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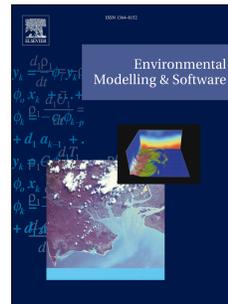
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1 ***Crop2ML: An open-source multi-language modeling framework for the***
2 ***exchange and reuse of crop model components***

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41 Highlights

- 42 • An open framework is developed (Crop2ML) to support the development, exchange, and reuse of
43 crop model components.
- 44 • Crop2ML provides a shared format to describe the specifications of model units and their
45 composition.
- 46 • Crop2ML generates model components considering crop simulation platforms artifacts.
- 47 • Crop2ML model life cycle can be managed using a user-friendly JupyterLab interface.
- 48 • Crop2ML can be used to import and export model components into many different crop
49 simulation platforms.

50 **Abstract**

51 Process-based crop models are popular tools to analyze and simulate the response of agricultural
52 systems to weather, agronomic, or genetic factors. They are often developed in modeling platforms
53 to ensure their future extension and to couple different crop models with a soil model and a crop
54 management event scheduler. The intercomparison and improvement of crop simulation models is
55 difficult due to the lack of efficient methods for exchanging biophysical processes between modeling
56 platforms. We developed Crop2ML, a modeling framework that enables the description and the
57 assembly of crop model components independently of the formalism of modeling platforms and the
58 exchange of components between platforms. Crop2ML is based on a declarative architecture of
59 modular model representation to describe the biophysical processes and their transformation to
60 model components that conform to crop modeling platforms. Here, we present Crop2ML framework
61 and describe the mechanisms of import and export between Crop2ML and modeling platforms.

62

63 **Keywords:** Crop model, Crop2ML, component-based software, model exchange and reuse.

64

65 **Software availability**

66 Software name: Crop2ML

67 Contact: Dr. Pierre Martre (pierre.martre@inrae.fr) or Dr. Christophe Pradal

68 (christophe.pradal@cirad.fr)

69 Year first available: 2020

70 Hardware required: IBM compatible PC or Apple

71 Operation System required: Windows 10, Mac OS X, or Linux

72 Program languages: CyML, Python

73 Availability: <https://doi.org/10.5281/zenodo.3911713>

74 1. Introduction

75 The wide range of crop process-based models (PBM) reflects the evolution of our knowledge of
76 the soil-plant-atmosphere system and the rich historical development for more than five decades
77 (reviewed in Jones et al. 2017; Muller and Martre 2019). The high diversity of PBM is due to their
78 multiple applications and the complexity of the system influenced by several factors, e.g. weather,
79 soil, crop management (Basso et al., 2013) and genotypic factors (Wang et al., 2019). Most of the
80 PBM are continuous models, formalized using ordinary differential equations, but are implemented
81 as discrete time simulation models using finite difference equations. They are commonly
82 decomposed into simpler biophysical functions (e.g. phenology, morphogenesis, resource
83 acquisition, pests and diseases impact) often implemented by recurrent equations with control flows.
84 Another common characteristic is that PBM simulate plant growth and development at the scale of
85 the canopy or average plant level without spatial dependence with a daily or sub-daily time step.

86 PBM are often implemented in modeling and simulation platforms at a higher level of abstraction
87 to facilitate model development (Rizzoli et al., 2008) . These platforms offer not only scalable,
88 modular, and robust modelling solutions but also the ability to analyze, evaluate, reuse and combine
89 models. The diversity of PBM led the crop modeling community to compare their performance and
90 to improve them by aggregating modelers' knowledge or by introducing improvements provided
91 from diverse research groups under the umbrella of large international collaborative projects such as
92 the Agricultural Model Intercomparison and Improvement Project (AgMIP; Rosenzweig et al. 2013).
93 Studies conducted in the context of model intercomparison and improvement exercises (e.g. Asseng
94 et al. 2013; Wang et al. 2017) pointed out the large uncertainty of PBM simulations and have
95 analyzed the sources of uncertainty or the processes involved. These intercomparison results showed
96 the potential and limits of PBM and highlighted the need to analyze models at the process level, but
97 also to exchange model components describing specific processes between simulation platforms
98 (e.g. Donatelli et al. 2014; Wang et al. 2017). The uncertainty of a PBM component may be related to
99 its validity domain, inputs, parameters, structure, and the underlying scientific hypotheses (Walker et
100 al., 2003). Epistemic uncertainty may arise from incomplete or lack of knowledge of these sources.
101 The uncertainty of PBM results from the aggregation of the uncertainty of each of its component
102 (Refsgaard et al., 2007). A framework that would allow the exchange of model components between
103 different platforms would give crop modelers the ability to test alternative hypotheses in the same
104 model, thus helping to reduce epistemic uncertainty.

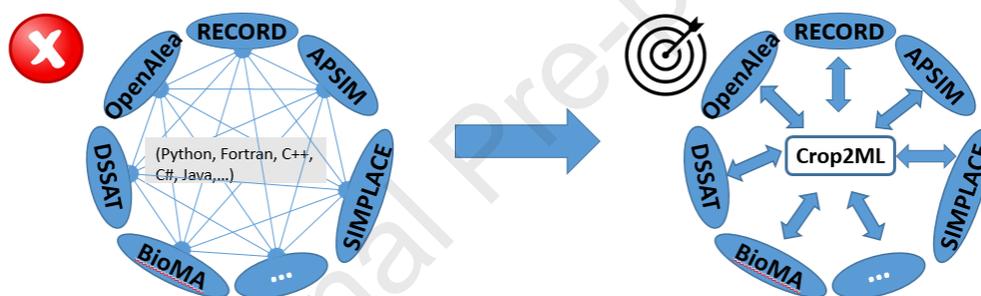
105 Although most crop simulation platforms provide modular approaches and reuse techniques,
106 there is little exchange of PBM components between them despite theoretical and application
107 interests. PBM components often contain source code developed in different programming

108 languages and are tightly coupled to the platforms. Therefore, model components are not seamlessly
109 reusable outside the modeling platforms in which they have been developed without recoding or
110 wrapping them (Holzworth et al., 2014; Rizzoli et al., 2008). Re-implementing a component in several
111 platforms is a tedious and cumbersome task and requires a minimum knowledge of the different
112 platforms. The wrapping solution treats components as black boxes taking little or no advantage of
113 the framework (Rizzoli et al., 2008) or as white boxes but with a high-level of complexity (Fernique &
114 Pradal, 2018; Pradal et al., 2008). Other reuse approaches in environmental modeling have been
115 explored. Declarative modeling can provide portability and facilitate integration between
116 independent, uncoordinated models (Athanasiadis and Villa (2013). However, model specifications
117 are seldom separate from implementation details. Model builders rely often directly on
118 implementation that hides the scientific content of a model (i.e. its algorithm) and its structure.
119 Moreover, the publication of PBM components in scientific journals does not provide sufficient
120 description associated with the modeled processes, which is a fundamental criterion for reuse
121 (Pradal et al., 2013). This raises the problem of reproducibility and reliability of scientific results that
122 are strongly linked to the platforms in which the models have been implemented and tested (Cohen-
123 Boulakia et al., 2017; Hinsen, 2016).

124 Visual domain-specific languages such as Simile (Muetzelfeldt and Massheder 2003) or Stella
125 (Richmond, 1985) provide a rich graphical interface to build models but become difficult to use for
126 complex models and require many widgets to represent graphically nested control flows. Multiscale
127 modelling and simulation frameworks (Marshall-Colon et al., 2017; Pradal et al., 2015) propose
128 model interface designs which enables communication of multi-language components as black box
129 components. Other declarative modelling languages are also used in the Systems Biology community
130 who have developed declarative open standard such as SBML (Hucka et al., 2010), CELLML (Cuellar et
131 al., 2003), or NEUROML (Le Franc et al., 2012) to describe biological models. However, crop modelers
132 generally use procedural modelling rather than a mathematical formalism like differential or reaction
133 equations as it is commonly done in System biology.

134 An alternative to the problem of PBM component reuse between PBM platforms is the use of a
135 centralized framework that enables the development of PBM components regardless of the
136 modeling platforms (Fig. 1). We followed this approach and developed a modeling framework called
137 Crop2ML (Crop Modelling Meta Language) that separates the structure of a model component from
138 its implementation. Given that the wrapping solution was excluded because of the lack of
139 transparency and high maintenance cost and that Crop2ML does not aim at replacing existing
140 modeling platforms or at simulating components within large modeling solutions (crop models), we
141 created a solution that generates components, from a metalanguage, for specific PBM platforms. It
142 provides a centralized PBM components repository to store model components in a standard format

143 to facilitate their access and reuse. This reuse approach is supported by the Agricultural Modeling
 144 Exchange Initiative (AMEI), which brings together some of the most widely used crop modelling and
 145 simulation platforms, including the Agricultural Production Systems sIMulator (APSIM, Holzworth *et*
 146 *al.*, 2018), the Biophysical Model Applications (BioMA; Donatelli *et al.*, 2010), the Decision Support
 147 System for Agrotechnology Transfer (DSSAT; Jones *et al.*, 2003; Hoogenboom *et al.*, 2019), OpenAlea
 148 (Pradal *et al.*, 2015), the REnovation and COORDination of agroecosystems modelling (RECORD;
 149 Bergez *et al.*, 2013), and the Scientific Impact assessment and Modeling Platform for Advanced Crop
 150 and Ecosystem management (Simplace; Gaiser *et al.*, 2013) and other crop models such as STICS
 151 (Brisson *et al.*, 2010) or *SiriusQuality* (Martre *et al.*, 2006). Here, we first present the main
 152 components of Crop2ML framework. Then we describe the mechanisms of importing and exporting
 153 between Crop2ML and PBM platforms. We then discuss our approach and present some
 154 perspectives.



155
 156 **Fig. 1.** From a combinatorial to a centralized exchange framework. The schema illustrates the
 157 reduction of import export links between platforms in a centralized (right) versus combinatorial
 158 exchange framework.

159 2. Crop2ML: a centralized framework for crop model components development and sharing

160 Crop2ML is a framework for crop model component development, exchange, and reuse between
 161 PBM platforms. It is designed following FAIR principles for research software (Lamprecht *et al.*, 2019)
 162 to provide:

- 163 • *Simplicity*: Model specifications are defined using a declarative language (eXtensible Markup
 164 Language [XML]; Bray *et al.*, 2008) with generic concepts shared between PBM platforms and
 165 model algorithms are encoded using a minimal language.
- 166 • *Transparency*: Models are shared as documented components in a well-defined format
 167 (Crop2ML format).
- 168 • *Flexibility*: Model units are composed with a shared abstract representation of model structure.

- 169 • *Findability*: Model specifications include rich metadata and are assigned a globally unique and
170 persistent identifier for each released version.
 - 171 • *Reusability*: Model components are transformed into PBM platform-compliant code to support
172 efficient interoperability.
 - 173 • *Reproducibility*: Model components can be executed and tested regardless of the PBM
174 platforms.
 - 175 • *Modularity*: Three levels of modularity of models are defined: (single) model units, composite
176 models and package. Package contains model units and composite as well as data. It provides
177 the flexibility to make different compositions based on these models.
- 178 We used the principles of Lamprecht et al. (2019) for assessing the FAIR-ness of Crop2ML framework
179 (Supplementary data Table C1).

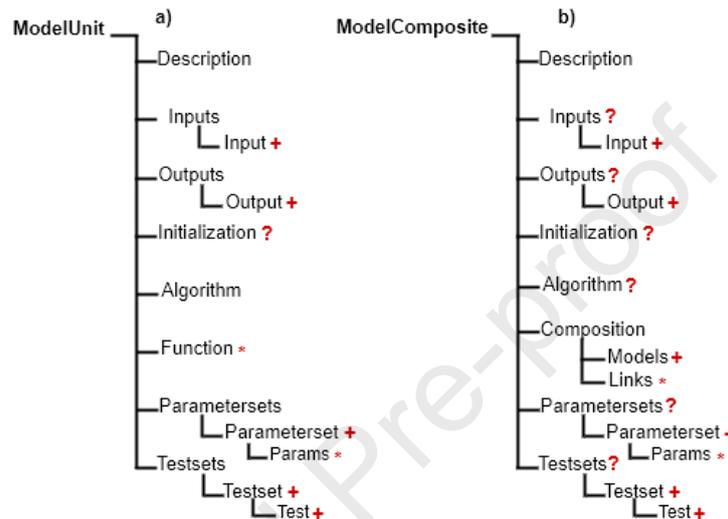
180 2.1. *Design and concepts of Crop2ML model specification*

181 Software modularity is one of the main criteria of reuse. Jones et al., (2001) proposed key elements
182 for modular model structure, which is an essential first step to enhance collaborative modelling
183 effort. Crop2ML follows and extends these principals. In most PBM, the system is decomposed into
184 compartments such as plant parts or soil layers that interact. For each compartment, different
185 processes are described and assembled in components to simulate the response of the
186 compartment. These processes can be subdivided into discrete, explanatory, independent
187 biophysical sub-processes, which could be individually modeled (*ModelUnit*) or composed
188 (*ModelComposite*). A modular model structure requires making an objective decomposition of the
189 system to avoid coarse granularity models, which limit reusability. A *ModelUnit* should not
190 encapsulate alternative assumptions and formalisms, making it easier to test them. In addition, the
191 management of input and output data, such as data access, logging, and file generation, must be
192 managed separately from the implementation of model component. These design principles foster
193 the reuse of components, which are intended to be integrated and simulated with a large variety of
194 input data formats in different PBM platforms. Moreover, to emphasis modularity, the temporal
195 integration loop must be removed from the model process implementation. This makes it possible to
196 reuse the same process with different modeling formalisms or simulation frameworks that manage
197 temporal dynamics of the simulation differently (e.g. different numerical integration techniques).

198 Crop2ML provides a level of abstraction that enables a shared representation of model
199 components between PBM platforms. A *ModelUnit* is defined with the following descriptive
200 elements (Fig. 2a):

- 201 • a model description;

- 202 • a list of inputs;
- 203 • a list of outputs;
- 204 • an initialization step of the state variables;
- 205 • a link pointing to the source of the model algorithm;
- 206 • a list of usual mathematical functions;
- 207 • a set of unit tests with parameterization shared between modeling platforms.



208
 209 **Fig. 2.** Crop2ML concepts for model specifications: *ModelUnit* (a) and *ModelComposite* (b). "+", one
 210 or more elements; "*", zero or more elements; "?", zero or one element.

211 A *ModelComposite* includes the same elements as a *ModelUnit*. In addition, it contains a list of
 212 *Models* and the links between them. (Fig. 2b). However, if control structures are necessary to express
 213 the behavior of a *ModelComposite*, the *Algorithm* can be explicitly provided.

214 The Crop2ML model specification is based on XML Language. XML is a widely used declarative
 215 metalanguage for describing or structuring data in a portable format with some descriptive elements.
 216 XML format is used in several PBM platforms for template parametrization and model simulation
 217 configuration (e.g. APSIM, BioMA, RECORD, Simplace, *SiriusQuality*). This reinforces our choice on
 218 this format since the transformation between different XML documents or in any language is
 219 relatively straightforward, allows using XML as a bridge between heterogeneous structures and it
 220 facilitates collaborative development. Moreover, the use of XML and a formal description of model
 221 specifications and their associated metadata facilitate machine readability and model exchange. In
 222 the following sections, we describe the concepts of Crop2ML model specifications.

223 2.1.1. Description

224 The core description of a Crop2ML model contains the name of the model, an identifier that
 225 ensures the provenance of the model and a version number (Fig. 3). The identifier of the model is
 226 specified to keep the property of the component. Since PBM are dynamic models, the time step is an
 227 important factor that is specified to allow a multi temporal-scale composition. In addition, other
 228 elements are described to provide rich metadata, including author names and affiliations, citable and
 229 findable references (e.g. doi) and a brief description of the model. The description also includes
 230 usage licenses compatible with the model dependencies.

```

231 <ModelUnit modelid="SQ.EnergyBalance.DiffusionLimitedEvaporation"
      name="DiffusionLimitedEvaporation"
      timestep="1"
      version="1.0">
  <Description>
    <Title>DiffusionLimitedEvaporation Model</Title>
    <Authors>Pierre Martre</Authors>
    <Institution>INRA Montpellier</Institution>
    <Reference> From the wheat simulation model SiriusQuality
      Published in 2009
      doi:http://dx.doi.org/10.1016/j.eja.2006.04.007
    </Reference>
    <Abstract>the evaporation from the diffusion limited soil when the surface has
      dried sufficiently to provide a significant barrier to water vapor diffusion
    </Abstract>
    <URI> http://www1.clermont.inra.fr/siriusquality/?page_id=547 </URI>
  </Description>
  ...
  </ModelUnit>

```

232 **Fig. 3.** Example of a Crop2ML ModelUnit core description.

233 2.1.2. Inputs – Outputs

234 In Crop2ML, a component takes parameter and variable values as inputs and produces variable
 235 values as outputs. A variable is a quantity which is given by the context of the experiment (input
 236 data) or calculated by the model (output data), while the value of a parameter is an input that can be
 237 specified by the modeler within a defined interval. Variables and parameters are distinguished with
 238 *input type* attributes and are categorized with *variable category* and *parameter category* attributes,
 239 respectively (Table 1).

240

Table 1
 Category, definition, and example of variables and parameters in Crop2ML.

Input Type	Category	Definition	Example*
Variable	State	Characterizes the behavior of a component	Leaf area index, weight of a plant part, canopy temperature
	Rate	Defines the change of one state variable	Transpiration rate, leaf growth rate
	Auxiliary	Intermediate variable computed by an auxiliary	Dry matter partitioning, shoot number

		function	
	Exogenous	Driven variables that do not depend on other variables of the system or component	Mean air temperature, wind speed
Parameter	Constant	Absolute constant	Boltzmann constant
	Soil	Soil parameter	N mineralization constant, maximum rootable soil depth
	Species	Crop parameter with fixed value for a species	Maximum respiration rate
	Genotypic	Crop parameter that can take different values for different genotypes (cultivars)	Phyllochron, grain filling duration

241 Crop2ML currently supports four basic types: integer, double, strings and logical. It also supports
 242 two collection types: lists and arrays, which contain a sequence of elements of basic types. They are
 243 explicitly specified in a *datatype* attribute, similar to the VarInfo type (Donatelli & Rizzoli, 2008). It
 244 also provides a common representation of date/time. The domain of validity of each variable is
 245 specified by *min* and *max* attributes. A measurement unit can also be associated to the variables and
 246 parameters. Fig. 4 gives an example of inputs and outputs specifications.

```

247 <Inputs>
    <Input name="deficitOnTopLayers" description="deficit On TopLayers"
      variablecategory="auxiliary" datatype="DOUBLE" default="5341"
      min="0" max="10000" unit="g/(m**2*d)" inputtype="variable"/>
    <Input name="soilDiffusionConstant" description="soil Diffusion Constant"
      parametercategory="soil" datatype="DOUBLE" default="4.2" min="0" max="10"
      unit="dimensionless" inputtype="parameter"/>
  </Inputs>
  <Outputs>
    <Output name="diffusionLimitedEvaporation"
      description="the evaporation from the diffusion limited soil "
      variablecategory="rate" datatype="DOUBLE" min="0" max="5000"
      unit="g/(m**2*d)"/>
  </Outputs>

```

248 **Fig. 4.** Example of input and output specifications of a Crop2ML model.

249 2.1.3. Initialization

250 State variables of Crop2ML *ModelUnits* and *ModelComposites* are initialized at the start of a
 251 simulation and are specified with an *Initialization* element. This element is optional, and the default
 252 values of state variables are used if it is omitted. Initialization may also be a function that assigns
 253 initial values to state variables. In this case, the *Initialization* element contains the path to the source
 254 code of the initialization function.

255 2.1.4. Algorithm

256 *Algorithm* elements link the model specifications with the model algorithm (Fig. 5). A model
 257 algorithm describes the behavior of a component in terms of a sequence of inputs, successive rules

258 or actions, conditions and a flow of instructions from inputs to outputs including mathematical
259 expressions. A model algorithm can be implemented in different programming languages. However,
260 Crop2ML proposes to encode the model algorithm in a shared language, CyML (Midingoyi et al.,
261 2020). The CyML source code is the common representation for model algorithm shared by the
262 supported languages and platforms (see Section 2.2.).

```
263 <Algorithm language="Cym1" filename="algo/pyx/diffusionlimitedevaporation.pyx" />
```

264 **Fig. 5.** Example of a link to an algorithm file.

265 2.1.5. *Function*

266 A function is a utility routine that can be called from the model algorithm or from other
267 functions. It reduces the code length and improves the readability of the encoded algorithm. If a
268 model needs an external function, this function must be declared in the model specification by
269 referencing the path where the function is implemented. A function can also be used for model
270 adaptations such as temporal aggregation or integration, unit conversion to link model components
271 without changing their algorithms. Crop2ML provides a shared library of mathematical functions in
272 different languages such as standard functions, interpolation, or upper and lower bound functions.
273 Modelers can use these functions in their own algorithm, implemented in the CyML language.

274 2.1.6. *Parameter sets and test sets*

275 A Crop2ML model specification includes one or more sets of model parameterizations used for
 276 different unit tests (Fig. 6). A parameterization is a set of values assigned to an input parameter of a
 277 model. It is described by a name and a description. A unit test in Crop2ML is described in the *Testsets*
 278 element and allows comparing estimated and expected outputs values. Several unit tests can be
 279 specified. They are described by their *name*, their *description* and the name of *parameters set*
 280 associated to them. Each test provides a list of values assigned to each variable and the expected
 281 values of the model outputs. A numerical precision could be associated with the *output* of the test to
 282 check its validity.

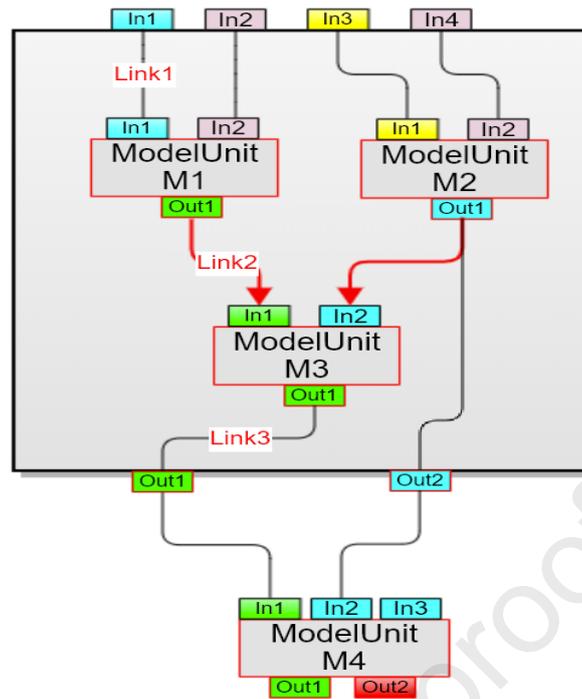
```

<Parametersets>
  <Parameterset name="set1" description="some values in there" >
    <Param name="soilDiffusionConstant">4.2</Param>
  </Parameterset>
</Parametersets>
<Testsets>
  <Testset name="first" parameterset = "set1" description="some values in there" >
    <Test name ="test1">
      <InputValue name="deficitOnTopLayers">5341</InputValue>
      <OutputValue name="diffusionLimitedEvaporation" precision ="3">6605.505</OutputValue>
    </Test>
  </Testset>
</Testsets>
  
```

283
 284 **Fig. 6.** Example of parameterization and unit tests in Crop2ML.

285 2.1.7. Model links

286 Model links are specified in a *ModelComposite* and depict how *ModelUnits* or *ModelComposites*
 287 are interconnected. A *ModelComposite* is a port graph (Andrei & Kirchner, 2009) that defines a
 288 dataflow where nodes are *ModelUnits*, and ports are inputs and outputs of the *ModelUnits*. Edges
 289 are oriented links connecting output ports of a source *ModelUnit* to the input ports of a target
 290 *ModelUnit* (Fig. 7). Three types of links must be specified: *InternalLink* is the connection between an
 291 input of one sub-model and the output of another sub-model, *InputLink* is the connection between
 292 an input port of a sub-model and an input port of the composite model, and *OutputLink* is the
 293 connection between a *ModelUnit* or *ModelComposite* output port, that can be either a *ModelUnit* or
 294 *ModelComposite*, and a *ModelComposite* output port. These connections show the hierarchical
 295 structure of a *ModelComposite*. This modeling approach enhances reusability and has been used with
 296 success (Wyatt, 1990).



297

298 **Fig. 7.** Graph of a *ModelComposite*. Three *ModelUnits* (M1 to M3) are connected to form a first level
 299 of composition, which is linked to a fourth *ModelUnit* (M4). Link1 is an *InputLink*, Link2 is an
 300 *InternalLink*, and Link3 is an *OutputLink*. *OutputLink* is specified to clearly define the outputs of the
 301 *ModelComposite*, which necessarily include all state variables. Each model component has input
 302 ports In1, In2, ..., and output ports Out1, Out2, ..., where 1, 2, ..., are internal (local) port numbers.

303 2.2. *CyML: the common modelling language of biophysical processes in crop models*

304 We defined a set of common features resulting from the intersection of the programming
305 languages supported by PBM platforms to propose a shared modelling language. A design choice was
306 to define a subset of an existing language that can provide these common features. We needed a
307 widely used high-level language with a low learning curve so that modelers with basic programming
308 skills could efficiently use it. The transformation of a language with dynamic typing can make code
309 transformation into programming languages with static typing ambiguous. Therefore, we choose
310 Cython, a high-level language that combines the expressive power of Python language with explicit
311 type declaration of C language (Behnel et al., 2011). It is compiled directly in efficient C code, which
312 improves runtime speed and makes it possible to interact with C, C++ and Fortran source code.
313 However, not all Cython syntax can be directly transformed in all target languages. For instance, the
314 yield statement and anonymous functions are not supported by Fortran. Therefore, we defined CyML
315 (Cython Meta Language), a sub-set of Cython to address the implementation of the model algorithm
316 (Midingoyi et al., 2020).

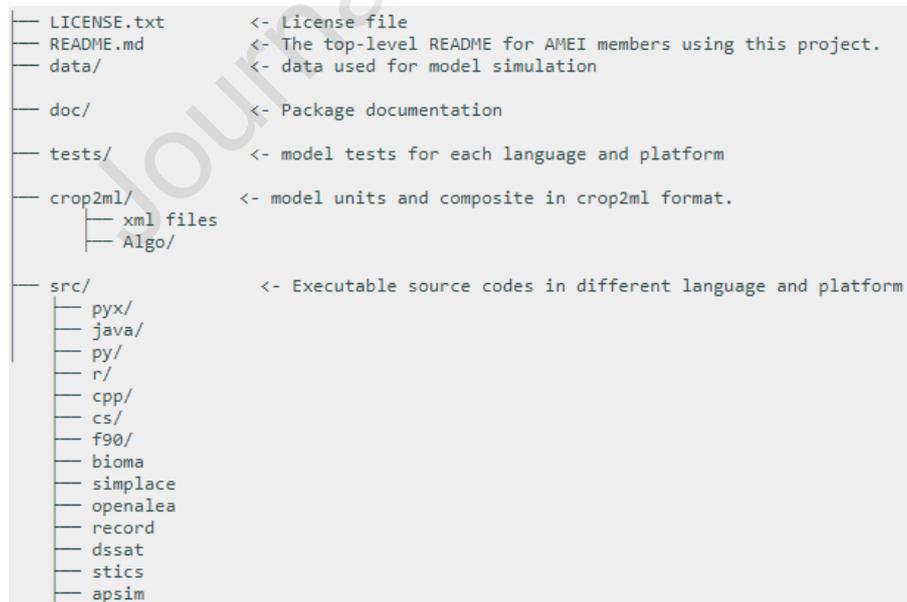
317 We use CyML as a pivot language between various platform languages, which can be mapped to
318 their syntax and semantics. The structure and syntax of CyML, as well as its transformation system to
319 various languages and platforms is detailed in Midingoyi *et al.*, (2020). In brief, CyML supports
320 datatypes defined in the model specification and provides standard mathematical functions and
321 operators. In addition to local variable declaration and assignment statements, control structures are
322 used in the flow of instructions described by the encoded algorithms. These include conditional
323 statements (if, elif and else) to check if a condition is satisfied before addressing part of an algorithm,
324 sequential statement (for loop) with an incremental index on a data collection, and a repetitive
325 statement (while) used to repeat part of an algorithm while a condition is satisfied. These structures
326 can be nested. To support modular designs and the reuse of *ModelUnits* and functions, CyML
327 provides import mechanisms, which assumes that imported *ModelUnits* or functions are referenced.

328 Crop2ML framework provides a source-to-source transformation system (CyMLT) which converts
329 CyML source code into procedural (Fortran, Python, C++), object-oriented (Java, C#, C++, Python) and
330 scripting or functional (R, Python) languages (Midingoyi et al., 2020). CyMLT implementation relies
331 on the transformation of the abstract syntax tree (AST) generated from the syntax analysis of the
332 CyML code. The AST is transformed to a self-contained representation of the source code called
333 Abstract Semantic Graph, which is independent of the source language. CyMLT proposes a unique
334 approach to transform the Abstract Semantic Graph into readable source code in many different
335 languages. The generated code is independent from the transformation system and can be run

336 outside the Crop2ML framework. The transformation system integrates model documentation based
 337 on the model specification into generated code.

338 2.3. Crop2ML model package

339 In the context of large projects and collaborative work, it is useful to define some requirements
 340 or standards to facilitate common exchange. Crop2ML provides a logical, standardized but flexible
 341 support to facilitate model sharing between modeling platforms through the definition of a directory
 342 structure (Fig. 8). This template includes a folder that contains model description and associated
 343 algorithms, a repository of source code for each language and modeling platforms. It also includes a
 344 folder containing input data for a *ModelComposite* simulation, and a folder containing the unit tests.
 345 To save time and avoid omission of mandatory files or folders during package creation, we created a
 346 cookiecutter (Roy, 2017) template that automatically generates Crop2ML package templates
 347 (<https://crop2ml.readthedocs.io/en/latest/user/package.html>). It increases model reusability by
 348 automatically generating a project that follows shared guidelines. Any *ModelUnit* or *ModelComposite*
 349 can be extracted as a stand-alone model from an existing package, tested, reused, or integrated in
 350 other *ModelComposite* or package. The notion of package-dependency increases the modularity of
 351 Crop2ML and avoids model duplicity.



352

353 **Fig. 8.** Tree view of the structure of a Crop2ML model component package.

354 2.4. Crop2ML model lifecycle management

355 Crop2ML aims at collaborative model development that supports the entire model lifecycle,
356 including model creation, editing, verification, validation, transformation, composition, and
357 documentation. Therefore, we developed tools and services to support all the steps of a Crop2ML
358 model lifecycle.

359 2.4.1. Model analysis

360 Crop2ML models conform to a specific Document Type Definition (DTD) that describes Crop2ML
361 concepts. Model analysis verifies if the model specifications are a well-formed XML document
362 validated by Crop2ML DTD. The analysis of a *ModelComposite* consists of checking model
363 composability through port datatypes and units. Most XML editors can check the validity of an XML
364 document against a DTD but the Crop2ML software environment (see Section 3.2) ensures this.

365 2.4.2. Model validation

366 Crop2ML model components can be validated by executing unit tests. It consists of using the
367 parameter and variable values from the model specification to produce unit tests in different
368 languages. Unit tests are generated in Jupyter notebook format, a document format for publishing
369 source codes and reproducible computational workflows that could be executed in the appropriate
370 kernel in Crop2ML software environment. This format is useful for code and documentation
371 publishing and real-time collaboration when running on a remote server (Kluyver et al., 2016). Unit
372 tests may also be associated with a model publication.

373 2.4.3. Model transformation

374 The success of Crop2ML model reuse through a white box approach comes from its ability to
375 generate model components that conform to platform requirements. The transformation of a model
376 component from a platform to another one goes through Crop2ML model representation. It relies
377 on a system of transformation to and from Crop2ML and the platforms.

378 For some PBM platforms, meta-information of model components are described inside their
379 implementation as documentation. For other platforms meta-information are encoded in a textual or
380 visual programming language. CyMLT generates from Crop2ML model either appropriate
381 documentation or variables and parameters specifications based on the artifacts of the target
382 platforms. In addition, CyMLT generates model component algorithms in various languages. Given a
383 model component provided by a platform, meta-information are extracted by identifying Crop2ML

384 concepts inside the component to generate Crop2ML model meta-information. Moreover,
385 algorithms in CyML are produced to obtain a complete Crop2ML model.

386 2.4.4. Model documentation

387 Sharing model knowledge requires detailed information on the model. Crop2ML generates
388 model documentation from the model specification. From the relationships between the *ModelUnits*
389 of a *ModelComposite*, the diagram flow of the *ModelComposite* is generated. It may constitute part
390 of the model documentation and gives a first description of the model component. This allows
391 groups of modelers to understand the model structure and evaluate the component.

392 3. Crop2ML software environment and tools

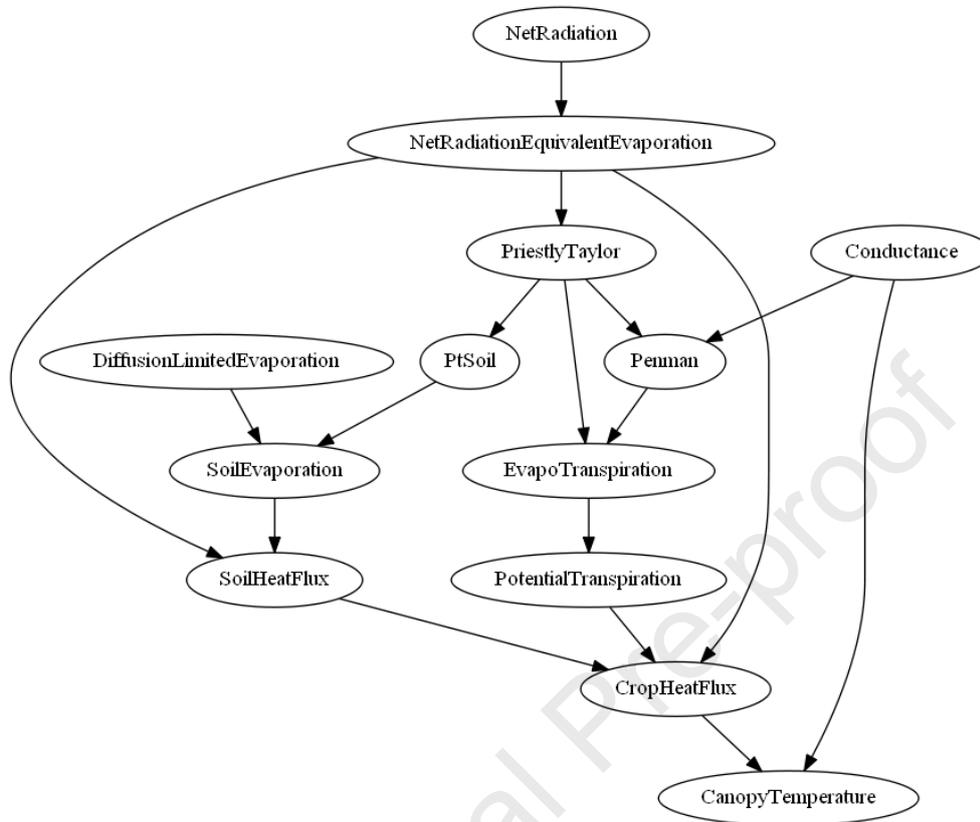
393 3.1. PyCrop2ML: a Python library for Crop2ML

394 Pycrop2ML is an open, modular, and extensible library developed in Python that implements all
395 the steps of Crop2ML model lifecycle. It is designed to support the current Crop2ML model
396 specifications but can easily be adapted to support future versions. Pycrop2ML can be integrated
397 into other software projects as a plug-in. It allows:

- 398 • Verifying a Crop2ML model: This is ensured through a model parser based on the Crop2ML DTD.
- 399 • Transforming a Crop2ML *ModelUnit* to source code: PyCrop2ML integrates CyMLT that
400 generates model components that conform to PBM platform requirements.
- 401 • Transforming a CyML source code to various languages: Regardless of Crop2ML model
402 specifications, any CyML source code can also be transformed into the target languages. This
403 source code can be used as auxiliary functions for Crop2ML model development.
- 404 • Transforming source code to Jupyter notebook format: Each *ModelUnit* source code generated
405 can be translated as a cell of Jupyter notebook, as well as, each unit test, allowing its execution
406 in Crop2ML JupyterLab environment.
- 407 • Transforming a Crop2ML *ModelComposite*: A Crop2ML *ModelComposite* provided as a directed
408 graph can be transformed to source code as a sequential order of the submodels.
- 409 • Visualizing a *ModelComposite*: Pycrop2ML provides a function to visualize a *ModelComposite*
410 with the links between *ModelUnits* (Fig. 9).

411 PyCrop2ML is written in Python and can be executed via a command-line interface, inputting
412 either a Crop2ML package or CyML source code, as well as the target language or platform for
413 transformation. Users with no knowledge of the Python language can easily run PyCrop2ML via the

414 command line. The PyCrop2ML library incorporates three crop model components as model
 415 examples that can be used to test the different functionalities.



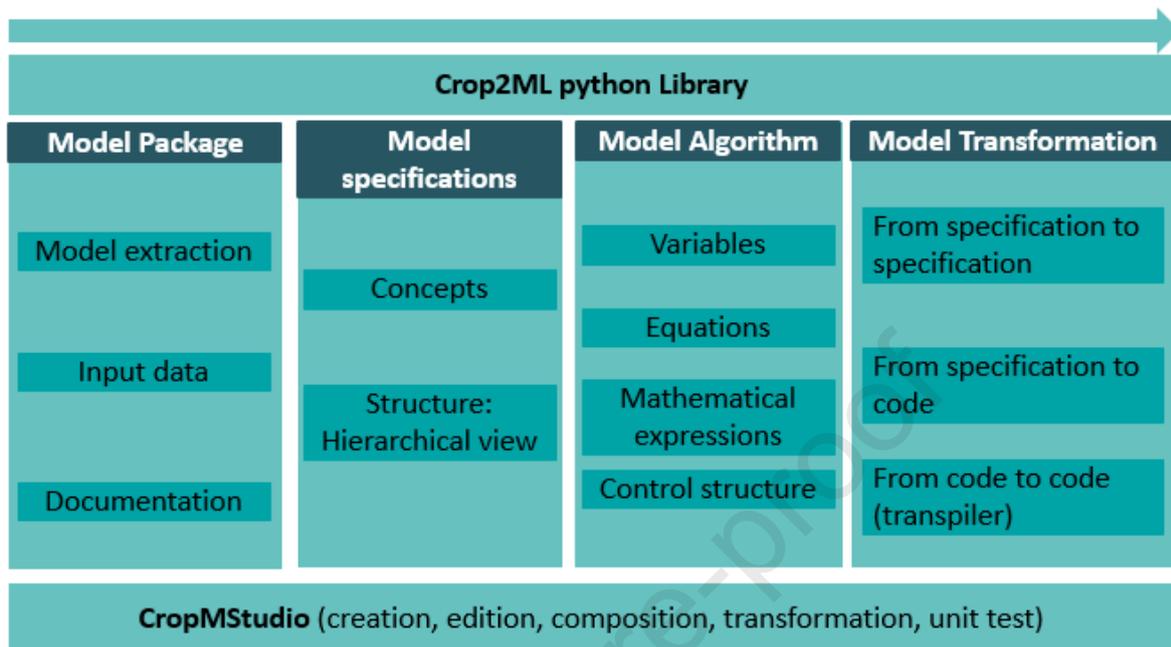
416

417 **Fig. 9.** Visualization of energy balance *ModelComposite* provided from SiriusQuality wheat model
 418 developed with the BiOMA platform. Ellipses are *ModelUnits* and arrows represent the link between
 419 two *ModelUnits*

420 3.2. CropMStudio: A JupyterLab environment for Crop2ML model life cycle management

421 Crop2ML model specifications can be created or edited using any XML editor. However, to fulfil
 422 our objective of collaborative model development accessible to modelers with no specific
 423 programming skills, we developed a user-friendly interface based on the PyCrop2ML package to
 424 manage the lifecycle of Crop2ML model components (Fig. 10). Since Crop2ML models are
 425 transformed in different languages, it is useful to execute the unit tests in a single environment. Our
 426 solution, named CropMStudio, uses the JupyterLab environment (<https://jupyterlab.readthedocs.io>),
 427 an open-source web application that allows working with code in different languages through
 428 different language backends kernels. We installed Python, Java, C#, C++, R and Fortran kernels to
 429 execute *ModelUnit* tests. The current version of CropMStudio can be accessed through a web
 430 browser and run locally like a desktop application. Another motivation to use JupyterLab is to make

431 publication results reproducible in a shared environment based on the capacity to produce
 432 interactive and readable code documents (Kluyver et al., 2016).



433
 434 **Fig. 10.** Schematic representation of the Crop2ML framework showing Crop2ML model lifecycle from
 435 the creation of a package to model transformation.

436 4. Interoperability between various simulation platforms

437 The interoperability between simulation platforms is based on two transformation processes
 438 (import and export) via Crop2ML. The import process consists of transforming any platform model
 439 component to Crop2ML model. The export process consists of transforming Crop2ML models to any
 440 platform. Detailed descriptions of the import/export mechanisms in five widely used platforms with
 441 different architectures (BioMA, DSSAT, Record, OpenAlea, SIMPLACE) are provided in Supplementary
 442 data (Appendix C). Table 2 summarizes the interoperability of model components between these
 443 platforms. Platforms are based on various programming languages, which requires the definition of
 444 transformation rules between CyML and various languages including C# (BioMA), Java (Simplace),
 445 C++ (Record), Python (OpenAlea) and Fortran (DSSAT) in both directions. We identified the levels of
 446 granularity of modeling processes that correspond to Crop2ML concepts such as *ModelUnit* and
 447 *ModelComposite* in each platform. We also considered how documentation or model specifications
 448 are described in these platforms.

449 The export process, from Crop2ML to platforms, is automatically done in BioMA, OpenAlea and
 450 Simplace. The modularity principle in BioMA matches Crop2ML, which allows associating simple and

451 composite BioMA strategies with Crop2ML *ModelUnit* and *ModelComposite*, respectively. Moreover,
452 all the Crop2ML elements are well translated into the VarInfo type attributes (Donatelli & Rizzoli,
453 2008), and Crop2ML model algorithms are transformed to a method of a strategy class that takes
454 generated domain classes as inputs. OpenAlea relies on two families of approaches: component-
455 based architecture and scientific workflows. Thus, Crop2ML exports *ModelUnits* as OpenAlea
456 components and *ModelComposite* as OpenAlea workflows. *ModelComposite* can thus be visualized
457 and edited using VisuAlea, the visual programming environment in OpenAlea. Widgets of *ModelUnit*
458 are automatically generated based on the type of inputs that is mapped to an OpenAlea interface.
459 Simplace is based on the concept of software units, called SimComponents as the smallest building
460 blocks that map with *ModelUnits*. *ModelComposite* are converted into a combination of
461 SimComponents (SimComponentGroup). Variables and parameters descriptions are automatically
462 included in the SimComponents descriptive part.

463 In DSSAT and Record the export process is many automatic but some aspects need to be done
464 manually. In DSSAT, Crop2ML transformation system generates a submodule in Fortran 90 for each
465 *ModelUnit*. It also generates a sequence of submodules calls for composite models. One issue that
466 makes this transformation not completely automatic is that Crop2ML does not manage the handling
467 of input and output files. Therefore, it requires to manually add the input and output methods into
468 the generated submodules. The concepts of atomic and coupled models in Record are mapped with
469 those of Crop2ML. Thus, atomic model classes are generated in C++ to correspond to *ModelUnits*.
470 However, the configuration and simulation file (VPZ) representing the *ModelComposite* is manually
471 completed with further information such as the description of simulation result files.

472 The import process (from simulation platforms to Crop2ML) is only partially automatic. Platform
473 tools produce automatically the meta-information in Crop2ML format but algorithms are manually
474 converted into the CyML language that leads to a semi-automatic transformation. A complete
475 automatic transformation would require the implementation of source-to-source transformation
476 from platforms' language into CyML. In BioMA, VarInfo attributes are extracted from BioMA
477 strategies to produce Crop2ML model meta-information. The process of automatically retrieving the
478 *estimate* method to produce model algorithm in CyML is not implemented yet. The description of
479 component in OpenAlea is very generic compared to Crop2ML concepts. Although OpenAlea is
480 mainly built for Functional Structural Plant Modeling (FSPM) application, there is no plant domain
481 specific description associated with inputs and outputs such as units, categories of variables and
482 parameters. Thus, the generation of model description in Crop2ML is partial. It requires further
483 description of components that can be provided in documentation or by extending OpenAlea

484 concepts. Like BioMA, the SimComponent specific descriptors in Simplace allows generating
 485 *ModelUnits* meta-information. The *process* method (algorithm) is currently translated manually in
 486 CyML. Links between the different SimComponents (Unit) stored in the SimComponentGroup
 487 (Composition) are automatically exported to the Crop2ML structure. However, there is a loss of
 488 information since when a ModelUnit is activated or ignored it is not transferred to the Crop2ML
 489 structure. In DSSAT, unlike in the other platforms, the description of physiological processes is
 490 provided as documentation in submodules and it is not fully complete with respect to Crop2ML
 491 specifications. Inputs and outputs variables and their descriptions, units can be clearly identified,
 492 based on systematic platform guidelines. DSSAT submodules contain specific platform variables such
 493 as control variables that need to be removed to produce CyML model algorithms. In Record, as in
 494 DSSAT, there is no explicit specification of a model. The documentation of a model within its
 495 associated C++ class is used to generate partial *ModelUnit* meta-information. The parsing of the VPZ
 496 file, that contains the structure of composite models in Record, is used to generate a
 497 *ModelComposite*. However, it is not possible to represent retroaction loops in Crop2ML as it is done
 498 in Record with coupled models.

Table 2

Import and Export processes between Crop2ML and PBM platforms. A, automatic; P, partially automatic; M, manual.

PBM Platforms	From Crop2ML to PBM platforms				From PBM platforms to Crop2ML			
	A	P	M	To be completed	A	P	M	To be automatized
BioMA	✓				✓			Source transformation from C# to CyML
DSSAT		✓		Integration of I/O			✓	Source transformation from Fortran to CyML Remove I/O More variable description
Record		✓		Complete VPZ file	✓			Source transformation from C++ to CyML More variable description
OpenAlea		✓					✓	Source transformation from Python to CyML More variable description
SIMPLACE		✓					✓	Source transformation from SIMLPACE Java to CyML

499

500 In order to illustrate Crop2ML concepts and transformation results, a phenology and an energy
 501 balance models are used. Phenology, the timing of crop development is the heart of most crop

502 growth models and is an essential component of most crop modeling platforms. The energy balance
503 model involves interconnected components that allows estimating canopy temperature,
504 evapotranspiration, and heat transfer between the canopy and the air. These processes are
505 implemented as BioMA standalone components (Manceau & Martre, 2018) of the wheat PBM
506 *SiriusQuality* (He et al., 2012; Martre et al., 2006). The two components were converted into
507 Crop2ML packages, and then automatically translated into different languages and model
508 components that conform to different PBM platforms. These packages are presented in Appendixes
509 A and B. In Table 3 we illustrate how to represent a parameter and an algorithm in a Crop2ML Model
510 Unit and its translation with CyMLT in Record, BioMA, and DSSAT. The implementations of the model
511 differ between the platforms. For instance, DSSAT defines a subroutine with all the variables as
512 argument, Record defines a class method (compute) with the variables as attributes of the class and
513 uses specific operator “()” to manage temporal variables, while BioMA defines a class method
514 (CalculateModel) that takes as argument data structures implementing each category of variables
515 (state, rate, auxiliary, exogenous). The aim is to provide to the platforms alternative model
516 components that could easily replace their corresponding components to analyze the effects of new
517 hypotheses into their modeling solutions.

518 The sequence of ModelUnits that compose a Crop2ML ModelComposite is formally modeled as a
519 directed acyclic graph. This means that there is no feedback loop or retroaction at a given time step,
520 instead they are usually represented by a cycle in the *ModelComposite*. Alternatively, a state variable
521 can be defined explicitly as two variables with respect to the current and the previous time. Thus, a
522 composite model may take as input a state variable at previous time and a state variable at current
523 time as output, making implicitly a loop with respect to time advance. Another way to represent
524 feedback inside a time step is to associate an explicit algorithm to the ModelComposite that defines
525 how to run it. However, this feature is not supported by two simulation platforms (OpenAlea and
526 RECORD).

Table 3. Declaration of the inputs and algorithm of a Crop2ML ModelUnit of the Penman-Monteith evapotranspiration model and the equivalent source code generated by CyMLT for Record, BioMA, and DSSAT. The declaration of a single variable is given as an example. 23

Framework or platform / language	Variable declaration	Algorithm
Crop2ML / CyML	<pre><input name="rhoDensityAir" description="Density of air" parametercategory="constant" datatype="DOUBLE" min="0" max="10000" default="1.225" unit="kg/m**3" inputtype="parameter"/></pre>	<pre>def model_penman(float evapoTranspirationPriestlyTaylor=449.367, float hslope=0.584, float VPDair=2.19, float psychrometricConstant=0.66, float Alpha=1.5, float lambdaV=2.454, float rhoDensityAir=1.225, float specificHeatCapacityAir=0.00101, float conductance=598.685): ---- ---- cdef float evapoTranspirationPenman evapoTranspirationPenman = evapoTranspirationPriestlyTaylor / Alpha + 1000.0 * ((rhoDensityAir * specificHeatCapacityAir * VPDair * conductance) / (lambdaV * (hslope + psychrometricConstant))) return evapoTranspirationPenman</pre>
Record / C++	<pre>//Model parameters /** * @brief Density of air (kg/m**3) */ double rhoDensityAir; //...</pre>	<pre>virtual void compute(const vd::Time& /*time*/) { evapoTranspirationPenman = evapoTranspirationPriestlyTaylor() / Alpha + (1000.0d * (rhoDensityAir * specificHeatCapacityAir * VPDair() * conductance() / (lambdaV * (hslope() + psychrometricConstant)))); }</pre>
BioMA / C#	<pre>VarInfo v4 = new VarInfo(); v4.DefaultValue = 1.225; v4.Description = "Density of air"; v4.Id = 0; v4.MaxValue = 10000; v4.MinValue = 0; v4.Name = "rhoDensityAir"; v4.Size = 1; v4.Units = "kg/m**3";</pre>	<pre>private void CalculateModel(SiriusQualityEnergybalance.EnergybalanceState s, SiriusQualityEnergybalance.EnergybalanceState s1, SiriusQualityEnergybalance.EnergybalanceRate r, SiriusQualityEnergybalance.EnergybalanceAuxiliary a, SiriusQualityEnergybalance.EnergybalanceExogenous ex) { double evapoTranspirationPriestlyTaylor = r.evapoTranspirationPriestlyTaylor; double hslope = a.hslope; double VPDair = a.VPDair; double conductance = s.conductance; double evapoTranspirationPenman; evapoTranspirationPenman = evapoTranspirationPriestlyTaylor / Alpha + (1000.0d * (rhoDensityAir * specificHeatCapacityAir * VPDair * conductance / (lambdaV * (hslope + psychrometricConstant)))); r.evapoTranspirationPenman = evapoTranspirationPenman; }</pre>

DSSAT / Fortran	!	* name: rhoDensityAir	SUBROUTINE model_penman(evapoTranspirationPriestlyTaylor, ...)
	!	** description : Density of air	REAL, INTENT(IN) :: evapoTranspirationPriestlyTaylor
	!	** datatype : DOUBLE	...
	!	** min : 0	evapoTranspirationPenman = evapoTranspirationPriestlyTaylor / Alpha + &
	!	** max : 10000	(1000.0 * (rhoDensityAir * specificHeatCapacityAir * VPDair * &
	!	** default : 1.225	conductance / (lambdaV * (hslope + psychrometricConstant))))
	!	** unit : kg/m**3	END SUBROUTINE model_penman

528 5. Discussion

529 The Crop2ML framework enables a user to exchange and reuse biophysical components between
530 various PBM platforms through shared declarative specifications. The use of a minimal language to
531 describe the model algorithm once and the transformation system facilitates reuse of models'
532 components. *ModelUnits* and *ModelComposite* can be accessed and composed following a white box
533 approach. Therefore, the Crop2ML approach greatly increases the ability of modelers to share their
534 algorithms. The protocol will allow modelers to borrow components easily and will facilitate their
535 intercomparison and improvement in different PBM platforms.

536 5.1. How does Crop2ML address model reuse compared to other initiatives?

537 Some initiatives addressed model reuse by providing multi-scale and multi-language integrative
538 frameworks such as *Crops in silico* (Marshall-Colon et al., 2017) the Open Modeling Foundation
539 OpenMI (Buahin & Horsburgh, 2018). These frameworks can compose and simulate heterogeneous
540 models provided by different frameworks through a communication interface. The model
541 components are often wrapped and are represented as black-box components. All state variables are
542 not always exposed as model outputs, which may limit their integration in an existing modeling
543 solution. Therefore, these frameworks enhance model reuse in their own environment but they do
544 not address reusability with other PBM platforms. Many existing PBM platforms do not support the
545 coupling of models written in multiple languages (e.g. BioMA, APSIM next generation).

546 Donatelli and Rizzoli (2008) proposed a design pattern for platform-independent model
547 components to enhance modularity and to facilitate model reuse in several PBM platforms via simple
548 wrappers. However, this approach fixes the structure of the components. The lack of specification or
549 meta-information makes the reuse of model components between platforms difficult. Even in
550 component-based systems, explicit information about the component itself and its inputs and
551 outputs (types, units and boundary conditions) are required to ensure a syntactic composability and
552 to meet the specificities of the platforms. Moreover, the knowledge of the structure underlying the
553 source code of a component is also required to systematically extract model information (variables
554 and algorithms) for their transformation and integration in different platforms. We thus argue that
555 model component reuse is improved if it is supported by model specification. Crop2ML defines an
556 abstract representation of model design shared by PBM platforms through some shared concepts
557 enriching or extending those proposed by Athanasiadis et al. (2011) with other attributes and a
558 formal and shared description of unit tests. We included unit tests in Crop2ML specifications to
559 ensure model transformation validation and some imperative constructs for model dynamics.

560 Several initiatives have used declarative modeling to describe model specifications and address
561 model reuse issues. The approach proposed by Villa et al. (2006) is similar to ours but it is limited to
562 models where the dynamics of the modeled processes is represented by simple mathematical
563 expressions without control structures, which does not match crop modeling context. Hucka et al.
564 (2003) used MathML (Ausbrooks et al., 2003) to express interactions between variables through
565 mathematical formalisms well defined in the systems biology community. This approach is similar to
566 that of Rizzoli et al. (2008) and is useful when processes are governed by differential equations.
567 However, in the PBM context, simulation platforms use algorithms to describe processes rather than
568 mathematical formalisms with differential equations. Moreover, in PBM, variables that drive the
569 system are temporal series that change the behavior of the system at discrete time. This does not
570 require finding a general solution of recurrent equations used in crop models but rather estimating at
571 each time step the state variables of the system.

572 Automated model transformation is a core aspect of model-driven development (Cuadrado &
573 Molina, 2007). It uses Model-Driven Engineering (MDE) principles based on metamodeling concepts.
574 Crop2ML is in line with MDE. It defines structured concepts representing its metamodel, with which
575 all Crop2ML models are conform, and a model transformation to generate PBM platforms'
576 components. Model Driven Architecture (Brown, 2004) is a framework of MDE that provides several
577 standard languages (e.g., ATL, QVT, ETL, Henshin, VIATRA, and Stratego) for model transformation
578 (Jouault and Kurtev, 2006; Kurtev et al., 2006). Crop2ML is based on a transformation process
579 through a set of refinement of models and code with some extensible rules defined as templates in
580 Python. Most MDE approaches allow model to model or model to code transformation where a
581 model represents the specification in our case. However, the use of transformation language
582 standards was inappropriate in our context to unify transformation process towards many languages
583 with different paradigms (Bucchiarone et al., 2020). Crop2ML produces code in a target language but
584 also adapts the code to fit with PBM platform specificities. To our knowledge, model transformation
585 languages in MDE do not support code generation in multiple languages with extended features in
586 the same environment.

587 5.2. *Connecting Crop2ML to PBM platforms*

588 Given that Crop2ML datatypes do not handle complex data structures other than arrays and lists,
589 some compromises or transformation should be made to the import-export process on the platform
590 side with respect to handling other data structures used in platforms. As an example, BioMA provides
591 the Dictionary data type that is a set of keys associated with values to represent either input or
592 output variables. This data type is not handled by Crop2ML, and by most PBM platforms. As an

593 alternative, Dictionaries can be expressed in Crop2ML as two list datatype variables that represent
594 keys and values of the dictionary.

595 The simulation algorithm defining the feedback loop is explicitly described as control flow in
596 some platforms (e.g. BioMA) but this is not the case in other platforms (e.g. Record, where the VPZ
597 file representing the simulation model file is handled by the simulation engine VLE). Different
598 simulation engines are based on different models of computation used by the platforms such as
599 dataflow (e.g. OpenAlea), DEVS simulation (e.g. Record), control flow (e.g. BioMA, DSSAT, and
600 Simplace). These models of computation are used to coordinate the execution of the model. The
601 current version of Crop2ML framework does not take into account the specificities of simulation
602 engines and addresses components which can be sequentially composed.

603 The Crop2ML transformation system is designed to support the specificities of the target PBM
604 platforms. However, the semantic of a Crop2ML model is based on shared concepts to describe at a
605 high level a biophysical process by a discrete-time model. There is no semantic reason to support the
606 description of each instance of the concepts. For example, since we have not defined a convention to
607 name process variables, the integration of a Crop2ML component into a PBM modeling solution
608 requires adapting the name of its variables. In the future, we could annotate Crop2ML models to add
609 semantic information to make semantic links between any Crop2ML model variables or parameters
610 with those of model components of PBM platforms. This will also allow a semantic composability of
611 Crop2ML models instead of a syntactic composability that analyzes whether the pair of variables to
612 be linked are compatible. However, this would require the crop modeling community to agree on
613 shared semantics and ontologies of crop model variables and parameter representations. Until now
614 this has been a real challenge as the crop modeling community has not been too keen on adapting
615 standards (White et al., 2013). In addition, to facilitate the exchange and reuse of model
616 components, semantic descriptions of model variables and parameters would facilitate the linking of
617 crop models to plant phenomics data (Neveu et al., 2018).

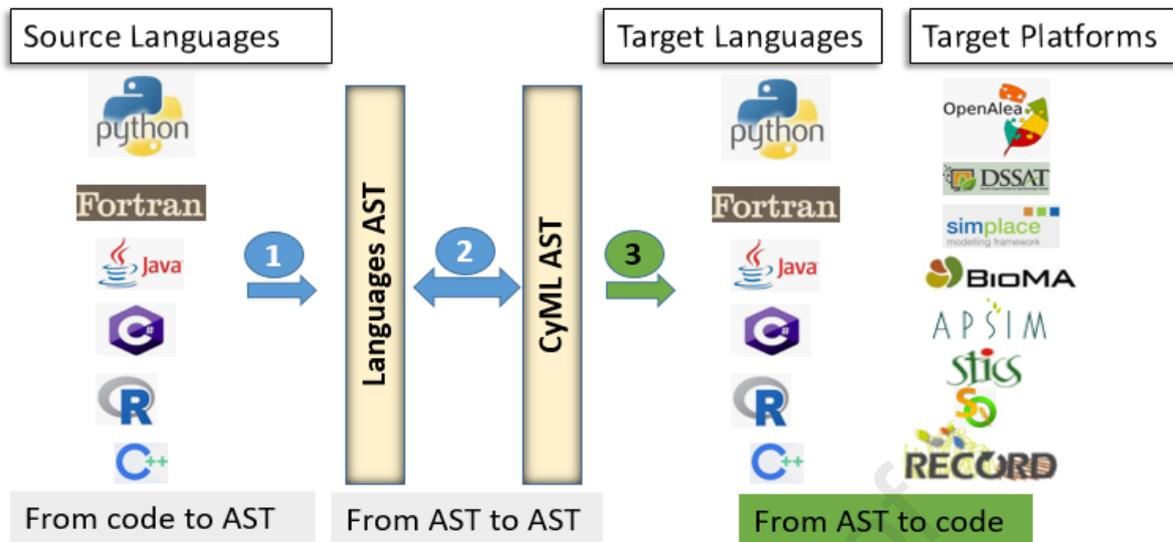
618 We were able to achieve fully-automatic export of Crop2ML model to several PBM platforms.
619 The import process into Crop2ML is more mixed regarding the overall differences between PBM
620 platforms. It is much easier to start with concepts shared and reused by PBM platforms than to start
621 from divergent views of model representations to achieve a particular result. Some PBM platforms
622 need to extend their concepts for model specification or to provide a rich model documentation in
623 order to produce complete Crop2ML model specifications. This reveals the need of a good level of
624 abstraction to represent a model in various PBM platforms. The higher the level of abstraction, the
625 further the description moves away from the platforms and the less easy it is to understand. On the
626 other hand, if the level of abstraction is too low, it is not always possible to represent all features of
627 the models present in the platforms.

628 5.3. *Future developments*

629 A common model repository infrastructure is essential for efficient model exchange (Glont et al.,
630 2018; Lloyd et al., 2008). Currently, Crop2ML model components are stored in Github repositories.
631 We aim to provide a Crop2ML model repository to store models in a shared format to make them
632 easily accessible and reusable by the plant and crop modeling community. This repository should aim
633 at hosting alternative biophysical processes. It will help modelers to operate on multiple model
634 components, compare processes, or evaluate the impact of the integration of alternative models of
635 biophysical processes in crop models. The success of the Crop2ML repository requires that the
636 community gives access to their models by feeding the repository, which will be curated by the AMEI
637 consortium to avoid error propagation.

638 Crop2ML has some limitations, which can be addressed in the next versions, either by extending
639 the model specifications with shared concepts or by adapting the target PBM platforms to Crop2ML
640 specification and language. It is an ongoing, long-term activity, to satisfy platform requirements and
641 facilitate Crop2ML model life-cycle management to make Crop2ML a standard for the plant and crop
642 modeling community.

643 The transformation of a model component of a PBM platform into a Crop2ML package requires
644 rewriting the model algorithms in the CyML language. This limit is currently being addressed by
645 extending the CyML transpiler to a bidirectional transpiler. Thereby, PBM platforms could provide
646 model algorithms in the language they use and the extended CyMLT will transform them in CyML and
647 target languages used by other PBM platforms (Fig. 11). This is a two-step process. First, the model
648 algorithms in the language of the source PBM platform will be parsed and an AST will be generated.
649 Second, the rules for transforming this AST into the CyML AST will be applied. The second step will
650 reuse the CyML transformation tool developed by Midingoyi *et al.* (2020) to produce model
651 algorithms compatible with other languages and PBM platforms.



652

653 **Fig. 11.** Schema illustrating CyML transformation extensibility to support bidirectional source
 654 transformation.

655

656 Other future developments of Crop2ML include:

- 657 • Enhance Crop2ML model repositories with model annotation to link publications to models for
 658 reproducibility;
- 659 • Add unit checks and conversions in Crop2ML to improve model validity;
- 660 • Define a methodology to link Crop2ML with plant structure representation for multiscale viewing
 661 and analysis;
- 662 • Define and implement an ontology of crop model variable and parameters to allow better
 663 Crop2ML model interpretation and improve transformation between PBM platforms and the
 664 integration of model component in complex modeling solutions.
- 665 • Extend Crop2MLab prototype by including bidirectional transformation and the creation of a web
 666 interface on a remote server in order to give users the possibility to handle Crop2ML model
 667 lifecycle without local installation.

668 6. Conclusion

669 At the interface between modeling and software engineering, this paper addresses plant and
 670 crop model component reuse by proposing the Crop2ML framework. Despite all the differences
 671 between PBM platforms, some common features can be identified that enabled model
 672 representation regardless of the platforms' specificities. Crop2ML provides structured concepts to
 673 support the definition of *ModelUnit* and *ModelComposite* and allows their transformation to make
 674 them compatible with PBM platforms at implementation level. Therefore, Crop2ML defines a new

675 unified crop model representation that considers the abstraction of PBM component features in
676 several PBM platforms. Moreover, Crop2ML uses a domain specific language to describe biophysical
677 processes and auxiliary functions to represent model dynamics based on a subset of the Cython
678 language, which can then be automatically transformed into different target languages. Crop2ML
679 proposes an open framework to manage all the steps of model lifecycle.

680 **Declaration of competing interest**

681 The authors declare that they have no known competing financial interests or personal
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689 BioMA component of *SiriusQuality* in Crop2ML.

690 **Appendix C. Supplementary data**

691 Supplementary data to this article can be found online at _____

692 **Appendix A. Crop2ML Engery balance component**

693 <https://doi.org/10.5281/zenodo.4292231>

694 **Appendix B. Crop2ML Phenology component**

695 <https://doi.org/10.5281/zenodo.4292245>

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Highlights

- An open framework is developed (Crop2ML) to support the development, exchange, and reuse of crop model components.
- Crop2ML provides a shared format to describe the specifications of model units and their composition.
- Crop2ML generates model components considering crop simulation platforms artifacts.
- Crop2ML model life cycle can be managed using a user-friendly JupyterLab interface.

Crop2ML can be used to import and export model components into many different crop simulation platforms.

Declaration of interests

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:

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