



Reproducibility in Modeling

Methods implemented in the ABM Simulation Framework FAME

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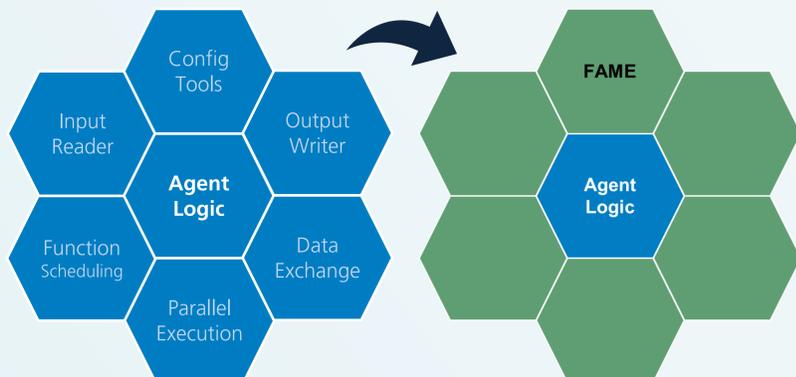
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FAME

open **F**ramework for **A**gent-based **M**odelling of **E**nergy systems

Aim: Reduce overhead code



- Support **model development** of energy-related Agent-Based Models (ABM)
- Provide **fast execution** of models at low overhead
- Assist **all project stages**: from rapid prototypes to large & complex models
- Grant **scalability**: from laptop to High-Performance Computing (HPC) cluster

In a nutshell

Write less code

- FAME organises**
 - Data exchange
 - Order execution
 - Inputs & Outputs
 - Parallelisation

Configure, not code

- FAME offers flexible**
 - Input
 - Output
 - Agent interactions

Scale to your needs

- FAME is portable**
 - Windows / Linux / Mac
 - Laptop / Server / HPC

No hidden costs

- FAME is open**
 - Apache 2.0
 - REUSE complaint

Full model support

- FAME helps developers/users**
 - Coding models
 - Configure scenarios
 - Run simulations



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

Components & workflow

FAME is divided into several components, each addressing a specific task:

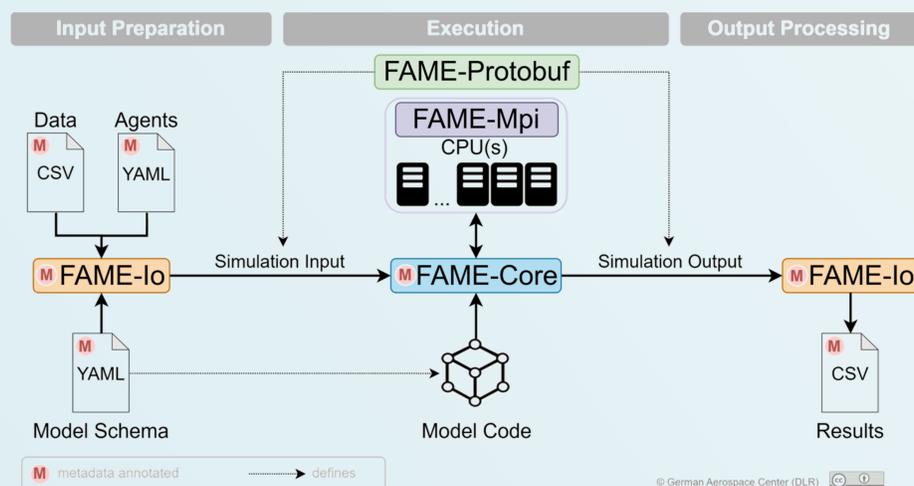
FAME-Core: Provides methods to create & run agent-based simulations (Java)

FAME-Io: Feeds input data to & extracts results from simulations (Python)

FAME-Mpi: Coordinates processes in multi-core mode using MPI (C++)

FAME-Protobuf: Defines file formats & message types (Google Protocol Buffers)

FAME-Gui: Drag & drop configuration of FAME-based models (Python)



Activity: FAME-Io



1191 commits | Last commit ≈ 2 days ago

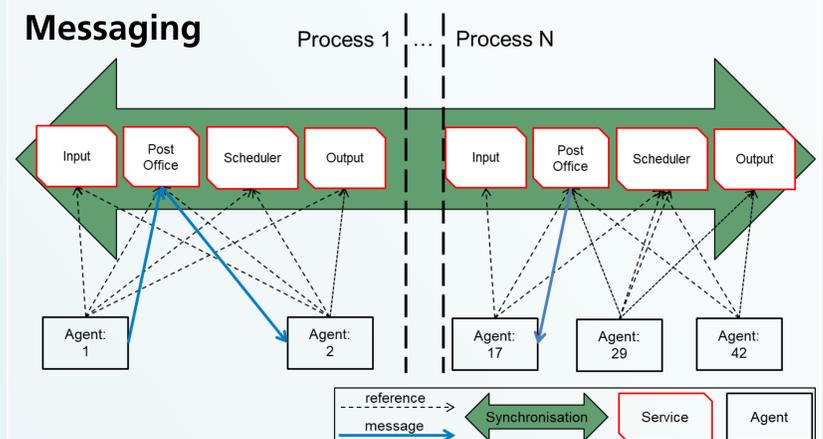
Concurrency

Basic idea

Zero knowledge: Modellers do not need to consider parallelisation

Parallelisable by default: every FAME model can be run concurrently

Messaging



Services: Each process has a set of Services that connect to local Agents

Synchronisation: Services update each other every tick

Agents: Have a unique ID; Agents don't share any objects

Messages: Are used to exchange data between Agents; 1 tick delay

Scheduling



Agents: Ask Scheduler to schedule Actions at specific simulated Time

Scheduler: Determines Time of next tick; calls Actions of non-idle Agents

Actions: Methods of Agents executed at specific Time

Simulated time: Non-continuous timestamp chain; 1s minimal resolution

Contracts: Define repeated Actions and Message deliveries

Result Reproducibility

Random number generation

Dedicated service: Provides random number generators to Agents

Seed: Derive seeds from configuration: Main seed + Agent ID + Counter

→ Random numbers are independent of process count involved

Comprehensive outputs

Single output file: includes model input, output, metadata

Log tool versions: of involved FAME-tools and models

Rerun capability: Rerun a simulation using output file as input

Metadata support

Everything: can have a metadata description

Describe: Models, agents, files, inputs, outputs

Free Format: Format of metadata is choice of modeler

Template Processor: Automatically extract metadata and assign to model output files; default: OEO metadata template

Outlook

Go Python: Enable Agents to be programmed in Python

Reduce boilerplate: Autogenerate message classes from schema file

Enhance verifications: Check contract consistency during configuration

Use FAME

See example: fully-fledged open energy market simulation AMIRIS: <https://helmholtz.software/software/amiris>

Ask questions: unsure how to use FAME? Contact us at: FAME@dlr.de

Get started: Create your own FAME model! Start from our example: <https://gitlab.com/fame-framework/fame-demo>

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