

# <sup>1</sup> FAME-Io: Configuration tools for complex agent-based simulations

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## Software

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## <sup>7</sup> Summary

We present FAME-Io, a Python package designed to help users and creators of agent-based simulation models (ABM) better manage the preparation and processing of their input and output datasets. The package was built with the needs of researchers in mind. FAME-Io was specifically developed to interface with the open framework FAME<sup>1</sup> and is published under the open Apache-2.0 licence. The software offers various logging capabilities, shell-integrated help and documentation, as well as extensive pre-run integrity checks and helpful warning messages. It also allows individual data components to be easily extracted and used in secondary workflows. The code itself is operating system independent and follows best practices in software development. Test coverage, at the time of writing, is 92% and the project uses continuous integration and offers frequent releases.

FAME-Io is designed as one of two main components of FAME, each addressing a particular aspect and user group within ABM development, see [Figure 1](#):

- <sup>18</sup> **FAME-Io** is a Python package<sup>2</sup> for ABM users. It supports configuring complex simulations and managing of files associated with FAME.
- <sup>19</sup> **FAME-Core** (Schimeczek, Deissenroth-Uhrig, et al., 2023) is a Java library for ABM modellers. It supports developing and executing ABM simulations.

<sup>24</sup> These two modules interact with additional supporting components, like FAME-Mpi<sup>3</sup>, FAME-Protobuf<sup>4</sup>, FAME-Gui<sup>5</sup> (in beta release) and FAME-Prepare<sup>6</sup> (under development). The demonstration project FAME-Demo<sup>7</sup> provides a ready-made example for modellers to investigate and experiment with.

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<sup>1</sup><https://gitlab.com/fame-framework/>

<sup>2</sup><https://pypi.org/project/fameio/>

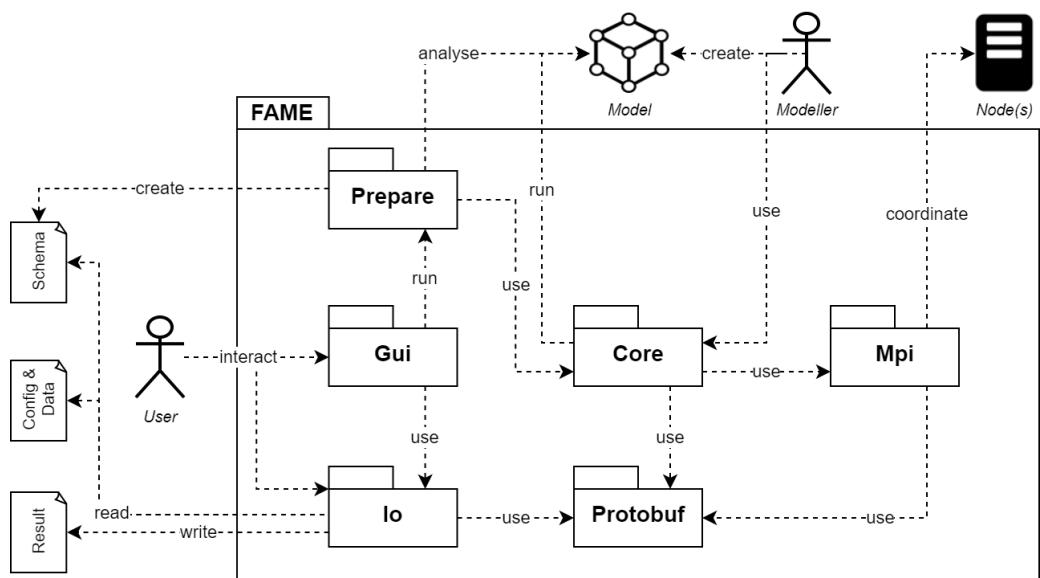
<sup>3</sup><https://gitlab.com/fame-framework/mpi>

<sup>4</sup><https://gitlab.com/fame-framework/fame-protobuf>

<sup>5</sup><https://gitlab.com/fame-framework/fame-gui>

<sup>6</sup><https://gitlab.com/fame-framework/fame-prepare>

<sup>7</sup><https://gitlab.com/fame-framework/fame-demo>



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**Figure 1:** Components of FAME and their interactions.

## 28 Configuration of FAME model inputs

29 Defining agents and their interactions is the first central task when creating any ABM  
 30 simulation. IN FAME, such agent interactions need not be hard-coded, but can be defined  
 31 when configuring the model. FAME-Io enables researchers to quickly create and handle  
 32 configurations of simulations, even complex ones with thousands of agents and numerous  
 33 interactions. To this end, FAME-Io allows splitting configurations into multiple files, each  
 34 dedicated to a specific aspect of a configuration. An overarching configuration comprises one  
 35 or more structured YAML files. Large data structures, like time series, can be imported from  
 36 external (CSV) files. FAME-Io supports easy definition of multiple similar agent interactions,  
 37 i.e., “Many-to-One”, “One-to-Many” and “Many-to-Many” messages, significantly reducing  
 38 repetitive parameterization tasks.

39 Once the simulation configuration files are created, they are validated by FAME-Io against  
 40 a schema derived from the associated FAME model. This validation is comprehensive and  
 41 checks the presence of each agent's mandatory parameters. It also tests the data type of each  
 42 parameter and possible value restrictions. Furthermore, FAME-Io evaluates the plausibility  
 43 of agent interactions, i.e., the correct type of data sent for each agent interaction. If errors  
 44 are detected, FAME-Io issues meaningful messages that help researchers to fix configuration  
 45 problems even before the actual simulation runs. In case no configuration errors are identified,  
 46 FAME-Io creates a single binary protobuf file that serves as input for the associated ABM  
 47 (executed with FAME-Core).

## 48 Conversion of FAME model outputs

49 Once a model is executed with FAME-Core, its entire output is stored in a single binary  
 50 protobuf file. FAME-Io converts this output into human-readable files with a standardized  
 51 output structure in CSV format. Researchers can customize this configuration process to suit  
 52 their needs. FAME-Io allows limiting the conversion to subsets of agents, it offers different  
 53 output formats, and the results of similar agents can be split or grouped. These options  
 54 facilitate later individual post-processing of the results. FAME-Io can also handle very large  
 55 output files, e.g., from simulations with thousands of agents. For computers with limited  
 56 resources, it offers a mode with a low-memory profile.

## 57 Statement of need

58 The energy system transition towards a sustainable future is one of the most urgent concerns  
59 of our time. Hence, the field of energy systems analysis deals with possible paths to a more  
60 sustainable future. However, the increasing complexity of energy markets combined with  
61 different policies and heterogeneous actors poses ever new challenges to this field (Pfenninger  
62 et al., 2014; Pye et al., 2021). ABM is perhaps the most suitable approach to tackle these  
63 challenges (Klein et al., 2019), because

- 64 ▪ the actors' perspective and their specific characteristics are taken into account (Frey et  
65 al., 2020; Kraan et al., 2018), and  
66 ▪ models remain computationally feasible even in large-scale simulations (Hansen et al.,  
67 2019).

68 FAME is a new software suite that helps researchers develop and use ABM simulations. FAME-  
69 models are highly configurable and allow the modelling workflow to be changed without affecting  
70 the model itself. This high flexibility of models, however, leads to the need for a sophisticated  
71 model configuration tool - FAME-lo addresses this need by providing powerful tools for complex  
72 ABM. It provides automatic translation between Java-coded agent specifications and highly  
73 organised configuration files. FAME-lo grants a convenient way to interact with FAME-based  
74 models, i.e. to input data and access their results. It also helps model users handling complex  
75 simulation configurations with hundreds of repeated interactions between agents.

## 76 State of the field

77 A scholarly search returns mostly review articles, e.g., by Ringkjøb et al. (2018). However,  
78 configuration support or input-output management is not an issue in any of them. Hence,  
79 we compare the two most popular ABM frameworks, NetLogo (Wilensky, 1999), which is  
80 widely used in teaching, and Repast (Collier & North, 2013), which was recently ported to  
81 Python. Unlike FAME-lo, both frameworks lack a dedicated input-output management and a  
82 configuration and validation tool. Help with parameterizing a NetLogo model is only provided  
83 by BehaviorSpace<sup>8</sup>. Given this lack of tools, we conclude that the need for a dedicated software  
84 is there and FAME-lo can fill this need.

## 85 Performance

86 FAME-lo offers high performance. It reads, validates, and converts configurations with  
87 approximately 25 agents, 150 contracts and one megabyte of time series data in less than five  
88 seconds on a desktop PC. However, for very large configurations, FAME-lo can run on multiple  
89 computing cores for further speed increase. The YAML and CSV file types used by FAME-lo  
90 are easily readable by both humans and machines. Thus, the configuration, execution, and  
91 result evaluation of FAME-based models can be easily integrated into automated workflows  
92 via FAME-lo. This paves the way for the use of FAME models as part of computationally  
93 intensive approaches, such as extensive parameter sweeps or genetic algorithms.

## 94 Applications

95 FAME was designed for energy systems modelling, but is not limited to this field. Thus, it  
96 can also be used in other domains for ABM simulations with similarly structured problems.  
97 FAME-lo is an integral part of FAME and currently the only tool available for configuring  
98 FAME simulations and accessing their results.

99 A large and powerful FAME-based model is AMIRIS (Schimeczek, Nienhaus, et al., 2023)<sup>9</sup>, an  
100 agent-based market model for the investigation of renewable and integrated energy systems.

<sup>8</sup><https://ccl.northwestern.edu/netlogo/docs/behaviorspace.html>

<sup>9</sup><https://dlr-ve.gitlab.io/esy/amiris/home/>

101 AMIRIS is also open-source and is being used in current projects in the field of energy systems  
102 analysis. Specifically, FAME-lo has been used to integrate AMIRIS into multi-model workflows  
103 with automatic model configuration and result evaluation in the following projects: TradeRES<sup>10</sup>,  
104 ERAFlex II<sup>11</sup>, UNSEEN<sup>12</sup>, and VERMEER<sup>13</sup>. Hence, it has already been applied for various  
105 scientific publications (Frey et al., 2020; Nitsch et al., 2021; Safarazi et al., 2020).

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