF) input and output along with conversion tools for ASCII legacy inputs, 4) ability for code to run both serially on personal computers and using message passing interface...
(MPI) on computing clusters, 5) output file description and metadata handled via a web interface, 6) model parameters read in from an external file, 7) containerization, and 8) extensive benchmarking of model state (via checksums) and performance. Each of these features will be expanded upon in the following sections.

In Section 2 we describe the submodels, CLASS and CTEM that form the scientific basis of CLASSIC and the initial state of the model framework at the start of CLASSIC development. Section 3 details the model developments implemented along with tools to help existing users migrate to the new model framework. Section 4 outlines our model evaluation and benchmarking approach while Section 5 presents the present state of CLASSIC as evaluated by the AMBER protocol. Section 6 describes future technical and scientific directions for CLASSIC.

2 Model Description

2.1 Model physics: CLASS

The Canadian Land Surface Scheme (CLASS) was initiated in 1987 to produce a 'second generation' land surface scheme, characterized by more soil moisture and thermal layers with a separate treatment of the vegetation canopy (Verseghy, 2000) for inclusion in the Canadian general circulation model (GCM) and has been under continual development since. The first publications introducing CLASS described the physics calculations for movement of heat and water through the soil and snow layers (Verseghy, 1991), and the physics algorithms for energy and moisture fluxes within the vegetation canopy as radiation and precipitation cascade through it, with an explicit thermal separation of the vegetation from the underlying ground (Verseghy et al., 1993). Development of CLASS has been predominantly within Environment Canada, a federal department of the Government of Canada (later renamed Environment and Climate Change Canada; ECCC) with the exception of an organized community effort as part of the Canadian Climate Research Network (1994 - 1997; Verseghy, 2000) as well as ad-hoc collaborations with the broader research community. CLASS is presently at version 3.6.2 (Verseghy, 2017). While CLASSIC includes a biogeochemical component (CTEM; described below), it is possible to turn it off and run CLASSIC with specified structural vegetation attributes such as tree height, plant area index, etc, which is desirable in some situations. As a land surface scheme, CLASS simulates the fluxes of energy, momentum, and water between the atmosphere and land surface (Figure 1).

CLASS operates on a sub-daily time-step that varies depending on its application. When within the framework of the most recent version of the Canadian Earth System Model (CanESM5; Swart et al., 2019), it operates at a time step of 15 minutes, consistent with the atmospheric component. A time-step of 30 minutes is used for offline, uncoupled simulations where it is driven with observed meteorological data. Differing measurement heights for the observed temperature and wind speed are accounted for. Since the forcing data typically does not separate precipitation into its rainfall and snowfall components, three options are available in the CLASSIC modelling framework. 1) a strict threshold of 0 °C below which all precipitation is considered snowfall, 2) a linear gradual transition of rainfall into snowfall as temperature decreases from 2 °C to 0 °C, and 3) a non-linear transition from rainfall to snowfall as temperature decreases from 6 °C to 0 °C. Here, we use option 1.
Vegetation characteristics are described by rooting depth, canopy mass, leaf area index (LAI), and vegetation height, which are then used in calculations of the transfers of energy, water, and momentum with the atmosphere. The physical land surface processes in CLASS are modelled using plant functional types (PFTs). Presently, global vegetation in CLASS is represented using four PFTs - needleleaf trees, broadleaf trees, crops and grasses. The structural vegetation characteristics of these PFTs are modelled as a function of driving meteorological data and atmospheric CO$_2$ concentration by CTEM, which handles the biogeochemical processes, as explained in the next section. The number of ground layers in CLASS can vary depending upon...
application. The standard offline model setup currently uses twenty ground layers starting with 10 layers of 0.1 m thickness, gradually increasing to a 30 m thick layer for a total ground depth of over 61 m, while three ground layers with thicknesses of 0.1, 0.25 and 3.75 m are used in CanESM5. Water fluxes are calculated for soil layers within the permeable soil depth of the ground column, but not the underlying bedrock layers, whereas temperatures are calculated for both soil and bedrock layers (depicted in Fig. 1). Both water fluxes and ground temperatures are calculated each time step. The permeable soil depth varies geographically and is read in from the model initialization file. Also calculated each time-step are the temperature, mass, albedo, and density of the single layer snow pack (where it exists), the temperature and interception and storage of rain and snow on the vegetation canopy, and the temperature and depth of ponded water on the soil surface. Each grid cell is independent with no lateral transfers of heat or moisture between them. While runoff can be routed to estimate streamflow, such as in CanESM5 using the approach of Arora et al. (1999) or the MEC-Surface & Hydrology System (MESH) framework (Pietroniro et al., 2007), in both cases, once the runoff leaves a grid cell, it does not modify the soil moisture of downstream grid cells.

Application and evaluation of the performance of CLASS over the course of its development has been extensive with dozens of publications using the model at point, regional, and global scales both in coupled and offline modes. CLASS has been applied in an offline context, i.e. forced with observed meteorology (e.g., Bailey et al., 2000; Kothavala et al., 2005; Roy et al., 2013; Bartlett et al., 2006; Brown et al., 2006; Wu et al., 2016; Verseghy and MacKay, 2017; Melton et al., 2019d), as the physical land surface component of regional climate models, e.g. CRCM (Ganji et al., 2015; Paquin and Sushama, 2014) and CanRCM (Scinocca et al., 2016), and integrated into each version of the Canadian Atmospheric Model (CanAM; von Salzen et al., 2013), Coupled Global Climate Model (CanCM), and Earth System Model (CanESM; Arora et al., 2011; Swart et al., 2019) since the early 1990s.

2.2 Model biogeochemistry: CTEM

Development of the Canadian Terrestrial Ecosystem Model (CTEM) began in the early 2000s at Environment Canada in response to the need for a land surface carbon cycle component for the CanESM. CTEM, which is presently at version 2.0 (Melton and Arora, 2016), couples with CLASS through the exchange of information describing the state of the physical land surface and overlying vegetation. CLASS provides CTEM with physical land surface information including soil moisture, soil temperature and net radiation. CTEM uses this information to simulate photosynthesis and prognostically calculate carbon in its three live vegetation components (leaves, stem, and roots) and two dead carbon pools (litter and soil). The carbon amounts in these five carbon pools evolve prognostically, in the default CLASSIC configuration, for nine PFTs that map directly onto the four PFTs used by CLASS. Needleleaf trees are divided into their deciduous and evergreen types, broadleaf trees are divided into cold deciduous, drought deciduous, and evergreen types, and crops and grasses are divided based on their photosynthetic pathways into C_3 and C_4 versions. The finer distinctions between PFTs in the CTEM PFTs is required for modelling biogeochemical processes. For instance, simulating leaf phenology prognostically requires the distinction between deciduous and evergreen versions of broadleaf trees. However, once the LAI has been dynamically determined by CTEM, CLASS only needs to know that this PFT is a broadleaf tree since the physics calculations do not require information about underlying the
deciduous or evergreen nature of the leaves. The prognostic carbon masses of leaf, stem, and root simulated by CTEM are
used to calculate the structural vegetation characteristics required by CLASS: rooting depth (using a dynamic root distribution; Arora and Boer, 2003), canopy mass, LAI, and vegetation height (Arora and Boer, 2005a). Other than these structural vegetation attributes, CTEM also provides canopy conductance values to CLASS based on photosynthesis calculations at the CLASS time step, as explained in Arora and Boer (2003). The remainder of the biogeochemical process simulated by CTEM (Figure 1) and the resulting vegetation dynamics operate on a daily time step. CTEM models all primary terrestrial ecosystem processes including maintenance and growth respiration (Arora and Boer, 2005a); heterotrophic respiration (Melton et al., 2015); tissue turnover, allocation of carbon, and phenology (Arora and Boer, 2005a); disturbance (fire; Arora and Boer, 2005b; Arora and Melton, 2018); competition for space between PFTs (Arora and Boer, 2006; Melton and Arora, 2016), and land use change (Arora and Boer, 2010).

CTEM also dynamically calculates wetland extent, methane emissions from wetlands and fires, and methane uptake by soils (Curry, 2007) as described in Arora et al. (2018). To determine the wetland extent, as the liquid soil moisture in the top soil layer increases above latitudinally dependent threshold values, the wetland fraction increases linearly up to a maximum value, equal to the flat fraction in a grid cell with slopes less than 0.2%. Methane emissions from wetlands are calculated by scaling the heterotrophic respiration flux from the model’s litter and soil carbon pools. A separate submodule simulates peatland specific processes following Wu et al. (2016).

Both CLASS and CTEM have the capability to be run in a ‘mosaic’ configuration in which grid cells are divided into separate and independent tiles that simulate their own energy and water balance. There is no lateral transfer of energy or water between tiles that lie within a grid cell. As a result, the soil moisture in the permeable soil layers and the temperature of soil and bedrock layers in each tile evolve independently of each other, despite being driven with same meteorological data. A grid cell may be divided into tiles using any criteria, e.g. by each PFT on its own tile, different soil textures, or peatlands vs. uplands. In contrast, with the ‘composite’ or single-tile approach, structural vegetation attributes (LAI, rooting depth, vegetation height, canopy mass, and albedo) are averaged over all PFTs for use in energy and water balance calculations. Soil texture for permeable soil layers is also common to all PFTs within a grid cell resulting in a single liquid and frozen soil moisture for each permeable soil layer, and a single temperature for each soil and bedrock layer for the grid cell. The effect of tiling on the basis of fractional coverage of PFTs has been evaluated on estimation of the terrestrial carbon sink (Melton and Arora, 2014), competition between PFTs (Shrestha et al., 2016), and the use of soil texture clusters as a tiling criteria has been evaluated by Melton et al. (2017).

Terrestrial biogeochemical processes simulated by CTEM have been evaluated at scales from site-level to global (e.g. Peng et al., 2014; Melton and Arora, 2014; Melton et al., 2015; Melton and Arora, 2016). The fire subroutine has been extensively evaluated as part of the Fire Model Intercomparison Project (FireMIP; Hantson et al., 2016; Forkel et al., 2019) in addition to being used to estimate carbon cycle implications of the reduction in global wildfire since the 1930s (Arora and Melton, 2018). The parameterization for competition between PFTs has been evaluated at the site-level (Shrestha et al., 2016) as well as at the global scale (Melton and Arora, 2016). CLASS-CTEM has contributed to the Global Carbon Project’s methane assessment by providing wetland methane emissions (Poulter et al., 2017). An assessment of natural methane emissions simulated by CTEM
is presented in Arora et al. (2018) who use a one-box model of atmospheric methane together with prescribed anthropogenic and geological sources to reproduce atmospheric methane concentrations consistent with the observational record over the historical period. CLASS-CTEM has also contributed regularly to the Global Carbon Project’s annual carbon budget analyses since 2016 (Le Quéré et al., 2016, 2018b, a).

5 Model development

3.1 Motivation

At the start of our project to transform CLASS-CTEM into CLASSIC, the model code base was not well-suited to modern computing and model development practices. CLASS-CTEM was written following the Fortran 77 standard, documentation was provided in stand-alone documents that often significantly lagged model development or scientific papers, code management did not use modern version control systems, model parameters such as the number of soil layers and PFTs were hard-coded into subroutines, and the offline framework (as opposed to the version coupled into the family of CanESM models) used fixed format ASCII text files for model inputs and outputs. While these issues are not unique to CLASS-CTEM, they made model development and evaluation challenging. For example, while ASCII inputs are reasonable for site-level simulations, they quickly become difficult to prepare and manage over large modelling domains, which entailed handling thousands of ASCII files that are poorly suited to parallel computation. Additionally as computer hardware or compiler versions change the model results are susceptible to issues of reproducibility due to a changing computational environment. While CLASS-CTEM has been used extensively by the Canadian research community, as evidenced by the publications cited in the previous section, the modelling framework was poorly designed for new users who faced a steep learning curve in attempting to run and understand the model. In our modernization and improvement of the CLASS-CTEM code base to develop the CLASSIC model, we have addressed these concerns and made several important changes to the model code, which we describe below.

3.2 Model developments to encourage community use

3.2.1 Containerization

A virtual machine (VM) emulates a physical computer allowing different operating systems to run on different hardware, for e.g., the Linux operating system on a Windows machine. A software container is a form of operating system virtualization, essentially a pared-down version of a virtual machine with only the necessary software required to complete the intended task. Both options have been used to facilitate the running of complex models. The Weather Research and Forecasting model (WRF; Hacker et al., 2016) has been implemented in both VM and container configurations, while the United Kingdom Chemistry and Aerosols (UKCA) composition-climate model has been implemented within the Met Office VM (Abraham et al., 2018). In choosing between a VM or container to provide users of CLASSIC, we chose a container due to its light computational overhead, and design geared to high performance computing (HPC) applications (Arango et al., 2017). The advantages of containers for scientific computing are numerous. Firstly, as the environment within the container is consistent across computer
systems, the reproducibility of the code is enhanced. Containers are encapsulated portable environments that can contain both the software and data dependencies, along with libraries needed by the software, ensuring all necessary environment variables are provided and consistent. Secondly, they prevent orphaning of old code due to environment changes. If a container image is retained, the container can recreate the same environment for the code each time it is run. This ability allows a simple means to rerun old code, without having to consider environment changes that would make running the code outside of the container onerous. Third, the use of a container makes it easier to get up and running with a model quicker as all dependencies are included in the container. Lastly, many HPC centres allow the use of containers on their systems reducing the need for users to reconfigure models for different server systems.

For CLASSIC, we use a Singularity container (https://www.sylabs.io/), which is portable to Linux, Mac, and Windows systems as well as HPC clusters, for which Singularity was specifically designed (Kurtzer et al., 2017). Tests of Singularity’s performance on HPC benchmarks have demonstrated a negligible performance overhead (Le and Paz, 2017). Singularity is already installed on many national, university and government HPC centres including Compute Canada, Argonne National Labs, Fermilab, National Supercomputer Centre (Sweden), amongst others. The CLASSIC container image is available from our Zenodo community page (https://zenodo.org/communities/classic/). Version 1.0 of our container contains all software libraries needed to run the model in serial or parallel (using MPI) as well as the capability to run our benchmarking software (described in Section 4.2). The general model workflow is outlined in Figure 2.

3.2.2 Code design and management

To permit flexibility in the code, CLASSIC’s parameters are read in at run time from a Fortran namelist file. This use of a namelist permits rapid testing of model sensitivity to parameter values, as the model does not require recompilation between tests. Additionally, new PFTs can be more easily added to the code as all parameters are located in the namelist file, rather than distributed throughout the code. However, by design, new PFTs cannot be introduced into CLASSIC without due care. Within the code, case structures are used at each PFT-level branching calculation to determine the code branch a particular PFT is assigned to. These case structures have integrated error checks, whereby an unknown PFT causes the model to abort with a flag thrown informing the user where in the code the model detected an unknown PFT. This safety check helps to encourage thoughtful introduction of new PFTs to the model.

In transitioning from CLASS-CTEM to CLASSIC, the code base was ported into the distributed version control system Git (https://git-scm.com/) and is distributed via the software development tool Gitlab (https://gitlab.com/ccma/classic). The Gitlab issues tracker and a Nabble forum (http://classic-message-board.158658.n8.nabble.com/) will be used to facilitate communication between CLASSIC users. The issues tracker can be used to submit bug reports, elaborate on new parameterizations, and discuss code changes whereas the forum can be used to ask other users for advice and assistance.

As a further aid to development, CLASSIC can output checksums to speed model development for code changes that should not impact upon the model outputs. The checksum subroutine computes content-based checksums (summation of the flipped bits, i.e. the 1 in binary representation) of several groups of variables after a run has completed.
Figure 2. CLASSIC model workflow. For each CLASSIC simulation the solid lines indicate mandatory model inputs or outputs. The OutputVariablesEditor can be used to edit the XML file that controls the model outputs and specifies their metadata (Section 3.2.8). Model parameters are read in from the CLASSIC parameters namelist file while the simulation options such as location of input files and model configuration options are read in from the CLASSIC job options file. The required model geophysical inputs vary depending on the model configuration chosen but meteorological inputs are always required. All model outputs can be switched on or off depending on the simulation except the model restart file. Checksums can be used to ensure changes to the model don’t impact model performance (see Section 3.2.2). The CLASSIC source code is combined with other supplementary information by Doxygen to produce the model documentation (see Section 3.2.3).

Checksums are an imperfect means of ensuring no logical changes, as two numbers may have different binary representations with the same number of flipped bits. However, as the number of variables checked increases, it becomes highly unlikely to render a false-positive. For example, if we take a data item of bit-length $n$, having $b$ flipped bits in its representation, the number of same-length data items with the same checksum can be expressed through the binomial coefficient

$$\binom{n}{b}.$$  \hspace{1cm} (1)

If we assume the worst case where $b = \frac{n}{2}$, then Equation 1 becomes

$$\binom{n}{n/2}.$$  \hspace{1cm} (2)

Dividing by the total number of possible values for an $n$-digit binary number, we find the probability of a false positive checksum to be:

$$\frac{\binom{n}{n/2}}{2^n}.$$  \hspace{1cm} (3)
So, for example, if the checksum is computed using only three 32-bit numbers, the probability of a false positive is about $2.87 \times 10^{-7}$.

### 3.2.3 Self-documenting Code

To modernize the CLASS-CTEM code base, it was first transformed from the fixed form (Fortran 77) to a free-form structure (allowing coding constructs permitted by the Fortran 2008 standard). Best practices suggest that it is most desirable to embed documentation within the software, as this increases the likelihood that when developers update code, they will update the documentation at the same time (Wilson et al., 2014). Our model documentation was then incorporated directly into the code using the syntax of the Doxygen documentation generator (http://www.doxygen.nl/). The use of Doxygen allows both the programmatic and scientific reasoning behind the code to be detailed within the generated manual along with variable dictionaries. This ensures that the documentation for a particular model version is always included in the model source code distribution, and lowers the burden for developers to maintain its currency as it can be edited as the source code is updated. The full Doxygen-generated CLASSIC documentation for version 1.0 can be found in the Zenodo archive and, for the most recent CLASSIC release, at https://cccma.gitlab.io/classic. As Doxygen is included in the CLASSIC container, the user can run ‘doxygen Doxyfile’ to generate a local version of the documentation as required.

### 3.2.4 Coding standards

As part of the code modernization effort, we refactored the code to follow a recently developed CLASSIC coding convention. Coding conventions help to ensure code portability, readability, and maintainability. In adopting our coding conventions, we have developed a tool (a ‘linter’) to enforce code quality that can be used on legacy code, or new code, to either change the code to meet the coding standards, or to flag suspect sections of the code for manual intervention. Our coding convention is available in the Supplementary Material and on the CLASSIC website. The linter, which is written in python, is available in the CLASSIC code repository.

### 3.2.5 Serial and parallel computations (MPI)

As CLASSIC can be run at the site-level as well as over gridded domains it is important to support both modes of operation while maintaining only one codebase. The newly created CLASSIC driver allows compilation for serial or parallel processing of the model code. For site-level simulations, serial compilation and running of the model is sufficient whereas, for speed, runs over gridded domains require parallel processing which uses message passing interface (MPI) directives within the CLASSIC code. Pre-compiler directives, used sparingly, allow the same code for both applications. The default compiler supplied within the CLASSIC container is the GNU compiler (gfortran for serial and mpif90 as wrapper around gfortran for parallel computation) with additional Makefile scripts for Intel and Cray compilers.
3.2.6 Simulation domains

CLASSIC is designed to be easily run over simulation domains from site-level to global. To run the model at a point location, the model input and output files can be either, also site-level with point-scale information, or regional (two-dimensional fields). For regional domains, the model grid can be specified either with one-dimensional vectors or two-dimensional grids of longitudes and latitudes. The 1D case includes regular grids equally spaced in longitude and latitude as well as regular Gaussian grids. The 2D case is more general and can support any rectangular grid. The model call determines whether a simulation is run across a region or at a point. When using a regular grid, the model domain can be set by calling CLASSIC with a longitude latitude box. For example, `-180.0/180.0/-90.0/90.0` specifies a global simulation, `-135.0/-105.0/45.0/90.0` a region encompassing western Canada, and `-105.0/50.5` a grid centred on 105°W and 50°30′N. When the input files are provided on an irregular grid, CLASSIC uses grid cell indexes, instead of geographic coordinates to delineate the domain for the simulation. Thus `-135.0/-105.0/45.0/90.0` could be specified on an irregular grid by the grid cell indexes of those corresponding coordinates, i.e. `39/150/50/150`.

3.2.7 NetCDF input/outputs

While CLASS-CTEM used ASCII text files for model input/output (I/O), this method is cumbersome over gridded domains. The CLASSIC framework uses netCDF files instead (https://www.unidata.ucar.edu/software/netcdf/). NetCDF is a machine independent data format that is self-describing, allowing extensive metadata within the files, and which is the most common data format within the land surface modelling community. Scripts are provided in the CLASSIC code repository to convert legacy ASCII input files (meteorology and model initialization files) into the new netCDF format. Additionally, for site level users, a Fortran tool is included to convert Fortran namelist files, which are easier to work with than the legacy ASCII files, into a netCDF file suitable for CLASSIC model initialization.

3.2.8 Output files (xml)

To best use netCDF as an output format requires writing the output file metadata at the time of the file’s creation. To facilitate this CLASSIC uses extensible markup language (XML) files that are edited through a web interface and the resulting files are validated using an adjacent schema. The web interface allows a user to configure the output variables, as well as add new variables and edit metadata, using any JavaScript-compatible browser. Once the changes are complete, the user may download an updated version of the modified XML configuration file for use by CLASSIC. Where possible CLASSIC output files are Climate and Forecast (CF)-compliant (http://cfconventions.org/) and use variable names consistent with the data request of the Coupled Model Intercomparison Project (CMIP6; http://clipe-services.ceda.ac.uk/dreq/mipVars.html).

3.2.9 Meteorological inputs

Meteorological inputs required by CLASSIC include downwelling shortwave and longwave radiation, surface precipitation rate, surface air pressure, specific humidity, wind speed, and air temperature together with a reference height at which these
quantities were measured. Within the CanESM framework these inputs are provided on the coupled model physics timestep meteorological variables are provided by the atmospheric component of the ESM at its time step of 15 minutes. For global offline simulations, reanalysis meteorological variables are typically available on either a 3 hour or 6 hour timestep, but it could be any multiple of the CLASSIC physics timestep. To convert the 3/6-hourly reanalysis meteorological data to the offline CLASSIC physics timestep of 30 minutes, CLASSIC disaggregates the coarse temporal resolution meteorology on the fly. Surface pressure, specific humidity, wind speed and surface temperature are linearly interpolated. Long-wave radiation is uniformly distributed across the 3/6-hourly reanalysis time period. Shortwave radiation is distributed diurnally using the day of year and a grid cell’s latitude with the maximum value occurring at solar noon. The total 3/6-hourly reanalysis time period precipitation amount determines the number of wet half-hours in each 3/6-hour period following Arora (1997). In a conservative manner, the total 3/6-hour period’s precipitation amount is then randomly spread across the wet half-hourly periods. For example, if the reanalysis meteorology is provided in 6 hour time periods, the non-zero precipitation for each 6 hours is spread across the number of wet 30 minute time steps as determined by the disaggregation scheme. If CLASSIC is being run with observed meteorology, such as from an eddy covariance tower, on a 15 - 30 minute timestep, the meteorology is used without modification.

4 Running and Benchmarking CLASSIC

4.1 Quick-start tutorial to run and benchmark CLASSIC

Appendix A contains a tutorial that guides the reader through downloading, compiling, and running CLASSIC over a set of FLUXNET sites. FLUXNET is a global network of micrometeorological tower sites that uses the eddy covariance method to measure the exchanges of carbon dioxide, water vapour, and energy between terrestrial ecosystems and the atmosphere (https://fluxnet.ornl.gov). FLUXNET data has been used extensively to improve and evaluate land surface models hydrology, energy, and carbon fluxes as well as model processes (e.g., Blyth et al., 2010; Stöckli et al., 2008; Melaas et al., 2013). The outputs of the model runs are then run through a benchmarking system described below. All figures presented here except the schematic figures (Figs. 1 and 2) are produced by the benchmarking software.

4.2 Automated Model Benchmarking (AMBER)

The increasing complexity of land surface models necessitates more advanced methods of model evaluation and assessment. Such methods have been developed in a number of collaborative projects including the Project for Intercomparison of Land Surface Parameterization Schemes (PILPS; Henderson-Sellers et al., 1993), the Global Land-Atmosphere Coupling Experiment (GLACE; Koster et al., 2006), the Protocol for the Analysis of Land Surface Models (PALS; Abramowitz, 2012), and the PALS Land Surface Model Benchmarking Evaluation Project (PLUMBER; Best et al., 2015), and the International Land Model Benchmarking (ILAMB; Collier et al., 2018).
ILAMB has developed a framework that summarizes model performance across multiple statistical metrics using a dimensionless skill score system implemented in an open-source benchmarking and diagnostics software tool for land surface model evaluation written in Python (Collier et al., 2018). The evaluation of CLASSIC presented in this study is based on ILAMB’s statistical framework, which we implemented in a new R package referred to as the Automated Model Benchmarking (AMBER) package (Seiler, 2019). The development of AMBER allowed us to tailor the ILAMB approach to CLASSIC model outputs allowing: (i) a seamless data ingestion that does not require pre-processing steps, (ii) the ability to evaluate CLASSIC in different simulation modes (i.e. global and regional simulations on differing grids, and site-level runs), and (iii) full control on how the statistical framework is implemented. AMBER uses a variety of observation-based reference data against which it evaluates CLASSIC. These data consist of global scale remote sensing-based products, eddy covariance flux towers and annual streamflow measurements at the mouth of major rivers. Site-level evaluations are based on eddy covariance flux tower data alone. All AMBER outputs from site-level evaluation of CLASSIC v. 1.0 are included in the model benchmarking archive (Table 2) with AMBER itself included in the CLASSIC software container.

4.2.1 Skill scores

In ILAMB (Collier et al., 2018) and AMBER the performance of a model is expressed through scores that range from zero to one, where higher values imply better performance. These scores are computed for each variable in five steps: (1) computation of a statistical metric, (2) nondimensionalization, (3) conversion to unit interval, (4) spatial integration, and (5) averaging scores computed from different statistical metrics. The statistical metrics considered are the bias, root-mean-square error, phase shift, interannual variability, and spatial distribution. For example, the scalar score for the bias is calculated as:

\[ \text{bias}(\lambda, \phi) = \bar{v}_{\text{mod}}(t, \lambda, \phi) - \bar{v}_{\text{ref}}(t, \lambda, \phi), \quad (4) \]

where \( \bar{v}_{\text{mod}}(t, \lambda, \phi) \) and \( \bar{v}_{\text{ref}}(t, \lambda, \phi) \) are the mean values in time \( t \), over a specified period of time, of a variable \( v \), which varies geographically as a function of longitude \( \lambda \) and latitude \( \phi \) for model and reference data, respectively. Nondimensionalization is achieved by dividing the absolute bias by the standard deviation of the reference data \( \sigma_{\text{ref}}(\lambda, \phi) \) that corresponds to the same period over which the mean of model simulated quantities is calculated:

\[ \varepsilon_{\text{bias}}(\lambda, \phi) = \frac{|\text{bias}(\lambda, \phi)|}{\sigma_{\text{ref}}(\lambda, \phi)}, \quad (5) \]

A bias score that ranges between zero and one is calculated next:

\[ s_{\text{bias}}(\lambda, \phi) = e^{-\varepsilon_{\text{bias}}(\lambda, \phi)}. \quad (6) \]

The spatial integration of \( s_{\text{bias}} \) leads to the scalar score:

\[ S_{\text{bias}} = s_{\text{bias}}. \quad (7) \]
The spatially integrated score values for the root-mean-square error ($S_{rmse}$), phase shift ($S_{phase}$), interannual variability ($S_{iav}$), and spatial distribution ($S_{dist}$) are found in a similar way although the actual calculation of the metrics is, of course, different as shown in Table A1. To demonstrate the intermediate statistical metrics for the component scores, the components of $S_{rmse}$ are listed in Table A2. The score values based on these metrics are then combined to derive a single overall score for each output variable:

$$S_{overall} = \frac{S_{bias} + 2S_{rmse} + S_{phase} + S_{iav} + S_{dist}}{1 + 2 + 1 + 1 + 1}. \quad (8)$$

$S_{rmse}$ is assigned twice as much value as the other metrics since we consider it more important than the other metrics.

Interpretation of benchmarking scores should be done carefully. While ILAMB uses ‘stoplight’ colours to distinguish different thresholds, at least in a visual sense, of model performance (e.g. see Figure 1 in Collier et al., 2018), we don’t adopt that approach. The use of thresholds may be a useful visual tool when distinguishing the performance of several models (as in Figure 1 of Collier et al., 2018), or versions of the same model, however here we are presenting the results of a single model and thus any thresholds chosen would be arbitrary. Following Collier et al. (2018) we also ‘... do not view these aggregate absolute scores as a determinant of good or bad models. We envision the scores as a tool to more quickly identify relative differences among models and model versions which the scientist must then interpret’. We also agree with Collier et al. (2018) that the absolute value of score is not particularly meaningful. As well, a perfect score is not achievable due to reference datasets having measurement error and uncertainty, a lack of consistency between datasets for the same variables, and a lower score may only highlight the need for an enhanced observational sampling effort, or the lack of an appropriate metric of model performance (Collier et al., 2018). Therefore, we do not use the model score as the basis to determine whether model performance is acceptable or deficient for different model outputs, but rather use these scores to define the initial model performance that future model developments will be compared against.

5 Evaluation of CLASSIC

We used three separate approaches to benchmark CLASSIC. First, we used FLUXNET sites (Pastorello et al., 2017) where CLASSIC was driven with observed meteorology and the model initialization file was set up to correspond to site conditions, i.e. vegetation composition and coverage, soil texture and permeable depth, based on publications describing the tower sites. These ‘site-level’ simulations were performed for the 31 FLUXNET sites listed in Table 1 and shown in Figure 3. We then evaluated the CLASSIC outputs against the eddy covariance (EC) tower derived quantities of net ecosystem exchange (NEE), gross primary productivity (GPP), ecosystem respiration (RECO), net radiation (RNS), ground (HFG), sensible (HFSS) and latent heat (HFLS) fluxes. The FLUXNET data is from the November 3rd, 2016 update to the FLUXNET2015 dataset. While these sites are limited in their spatial and temporal coverage, they do present the most realistic model setup which uses observed vegetation composition, soil textures and driving meteorology.

There are two limitations, which apply to most land surface models, when comparing model simulated data against observation-based estimates at FLUXNET sites. First, CLASSIC represents vegetation at the level of PFTs, and not at the species level.
That is, for example, at two temperate FLUXNET sites which have either Oak (Quercus) or Maple (Acer) deciduous broadleaf trees, CLASSIC represents these sites using the same set of model parameter values corresponding to its cold deciduous broadleaf PFT. While climate predominately determines energy, water, and carbon fluxes, in the real world, species level differences also play a role in modulating these fluxes. Second, the model simulated values are compared with observations only when the model’s carbon pools reach equilibrium after being driven repeatedly with available meteorological data. The chosen FLUXNET sites in our study have between 3 and 19 years of available meteorological data (Table 1). Since these available years of meteorological data may not be representative of the mean climate at a given site, and/or a given site may be recovering from events that occurred prior to the start of the EC tower installation, the comparison between observations and model simulated fluxes is somewhat confounded. The modelled annual NEE thus always sums to zero, by construction, while we know that in the real world, at most sites, GPP is currently higher than RECO making annual NEE positive with land acting as a sink of carbon in response to rising atmospheric CO₂ concentrations. Past episodic extreme climatic events (e.g. drought or extreme precipitation) and disturbances (e.g. prior timber harvest or tree planting such as site CZ-BKI (Sedlák et al., 2010)) or fire, e.g. at CG-Tch (Merbold et al., 2008)) that we are unable to take into account confound the mismatch between model and observation-based fluxes even more. In an ideal world, we would have historical meteorological data and disturbance history available at each site. All components of the carbon budget - GPP, RECO, and NEE - are affected by these issues, but since NEE is a residual of GPP and RECO, the model to observations comparison is confounded most significantly for NEE.

The other two benchmarking approaches drive CLASSIC with globally gridded data products of meteorological data, vegetation cover, soil permeable depth, and soil textures. In the first of these two approaches, CLASSIC is evaluated against globally gridded data sets for variables where observation-based data are available. The third and final benchmarking approach also uses gridded data products to drive CLASSIC, but the model outputs are evaluated against the entire FLUXNET2015 release of 212 sites by comparing the EC tower data against the model outputs from the CLASSIC grid cell that each tower lies within. The evaluation from these two approaches is presented in a companion paper that evaluates CLASSIC on a global scale (Seiler et al. 2019, in prep.).

5.1 Further performance metrics

We present here also the statistical metrics of root mean squared error and coefficient of determination (Pedregosa et al., 2011). The coefficient of determination is calculated as:

\[
R^2(y, \hat{y}) = 1 - \frac{\sum_{i=1}^{n}(y_i - \hat{y}_i)^2}{\sum_{i=1}^{n}(y_i - \bar{y})^2}
\]  

(9)
Figure 3. Locations of the 31 FLUXNET sites grouped by biome (Table 1). International Geosphere-Biosphere Programme (IGBP) land classification abbreviations used include Evergreen Broadleaf Forests (EBF), Evergreen Needleleaf Forests (ENF), Savannas (SAV), Deciduous Broadleaf Forests (DBF), Grasslands (GRA), Croplands (CRO), Open Shrublands (OSH), Closed Shrublands (CSH), Permanent Wetlands (WET), and Deciduous Needleleaf Forests (DNF).

where $\hat{y}$ is the predicted value of the $n$-th sample and $y_i$ is the corresponding observed value for total $n$ samples. $\bar{y}$ is found via

$$\bar{y} = \frac{1}{n} \sum_{i=1}^{n} y_i$$

and

$$\sum_{i=1}^{n} (y_i - \bar{y})^2 = \sum_{i=1}^{n} \epsilon_i^2$$

While the coefficient of determination is commonly termed $R^2$, in this formulation, the best possible goodness of fit is 1.0 while the $R^2$ value can go negative (as can be seen from Equation 9). The RMSE is calculated by

$$\text{RMSE}(y, \hat{y}) = \sqrt{\frac{1}{n} \sum_{i=0}^{n-1} (y_i - \hat{y}_i)^2}.$$  

5.2 Benchmarking observation-driven CLASSIC at FLUXNET sites

The overall AMBER scores for the comparison between the observation-driven CLASSIC outputs and the 31 FLUXNET sites are shown in Figure 4 for three energy fluxes (HFLS, HFSS, and RNS) and three carbon fluxes (GPP, NEE, and RECO). For the
Figure 4. AMBER site-level scores comparing CLASSIC to the 31 FLUXNET sites (Table 1). Higher scores indicate better model performance. Further AMBER plots are available in the benchmarking archive listed in Table 2.

In component scores, CLASSIC generally had higher phase and spatial distribution scores with lower scores for RMSE, which is weighted double the other metrics in calculating the overall score (see Section 4.2.1). Most overall scores are between 0.65 - 0.82 with the exception of NEE at 0.44. These scores are discussed more below.

5.2.1 Energy fluxes

The mean seasonal cycle of net radiation (RNS) simulated by CLASSIC compared against observation-based estimates from all FLUXNET sites is shown in Fig. 5. For each site, the plot title shows the site ID (Table 1) followed by the biome type in brackets. The average RMSE across the sites is 19 W m\(^{-2}\) with a median R\(^2\) of 0.86 (see Supplementary Table S1 for per site values). Sites with a pronounced seasonality, typically from the high latitudes, are better captured by CLASSIC compared to lower latitude sites where the seasonal cycle is a result of seasonality in cloudiness and precipitation. CLASSIC also generally has lower peak values than observations, which could indicate that the simulated albedo values are too high. The most poorly simulated sites include BR-Sa1 (EBF), ZM-Mon (DBF), RU-Che (WET), and RU-SkP (DNF). These sites, with the exception of BR-Sa1, have only 2 – 3 years of observations. The AMBER scores for RNS are all above 0.73 with an overall score of 0.82 and an RMSE score of 0.76, which are the highest scores for the energy fluxes.

The time series of the observed and simulated latent heat fluxes (HFLS) is shown in Figure 6. Only 23 of the 31 FLUXNET sites reported latent and sensible heat fluxes. While some of the observed fluxes appear questionable, e.g. both ZM-Mon and SD-Dem appear to lose HFLS seasonality for select periods of their records, in contrast, CLASSIC simulates a more consistent seasonal cycle for all years at these sites. CLASSIC consistently simulates lower latent heat fluxes than evident in the observations. However, non-closure of the energy budget at the FLUXNET sites is a well-known issue. The residual of net
radiation plus the ground heat flux minus the latent and sensible heat fluxes ranges from 10 - 30% of RNS across all FLUXNET sites (Wilson et al., 2002; Cui and Chui, 2019). Since CLASSIC enforces closure of the energy budget, this non-closure by the EC data complicates evaluation of simulated energy fluxes (both HFLS and HFSS). For example, the period of highest mismatch at AU-Tum also corresponds to years with the highest non-closure terms (Fig. 6 and Supplementary Figures S10 and S11) The mean RMSE across sites was 31.5 W m\(^{-2}\) with a median R\(^2\) of 0.08 (recall the coefficient of determination, as calculated here, can go negative, see Section 5.1. Supplementary Table S2 contains all per site values). The AMBER scores for HFLS were lowest in bias and RMSE (0.49 and 0.50, respectively) giving an overall score of 0.66.

The sensible heat fluxes (HFSS) simulated by CLASSIC were generally closer to the EC tower observations (Figure 7) than HFLS, but with a slight overestimate compared to observations. The mean RMSE across sites was 24.7 W m\(^{-2}\) with a median R\(^2\) of 0.31 (See Supplementary Table S3 for per site values). The AMBER scores for HFSS were higher than HFLS with the lowest component score being \(S_{rmse}\) with 0.52 and an overall score of 0.66, equal to the HFLS overall score.

### 5.2.2 Carbon fluxes

Figure 8 compares the simulated mean seasonal cycle of GPP to FLUXNET observations. While we removed implausible values from the observations, i.e. negative GPP, we did retain other data points that we suspect could be spurious (e.g. 2005 GPP at SD-Dem in Supplementary Fig. S16). Over all 31 sites, the CLASSIC outputs have an average RMSE of 0.07 kg C m\(^{-2}\) month\(^{-1}\) and a median R\(^2\) of 0.15 (See Supplementary Table 4 for per site values). The AMBER scores for GPP are lowest for \(S_{rmse}\) at 0.48 with a \(S_{overall}\) of 0.69.

For biomes which are present at multiple sites (EBF, ENF, GRA, DBF), we can draw some general conclusions about the model performance. The EBF sites are primarily located in the low latitudes with the exception of sites in Turkey (GH-Ank) and France (FR-Pue). These mid-latitude EBF sites are for *Quercus ilex* (evergreen Oak) which don’t correspond well to the CLASSIC formulation for evergreen broadleaf PFT, which is typically assumed to be a tropical PFT. CLASSIC generally has a small bias, simulating somewhat lower GPP for the EBF sites (excepting GH-Ank; Fig. 8) with an RMSE for those sites of 0.08 kg C m\(^{-2}\) month\(^{-1}\). This bias is not evident for the ENF sites where CLASSIC generally simulates the sites well with a mean RMSE of 0.05 kg C m\(^{-2}\) month\(^{-1}\) but with one Canadian site (CA-Qfo) significantly underestimated. The deciduous broadleaf forest biome sites indicate that CLASSIC may generally underestimate GPP for this biome. The grassland sites are generally well simulated (mean RMSE of 0.5 kg C m\(^{-2}\) month\(^{-1}\)) with the exception of US-Wkg where CLASSIC is unable to capture the peak GPP values. The Kendall grassland (US-Wkg) is a semi-arid site in southern Arizona, USA. The mean annual rainfall is 345 mm, of which most falls in the summer months (Scott et al., 2010). Due to the extreme moisture stress caused by the low precipitation and heavy drought in 2003 – 2006, CLASSIC is unable to properly establish the vegetation and attain the productivity level observed by the EC tower. This site demonstrates the challenges inherent in this kind of comparison. At US-Wkg, the native vegetation cover of \(C_4\) grasses interspersed with \(C_3\) shrubs was intact prior to the early 2000s drought. Post-drought most of the native bunchgrasses were dead, along with high shrub mortality, leading to the establishment of an invasive grass (Scott et al., 2010). The site is also lightly to moderately grazed. Changes in vegetation such as this, along with unique drought conditions (which are only partly captured in the observation period) are difficult to capture adequately by
CLASSIC if the goal is to evaluate a model across a large number of sites to capture the diversity of biomes globally. Other sites present similar difficulties to replicate the observed GPP due to complex histories or conditions. As a result it is important to compare across several sites to reduce the importance of the site-level conditions and to observe the model performance on the whole. The other biomes have too few sites to interpret the results in detail. CLASSIC presently does not represent shrubs as separate PFTs so the shrubland biomes are simulated with trees PFTs in place of shrubs.

Seasonal plots of simulated ecosystem respiration (RECO) are compared against FLUXNET estimated values in Fig. 9. CLASSIC has generally lower variability in RECO than suggested by EC-tower derived estimates, especially at low latitudes sites such as BR-Sa1, GF-Guy, MY-PSO, ZM-Mon, and SD-Dem. In evergreen needleleaf forest sites, the simulated RECO is generally higher than the FLUXNET values. This overestimate is consistent with high GPP values at these sites. Similarly, the low GPP simulated for the evergreen broadleaf forests yields a low biased RECO. MY-PSO is a notable exception with CLASSIC simulating smaller GPP but larger RECO than the EC-tower derived quantities. Also, while RU-SkP (DNF) was reasonably well simulated for GPP, CLASSIC simulates a much larger RECO than the EC-tower derived value. The mean RMSE across all sites is 0.06 kg C m\(^{-2}\) month\(^{-1}\) with a median R\(^2\) of 0.12 (See Supplementary Table 5 for per site values). The AMBER scores for RECO are similar to the GPP scores with a slightly lower overall score (0.65) but a higher \(S_{rmse}\) of 0.49.

NEE is directly observed at eddy-covariance towers. The comparison between model and observations for NEE is, however, confounded due to the issues discussed in Section 5. Due to these factors, the AMBER NEE scores are lower than all others presented here with an \(S_{rmse}\) of 0.33, and \(S_{dist}\) of 0, and an overall score of 0.44 (Site-level plots of NEE are included in the supplementary material).

6 Future directions for CLASSIC development

CLASSIC remains under active development. Scientific developments include the introduction of non-structural carbohydrates (NSC) (Asaadi et al., 2018), which addresses a known issue of a delayed spring leaf out for deciduous tree species. The NSC work also lays the groundwork for incorporation of a N-cycle in CLASSIC that is presently in development. Other works in progress includes the incorporation of high-latitude shrub PFTs for both the physical and biogeochemical aspects of this growth form. Following on from recent work demonstrating CLASSIC’s capable performance in simulating the physics of permafrost regions (Melton et al., 2019d), CLASSIC’s bulk soil C pools will be replaced with an explicit tracking of soil C per soil layer that is better suited to simulate permafrost C dynamics. As part of this work, a C tracer is being incorporated to allow tracking of \(^{14}\)C or other isotopes through the model C pools. Planned technical developments include further modularization of the physics code, greater adoption of data structures to move away from lengthy argument lists, and modifications to allow the same code to be seamlessly used offline as well as in the CanESM framework. While future releases of CanESM will include CLASSIC, CanESM5 contains older model versions (CLASS v.3.6.2 and CTEM v. 1.1). Benchmarking developments include adding peatland EC tower sites to allow evaluation of our peatland module (Wu et al., 2016) as well as sites for biomes that are presently poorly represented.
7 Conclusions

CLASSIC is the open source community successor to CLASS-CTEM. CLASS and CTEM have decades of development behind them and have been extensively evaluated and applied at scales from site-level, involving evaluation against data from EC towers, to coupling to an atmosphere model as part of CanESM. While the science within CLASS-CTEM has continued to advance in response to changing scientific questions and applications, the technical aspects of CLASS-CTEM have fallen behind. Our study details the transformation of CLASS-CTEM from a primarily internally developed model to one designed to encourage community use and development. Table 2 summarizes the URLs of CLASSIC resources.

The work detailed here to create CLASSIC addresses the barriers to code sharing identified by Easterbrook (2014): portability, configurability, entrenchment, model-data blur, and provenance. Specifically, we have addressed code portability by providing a software container for CLASSIC to run in serial or parallel modes, along with rewriting the model I/O to use community standard netCDF format files. Configurability of CLASSIC is handled by providing extensive documentation of appropriate model setup via Doxygen, as well as benchmarking sites that can be used as examples of how to configure CLASSIC appropriately along with scripts to properly setup and run CLASSIC over 31 FLUXNET sites. Entrenchment is addressed most easily going forward as we adopt an open community approach to model development including the use of issue trackers and forums to outline new model developments and bug fixes. Model-data blur is reduced by using the FLUXNET sites for our site-level benchmarking as they required minimal processing prior to use. Lastly, provenance is made possible by our use of a Singularity software container and our benchmarking pipeline (Appendix A) where any user can reproduce the site-level plots and AMBER scores presented here. AMBER is made available to the community as a means to guide model development where low scores highlight possible areas for improvement. Additionally the use of AMBER can help ensure that improvements in one area of CLASSIC are not degrading the performance of other variables. We encourage members of the scientific community to use, develop, and contribute back to CLASSIC.

Code and data availability. The CLASSIC software container, CLASSIC site-level FLUXNET benchmarking data and AMBER reports, and all code both for CLASSIC v. 1.0 and for preparing the plots of model outputs as well as site-level data presented here are archived on the CLASSIC community Zenodo page. (Melton et al., 2019a, b, c). See Table 2 for the location of all resources.
Table 1. FLUXNET2015 sites used in the FLUXNET site-level benchmarking (Section 5.2).

<table>
<thead>
<tr>
<th>Site ID</th>
<th>Site name</th>
<th>Latitude</th>
<th>Longitude</th>
<th>Elevation (m)</th>
<th>Years</th>
<th>IGBP code</th>
<th>DOI</th>
</tr>
</thead>
<tbody>
<tr>
<td>AU-Tum</td>
<td>Tumbarumba</td>
<td>-35.656</td>
<td>148.1517</td>
<td>1200</td>
<td>2001 - 2013</td>
<td>EBF</td>
<td>10.18140/FLX/1440126</td>
</tr>
<tr>
<td>BR-Sa1</td>
<td>Santarem-Km67-Primary Forest</td>
<td>-2.8567</td>
<td>-54.9589</td>
<td>88</td>
<td>2002 - 2011</td>
<td>EBF</td>
<td>10.18140/FLX/1440032</td>
</tr>
<tr>
<td>CA-Q6o</td>
<td>Quebec - E. Boreal, Mature Black Spruce</td>
<td>49.6925</td>
<td>-74.3421</td>
<td>382</td>
<td>2003 - 2010</td>
<td>ENF</td>
<td>10.18140/FLX/1440045</td>
</tr>
<tr>
<td>CA-TPD</td>
<td>Ontario - Turkey Point Mature Deciduous</td>
<td>42.6353</td>
<td>-80.5577</td>
<td>260</td>
<td>2012 - 2014</td>
<td>DBF</td>
<td>10.18140/FLX/1440112</td>
</tr>
<tr>
<td>CN-Dan</td>
<td>Dangxiang</td>
<td>30.4978</td>
<td>91.0664</td>
<td>4313</td>
<td>2004 - 2005</td>
<td>GRA</td>
<td>10.18140/FLX/1440138</td>
</tr>
<tr>
<td>CZ-BK1</td>
<td>Bily Kriz forest</td>
<td>49.5021</td>
<td>18.5369</td>
<td>875</td>
<td>2004 - 2014</td>
<td>ENF</td>
<td>10.18140/FLX/1440143</td>
</tr>
<tr>
<td>DE-Tha</td>
<td>Tharanndt</td>
<td>50.9624</td>
<td>13.5625</td>
<td>385</td>
<td>1996 - 2014</td>
<td>ENF</td>
<td>10.18140/FLX/1440152</td>
</tr>
<tr>
<td>DK-Sor</td>
<td>Soroe</td>
<td>55.4859</td>
<td>11.6446</td>
<td>40</td>
<td>1996 - 2014</td>
<td>DBF</td>
<td>10.18140/FLX/1440155</td>
</tr>
<tr>
<td>FR-Pue</td>
<td>Puechabon</td>
<td>43.7413</td>
<td>3.5957</td>
<td>270</td>
<td>2000 - 2014</td>
<td>EBF</td>
<td>10.18140/FLX/1440164</td>
</tr>
<tr>
<td>GF-Guy</td>
<td>Guyaflux (French Guiana)</td>
<td>5.2788</td>
<td>-52.9249</td>
<td>48</td>
<td>2004 - 2014</td>
<td>EBF</td>
<td>10.18140/FLX/1440165</td>
</tr>
<tr>
<td>IT-Lav</td>
<td>Lavanne</td>
<td>45.9562</td>
<td>11.2813</td>
<td>1353</td>
<td>2003 - 2014</td>
<td>ENF</td>
<td>10.18140/FLX/1440169</td>
</tr>
<tr>
<td>IT-SRo</td>
<td>San Rossoire</td>
<td>43.7279</td>
<td>10.2844</td>
<td>6</td>
<td>1999 - 2012</td>
<td>ENF</td>
<td>10.18140/FLX/1440176</td>
</tr>
<tr>
<td>IT-Tor</td>
<td>Torgnon</td>
<td>45.8444</td>
<td>7.5781</td>
<td>2160</td>
<td>1999 - 2012</td>
<td>GRA</td>
<td>10.18140/FLX/1440237</td>
</tr>
<tr>
<td>MY-PSO</td>
<td>Pasoh Forest Reserve</td>
<td>2.9730</td>
<td>102.3062</td>
<td>147</td>
<td>2003 - 2009</td>
<td>EBF</td>
<td>10.18140/FLX/1440240</td>
</tr>
<tr>
<td>NL-Loo</td>
<td>Loobos</td>
<td>52.1666</td>
<td>5.7436</td>
<td>25</td>
<td>1996 - 2014</td>
<td>ENF</td>
<td>10.18140/FLX/1440178</td>
</tr>
<tr>
<td>RU-Che</td>
<td>Cherski</td>
<td>68.6130</td>
<td>161.3414</td>
<td>6</td>
<td>2002 - 2005</td>
<td>WET</td>
<td>10.18140/FLX/1440181</td>
</tr>
<tr>
<td>RU-Fyo</td>
<td>Fyodorovskoye</td>
<td>56.4615</td>
<td>32.9221</td>
<td>265</td>
<td>1998 - 2014</td>
<td>ENF</td>
<td>10.18140/FLX/1440183</td>
</tr>
<tr>
<td>RU-Hal1</td>
<td>Hakasia steppe</td>
<td>54.7252</td>
<td>90.0022</td>
<td>446</td>
<td>2002 - 2004</td>
<td>GRA</td>
<td>10.18140/FLX/1440184</td>
</tr>
<tr>
<td>RU-Skp</td>
<td>Yakutsk Spasskaya Pad larch</td>
<td>62.2550</td>
<td>129.1680</td>
<td>246</td>
<td>2012 - 2014</td>
<td>DNF</td>
<td>10.18140/FLX/1440243</td>
</tr>
<tr>
<td>US-WCr</td>
<td>Willow Creek</td>
<td>45.8059</td>
<td>-90.0799</td>
<td>520</td>
<td>1999 - 2014</td>
<td>DBF</td>
<td>10.18140/FLX/1440095</td>
</tr>
</tbody>
</table>

International Geosphere-Biosphere Programme (IGBP) land classification abbreviations used include Evergreen Broadleaf Forests (EBF), Evergreen Needleleaf Forests (ENF), Savannas (SAV), Deciduous Broadleaf Forests (DBF), Grasslands (GRA), Croplands (CRO), Open Shrublands (OSH), Closed Shrublands (CSH), Permanent Wetlands (WET), and Deciduous Needleleaf Forests (DNF).
Figure 5. Mean seasonal monthly net radiation for the 31 FLUXNET sites and CLASSIC simulated values. Shaded regions indicate the standard deviation across the sample years. The years of observations for each site as listed along with the site name and IGBP biome which correspond to evergreen broadleaf forest, EBF; evergreen needleleaf forest, ENF; grasslands, GRA; deciduous broadleaf forest, DBF; croplands, CRO; open shrublands, OSH; permanent wetlands, WET; and deciduous broadleaf forests, DBF.
Figure 6. Monthly observed and simulated latent heat fluxes for 23 FLUXNET sites. See caption of Figure 5 for biome names. Shaded regions indicate the standard deviation across the sample years. Table 1 lists the full site names.
Figure 7. Monthly observed and simulated latent heat fluxes for 23 FLUXNET sites. See caption of Figure 5 for biome names. Shaded regions indicate the standard deviation across the sample years. Table 1 lists the full site names.
Figure 8. The mean seasonal cycle of GPP for 31 FLUXNET sites as simulated by CLASSIC. IGBP biome of each site is listed alongside its FLUXNET site ID (see Table 1). Shaded regions indicate the standard deviation across the sample years. See caption of Figure 5 for biome names.
Figure 9. The mean seasonal cycle of RECO for 31 FLUXNET sites as simulated by CLASSIC. IGBP biome of each site is listed alongside its FLUXNET name (see Table 1). Shaded regions indicate the standard deviation across the sample years. See caption of Figure 5 for biome names.
Table 2. Summary table of CLASSIC Resources. Resources that do not list a version number will be updated with the most recent versions through time.

<table>
<thead>
<tr>
<th>Resource</th>
<th>URL</th>
</tr>
</thead>
<tbody>
<tr>
<td>CLASSIC webpage</td>
<td><a href="https://cccma.gitlab.io/classic">https://cccma.gitlab.io/classic</a></td>
</tr>
<tr>
<td>CLASSIC Gitlab code repository</td>
<td><a href="https://gitlab.com/cccma/classic">https://gitlab.com/cccma/classic</a></td>
</tr>
<tr>
<td>CLASSIC Community Zenodo page</td>
<td><a href="https://zenodo.org/communities/classic/">https://zenodo.org/communities/classic/</a></td>
</tr>
<tr>
<td>CLASSIC code for v. 1.0</td>
<td><a href="https://zenodo.org/record/3522407">https://zenodo.org/record/3522407</a></td>
</tr>
<tr>
<td>CLASSIC Singularity container v. 1.0 (including AMBER v. 0.1.5)</td>
<td><a href="https://zenodo.org/record/3525249">https://zenodo.org/record/3525249</a></td>
</tr>
<tr>
<td>Benchmarking data and outputs for CLASSIC v. 1.0 (including those generated by AMBER)</td>
<td><a href="https://zenodo.org/record/3525336">https://zenodo.org/record/3525336</a></td>
</tr>
<tr>
<td>AMBER CRAN page</td>
<td><a href="https://cran.r-project.org/package=amber">https://cran.r-project.org/package=amber</a></td>
</tr>
<tr>
<td>CLASSIC discussion forum</td>
<td><a href="http://classic-message-board.158658.n8.nabble.com/">http://classic-message-board.158658.n8.nabble.com/</a></td>
</tr>
</tbody>
</table>
Appendix A: Quick start guide to running CLASSIC at FLUXNET sites

This document will walk through the basic setup, compilation, and running of CLASSIC at FLUXNET sites.

Requirements

This guide assumes that the reader is using a Linux machine, and that they have Singularity and tar installed. The installation guide for Singularity can be found at https://sylabs.io/docs/#singularity. Tar can be installed from https://www.gnu.org/software/tar/. If using a Windows or Mac machine see instructions on setting up Singularity at https://sylabs.io/guides/3.3/user-guide/installation.html#install-on-windows-or-mac.

Obtaining the Source Code

The first step is setting up a directory structure. For the sake of this guide, we will work out of a directory located at, 
`/home/eccc`. So first we navigate to our chosen directory location:

```
cd /home/eccc
```

The CLASSIC source code is hosted on GitLab, and can be cloned with:

```
git clone https://gitlab.com/cccma/CLASSIC.git
```

If you are familiar with git’s workflow and wish to contribute to the codebase in the future, then it is a good idea to first fork a copy to your own GitLab account. Once you have a fork, you can clone the repository with:

```
git clone https://gitlab.com/**your_gitlab_username**/CLASSIC.git
```

Once the cloning process is complete, there should be a directory titled CLASSIC located in your directory:

```
/home/eccc/CLASSIC
```

Obtaining other necessary files

All other necessary files are found on Zenodo in the form of compressed packages. We will use the `tar` command to unpack them beside the CLASSIC repository. First, navigate to the CLASSIC community Zenodo page at https://zenodo.org/communities/classic and download the FLUXNET, and CLASSIC_container tar files. Your directory should now contain the following:

```
/home/eccc/CLASSIC
/home/eccc/CLASSIC_container.tar.gz
/home/eccc/FLUXNET.tar.gz
```

28
From the ‘/home/eccc’ directory, we will decompress the tarballs and move the decompressed files to their correct locations. The following list of commands will accomplish this:

```
tar xzvf CLASSIC_container.tar.gz

tar xzvf FLUXNET.tar.gz

mv CLASSIC_container.simg CLASSIC/

mv FLUXNET/TRENDY_CO2_1700_2018.nc CLASSIC/inputFiles/

cmp FLUXNET/FLUXNETsites CLASSIC/inputFiles/

mv FLUXNET/observationalDataFLUXNET CLASSIC/inputFiles/

cmp FLUXNET/benchmark_CLASIC_output CLASSIC/benchmark/

mv FLUXNET/benchmark_CLASIC_output CLASSIC/benchmark/

mv FLUXNET/benchmark_CLASIC_AMBER CLASSIC/benchmark/

rm -rf FLUXNET/
```

15 Using the container

Navigate to the CLASSIC root directory. The rest of the tutorial will assume that you stay in this working directory:

```
cd /home/eccc/CLASSIC
```

Since much of CLASSIC’s functionality relies on the libraries contained within the container, we execute commands through the container with this syntax:

```
singularity exec CLASSIC_container.simg [commands here]
```

Most of the calls like this are contained within the automated run scripts so their direct use is minimized. However, if you wish to deviate from the automated run scripts, that command will be of great use.

First, we’ll make sure we have a clean working directory.

```
singularity exec CLASSIC_container.simg make mode=serial clean
```

We are now ready to compile the source code for serial simulation. This is done with the command:

```
singularity exec CLASSIC_container.simg make mode=serial
```

The compilation process can take several minutes. If you get errors, check that you have the latest stable version of CLASSIC pulled from the repository.
Setting up FLUXNET site-level runs

With the FLUXNET sites in place and CLASSIC compiled, we can now setup the job options file(s) to run over the FLUXNET sites. To do this, run the job options script by invoking:

```
tools/siteLevelFLUXNET/prep_jobopts.sh
```

This script puts a new job-options file in every FLUXNET site directory tailored to that particular site. The ‘singularity exec’ command is not necessary for this particular script.

Running CLASSIC over FLUXNET sites

This guide will cover two methods for running the CLASSIC binary on FLUXNET sites. They can be run individually, or through the batch-run script.

Running all sites with the batch script (recommended)

A script is provided in the CLASSIC repository to run all FLUXNET sites, provided the directory structure of this document is followed. The script will run for all sites it finds within the inputFiles/FLUXNETsites directory and is invoked by:

```
tools/siteLevelFLUXNET/run_sites.sh
```

All model restart files are provided in a ‘pre-spunup’ state. Still, running all sites can a lengthy process on older machines, and sometimes you may not want or need all of them. To reduce the number of sites being run, remove the undesired site directories from inputFiles/FLUXNETsites. For example, if you don’t need to process ‘DK-Sor’, you could do:

```
 mv inputFiles/FLUXNETsites/DK-Sor inputFiles
```

This would simply move the site out of the way, and it could be placed back into the FLUXNETsites directory when needed.

Running individual sites (advanced)

Individual sites are run through the Singularity container directly referencing the job options file for that particular site. The command takes the form:

```
singularity exec [Location of CLASSIC container] [location of CLASSIC binary]
[Location of job options file for that site] [Longitude/Latitude of the site]
```

Since this is a site-level simulation the shorthand 0/0 or 0/0/0/0 can be used in place of the actual longitude and latitude of the site. Piecing it all together, if we want to run on the site AU-Tum, the command would be:

```
singularity exec CLASSIC_container.simg bin/CLASSIC_serial \
inputFiles/FLUXNETsites/AU-Tum/job_options_file.txt 0/0/0/0
```
If you decide to shell into the singularity container (recommended if you’re familiar with Singularity and terminal commands), then this becomes

bin/CLASSIC_serial inputFiles/FLUXNETsites/AU-Tum/job_options_file.txt 0/0/0/0

More information on running CLASSIC can be found in the CLASSIC manual at https://cccma.gitlab.io/classic/index.html.

5 Processing output

Whether you ran multiple sites with the batch job, or a single site individually, the output can be found in ‘outputFiles/FLUXNETsites/sitename/’. These outputs are in the form of netCDF files. More information on netCDF can be found at https://www.unidata.ucar.edu/software/netcdf/, but briefly, it is a machine-agnostic array-oriented form of data storage.

A script is provided to convert this data to csv format, which may be easier to work with if unfamiliar with netCDFs, as well as generate plots of several output variables against observational values of those variables. The script is run with:

```
tools/siteLevelFLUXNET/process_outputs.sh
```

NOTE: if not all sites were run with the batch script, errors will be seen in the output of this script. This is expected, and does not mean that the processing has failed.

Once the output plots have been generated, the ‘process_outputs’ script will run the Automated Model Benchmarking (AMBER) tool. More information on AMBER can be found at https://cran.rstudio.com/web/packages/amber/amber.pdf.

After the script completes, PDF copies of the plots are found in ‘outputFiles/plots’, while AMBER results are in ‘outputFiles/AMBER’.
Table A1. Equations used for computing spatially integrated scalar scores for each variable. The four steps refer to (1) the definition of the statistical metric, (2) nondimensionalization, (3) conversion to unit interval, and (4) spatial integration. The integration limits \( t_0 \) and \( t_f \) are the initial and final time step, respectively. The variables \( c_{\text{mod}} \) and \( c_{\text{ref}} \) refer to the monthly mean annual cycle of the model and reference data, respectively. \( \text{crmse} \) is the ‘centralized RMSE’, which is the RMSE of the anomalies. This ensures that the RMSE score excludes also considering the bias, which is captured by the bias score.

<table>
<thead>
<tr>
<th>Score</th>
<th>Step</th>
<th>Equation</th>
</tr>
</thead>
<tbody>
<tr>
<td>( S_{\text{bias}} )</td>
<td>1</td>
<td>( \text{bias}(\lambda, \phi) = \text{v}<em>{\text{mod}}(\lambda, \phi) - \text{v}</em>{\text{ref}}(\lambda, \phi) )</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>( \varepsilon_{\text{bias}} = \left</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>( s_{\text{bias}}(\lambda, \phi) = e^{-\varepsilon_{\text{bias}}(\lambda, \phi)} )</td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>( S_{\text{bias}} = s_{\text{bias}} )</td>
</tr>
<tr>
<td>( S_{\text{rmse}} )</td>
<td>1</td>
<td>( \text{rmse}(\lambda, \phi) = \sqrt{\frac{1}{t_f-t_0} \int_{t_0}^{t_f} \left( \text{v}<em>{\text{mod}}(t, \lambda, \phi) - \text{v}</em>{\text{ref}}(t, \lambda, \phi) \right)^2 dt} )</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>( \varepsilon_{\text{rmse}}(\lambda, \phi) = \frac{\text{crmse}(\lambda, \phi)}{\sigma_{\text{ref}}(\lambda, \phi)} )</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>( s_{\text{rmse}}(\lambda, \phi) = e^{-\varepsilon_{\text{rmse}}(\lambda, \phi)} )</td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>( S_{\text{rmse}} = s_{\text{rmse}} )</td>
</tr>
<tr>
<td>( S_{\text{phase}} )</td>
<td>1</td>
<td>( \theta(\lambda, \phi) = \max(\text{c}<em>{\text{mod}}(t, \lambda, \phi)) - \max(\text{c}</em>{\text{ref}}(t, \lambda, \phi)) )</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>not applicable, as units are consistent across all variables</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>( s_{\text{phase}}(\lambda, \phi) = \frac{1}{2} \left[ 1 + \cos \left( \frac{2 \pi \theta(\lambda, \phi)}{365} \right) \right] )</td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>( S_{\text{phase}} = s_{\text{phase}} )</td>
</tr>
<tr>
<td>( S_{\text{iav}} )</td>
<td>1</td>
<td>( iav_{\text{ref}}(\lambda, \phi) = \sqrt{\frac{1}{t_f-t_0} \int_{t_0}^{t_f} \left( \text{v}<em>{\text{ref}}(t, \lambda, \phi) - c</em>{\text{ref}}(t, \lambda, \phi) \right)^2 dt} )</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>( iav_{\text{mod}}(\lambda, \phi) = \sqrt{\frac{1}{t_f-t_0} \int_{t_0}^{t_f} \left( \text{v}<em>{\text{mod}}(t, \lambda, \phi) - c</em>{\text{mod}}(t, \lambda, \phi) \right)^2 dt} )</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>( s_{\text{iav}}(\lambda, \phi) = e^{-\varepsilon_{\text{iav}}(\lambda, \phi)} )</td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>( S_{\text{iav}} = s_{\text{iav}} )</td>
</tr>
<tr>
<td>( S_{\text{dist}} )</td>
<td>1</td>
<td>( \sigma = \frac{\sigma_{\text{mod}}}{\sigma_{\text{ref}}} )</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>not applicable</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>not applicable</td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>( S_{\text{dist}} = 2(1 + R)/(\sigma + \frac{1}{\sigma})^2 ), where ( R ) is the spatial correlation coefficient of ( \text{v}<em>{\text{ref}}(\lambda, \phi) ) and ( \text{v}</em>{\text{mod}}(\lambda, \phi) )</td>
</tr>
</tbody>
</table>
Table A2. Site-level statistical metrics for calculating $S_{rmse}$ at the 31 FLUXNET sites. $crmse$ is the ‘centralized RMSE’, which is the RMSE of the anomalies. This ensures that the RMSE score excludes also considering the bias, which is captured by the bias score.

<table>
<thead>
<tr>
<th>variable</th>
<th>unit</th>
<th>rmse</th>
<th>crmse</th>
<th>$\sigma_{ref}$</th>
<th>$\epsilon_{rmse}$ (-)</th>
<th>$S_{rmse}$ (-)</th>
</tr>
</thead>
<tbody>
<tr>
<td>GPP</td>
<td>gC m$^{-2}$ day$^{-1}$</td>
<td>2.26</td>
<td>1.88</td>
<td>2.74</td>
<td>0.78</td>
<td>0.48</td>
</tr>
<tr>
<td>HFLS</td>
<td>W m$^{-2}$</td>
<td>32.37</td>
<td>22.31</td>
<td>31.50</td>
<td>0.72</td>
<td>0.50</td>
</tr>
<tr>
<td>HFSS</td>
<td>W m$^{-2}$</td>
<td>25.10</td>
<td>19.55</td>
<td>32.03</td>
<td>0.70</td>
<td>0.52</td>
</tr>
<tr>
<td>NEE</td>
<td>gC m$^{-2}$ day$^{-1}$</td>
<td>1.95</td>
<td>1.56</td>
<td>1.48</td>
<td>1.24</td>
<td>0.33</td>
</tr>
<tr>
<td>RECO</td>
<td>gC m$^{-2}$ day$^{-1}$</td>
<td>1.90</td>
<td>1.29</td>
<td>1.82</td>
<td>0.75</td>
<td>0.49</td>
</tr>
<tr>
<td>RNS</td>
<td>W m$^{-2}$</td>
<td>19.34</td>
<td>12.32</td>
<td>48.56</td>
<td>0.28</td>
<td>0.76</td>
</tr>
</tbody>
</table>
Author contributions. J.M. lead the conversion of CLASS-CTEM to CLASSIC, coordinated activities, and wrote the first draft of the manuscript. V. A. created CTEM, originally suggested the possibility of converting CLASS-CTEM to a community model, and contributed to the planning and implementation of CLASSIC along with contributing text. C. S. wrote AMBER, provided AMBER outputs and wrote sections of the manuscript. E. W-C wrote CLASSIC code around netCDF I/O, the MPI incorporation, the meteorological disaggregation, and the XML editor and also contributed text to the manuscript. E.C. worked on the model compilation scripts, aspects of CLASSIC code, and its use on supercomputers. L. T. processed FLUXNET data and setup the model initialization files for the tower sites used in the site-level evaluation. M.F. wrote and applied the linter used to refactor the code, drafted the coding standards documents, developed the parallel container recipe, and wrote the plotting scripts. All authors contributed to the final manuscript.

Competing interests. The authors declare no competing interests

Acknowledgements. The authors wish to acknowledge the meticulous work of Diana Verseghy who lead development of the CLASS model from its origins until her retirement in January, 2017. We thank C. Le Quéré for allowing us to distribute her CO₂ record that was originally made for the TRENDY project. L.T.’s visit to CCCma was funded by Hamburglobal.

This work used eddy covariance data acquired and shared by the FLUXNET community, including these networks: AmeriFlux, AfriFlux, AsiaFlux, CarboAfrica, CarboEuropeIP, CarboItaly, CarboMont, ChinaFlux, Fluxnet-Canada, GreenGrass, ICOS, KoFlux, LBA, NECC, OzFlux-TERN, TCOS-Siberia, and USCCC. The ERA-Interim reanalysis data are provided by ECMWF and processed by LSCE. The FLUXNET eddy covariance data processing and harmonization was carried out by the European Fluxes Database Cluster, AmeriFlux Management Project, and Fluxdata project of FLUXNET, with the support of CDIAC and ICOS Ecosystem Thematic Center, and the OzFlux, ChinaFlux and AsiaFlux offices.


Haverd, V., Smith, B., Nieradzik, L., Briggs, P. R., Woodgate, W., Trudinger, C. M., Canadell, J. G., and Cuntz, M.: A new version of the CABLE land surface model (Subversion revision r4601) incorporating land use and land cover change, woody vegetation demography,


