

2025

modeFRONTIER Product Update

Streamlined process automation
for AI data-driven modeling





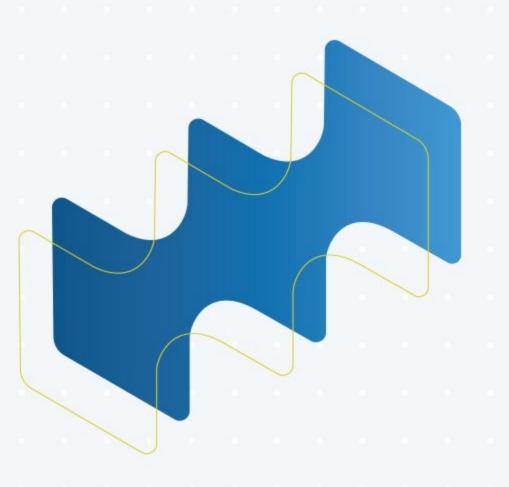
Agenda

Streamlined operations with process automation and system integration

Enabling data-driven modeling with modeFRONTIER evolved process automation and integration

modeFRONTIER roadmap 2026/2027

USERS' MEETING NORTH AMERICA

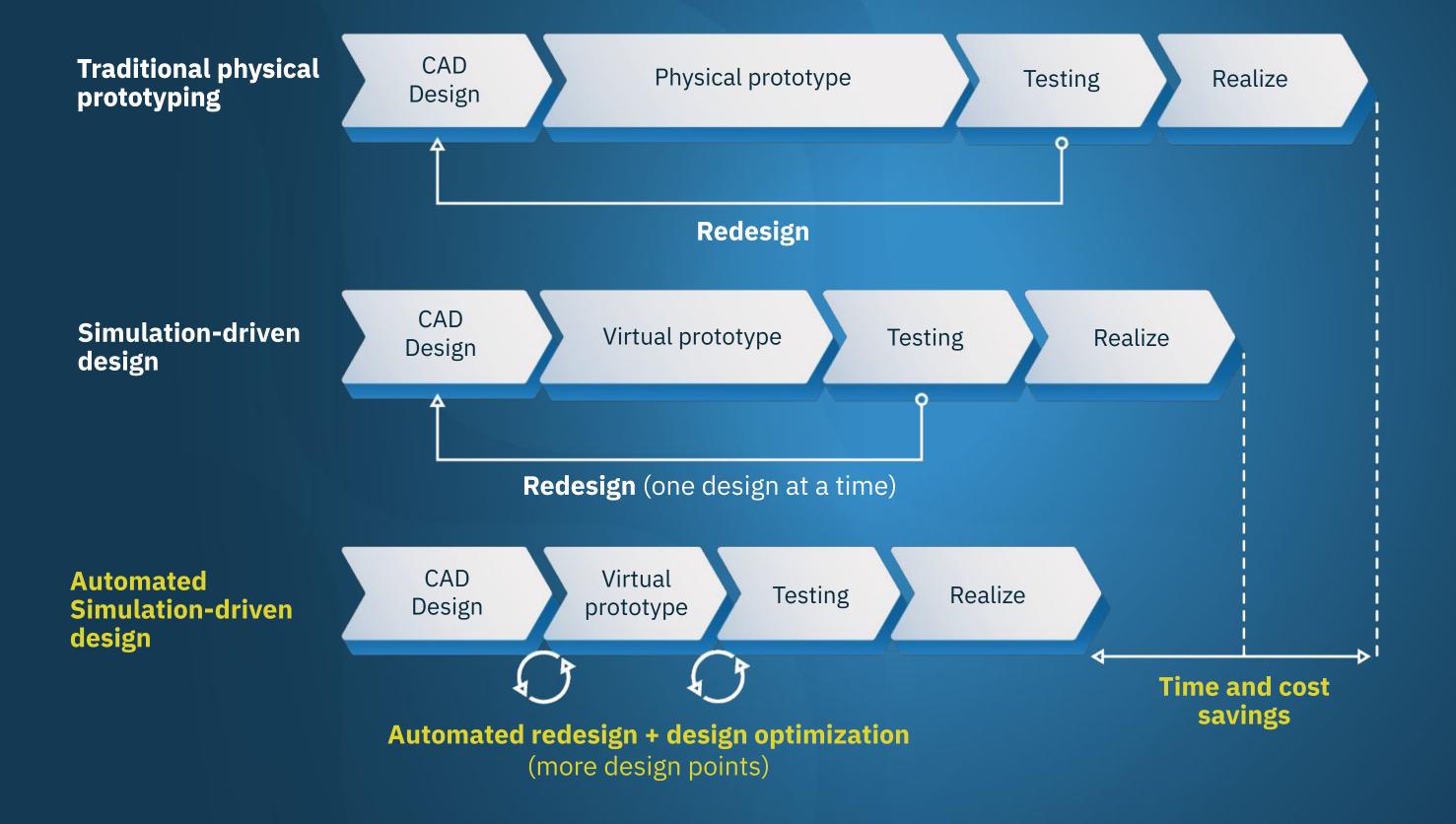




Streamlined operations with process automation and system integration

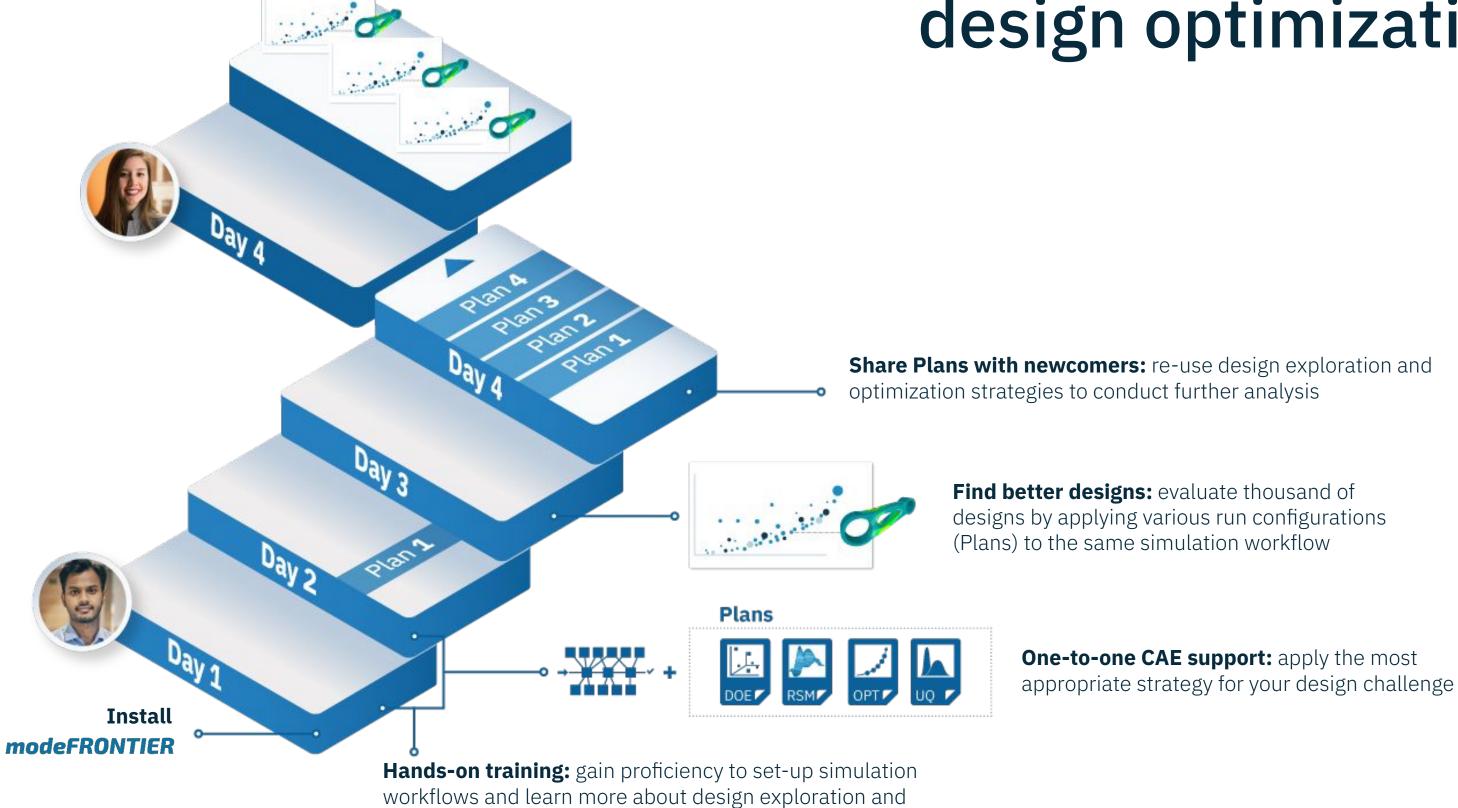


Business challenge: time and cost savings





The ideal scenario for effective design optimization



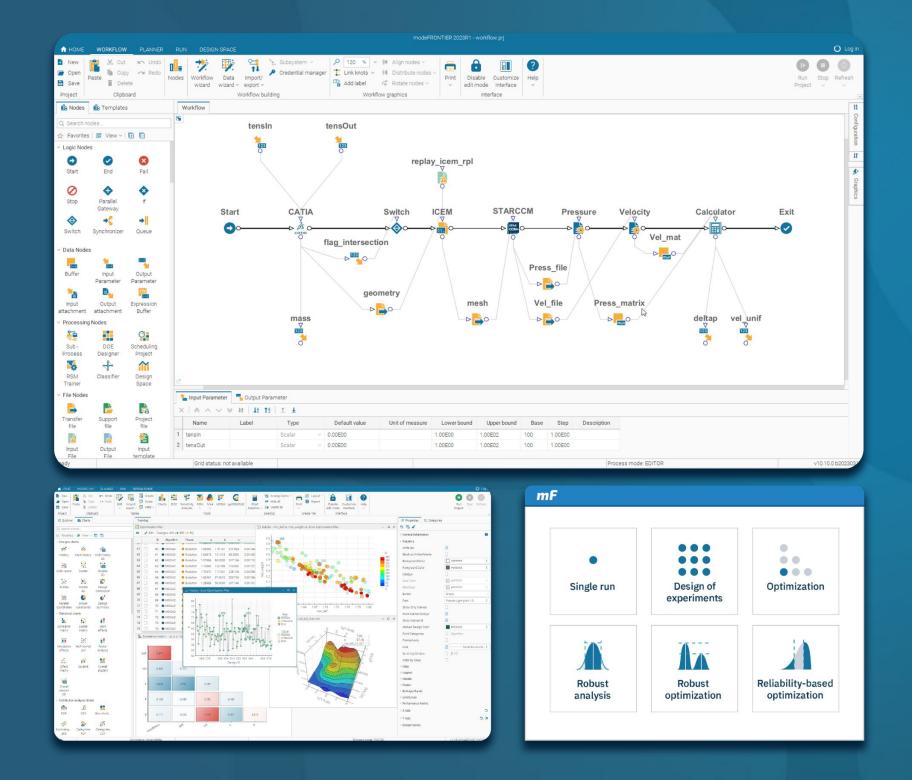
optimization strategies





modeFRONTIER

The vendor agnostic software for process automation and design optimization.

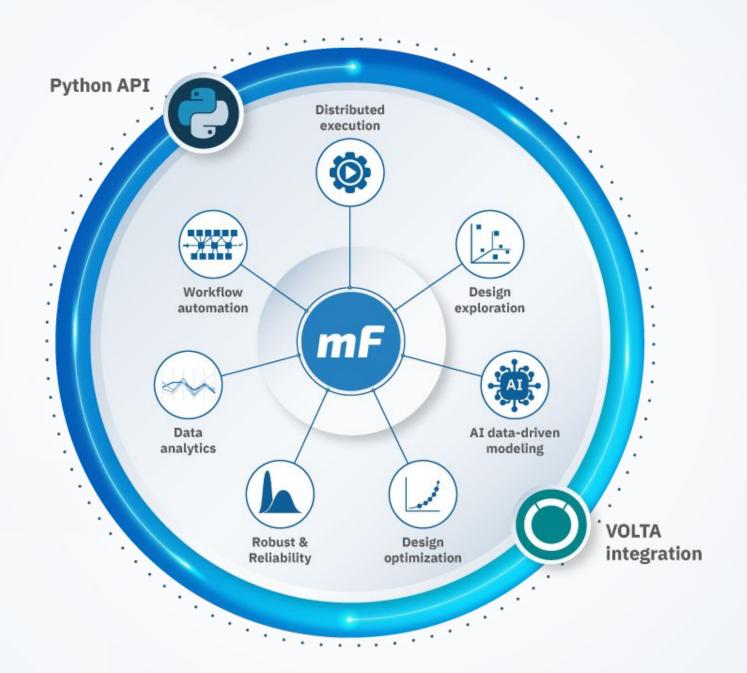


Automatically explore the design space with AI-enabled methods and accelerate product development.



modeFRONTIER **Machine Learning** and Deep Learning algorithms **Optimization** algorithms autonomous mode Today **Advanced** postprocessing Python API Multiple design exploration strategies set-up **Multi-objective** robust design optimization Proprietary pilOPT algorithm: explore + exploit in one click Design space statistics and Response Surface Models (RSM) **Proprietary Multi-Objective Genetic Algorithm** (MOGA-II) Vendor agnostic graphical simulation workflow 1999





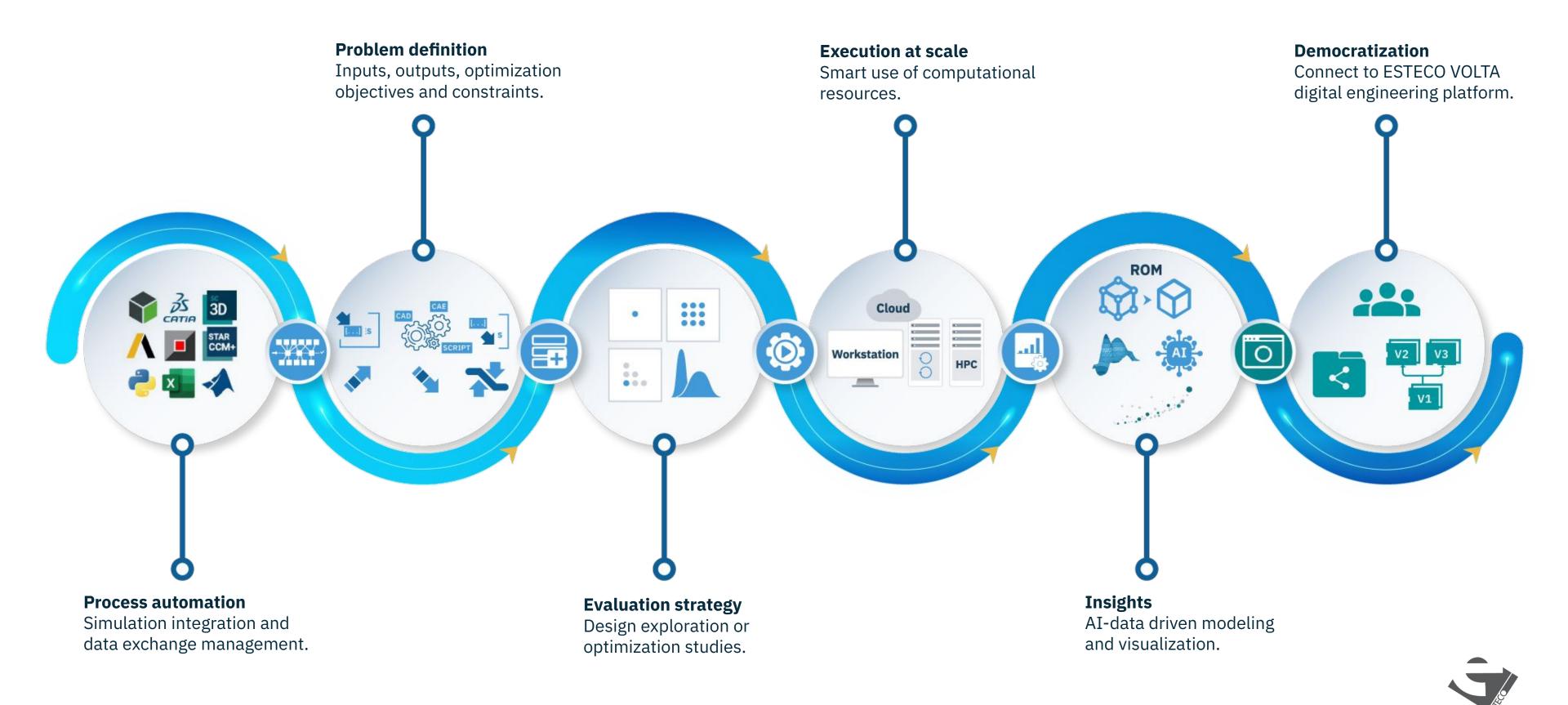
- Usability: workflow and Planner environment
- Enhanced integration with engineering applications



Enabling data-driven modeling with modeFRONTIER evolved process automation and integration



The modeFRONTIER process flow



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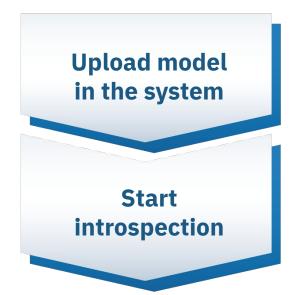






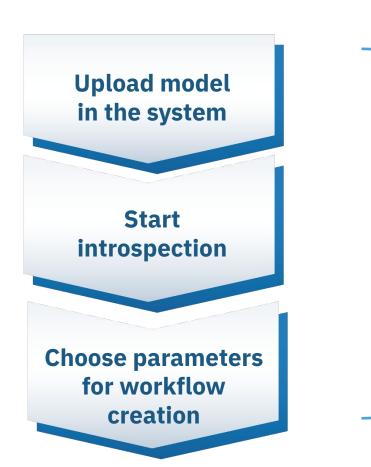


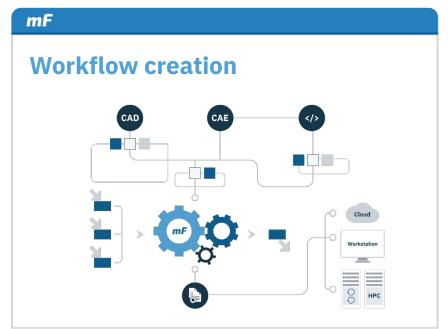


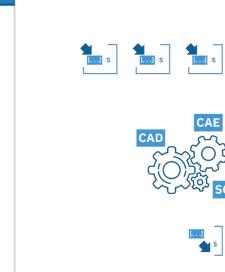












Input variables

Define the design space

Black box

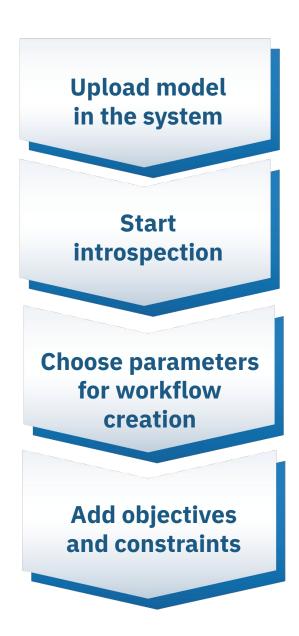
Compute outputs based on inputs

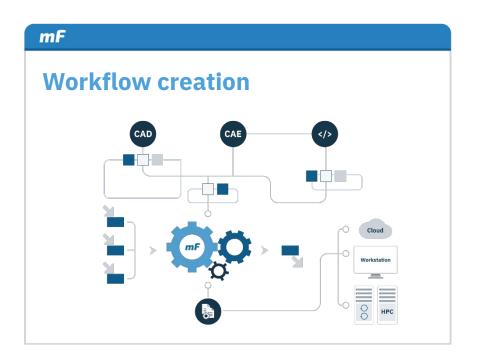
Output variables

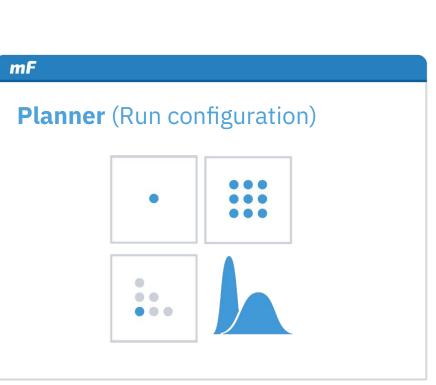
Response of the system

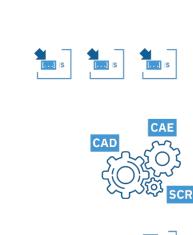












Input variables

Define the design space

Black box

Compute outputs based on inputs

Output variables

Response of the system



Evaluation strategy

Single run, DOE, Design Optimization, or Robust and Reliability analysis



Objectives

A quantity you want to minimize or maximize

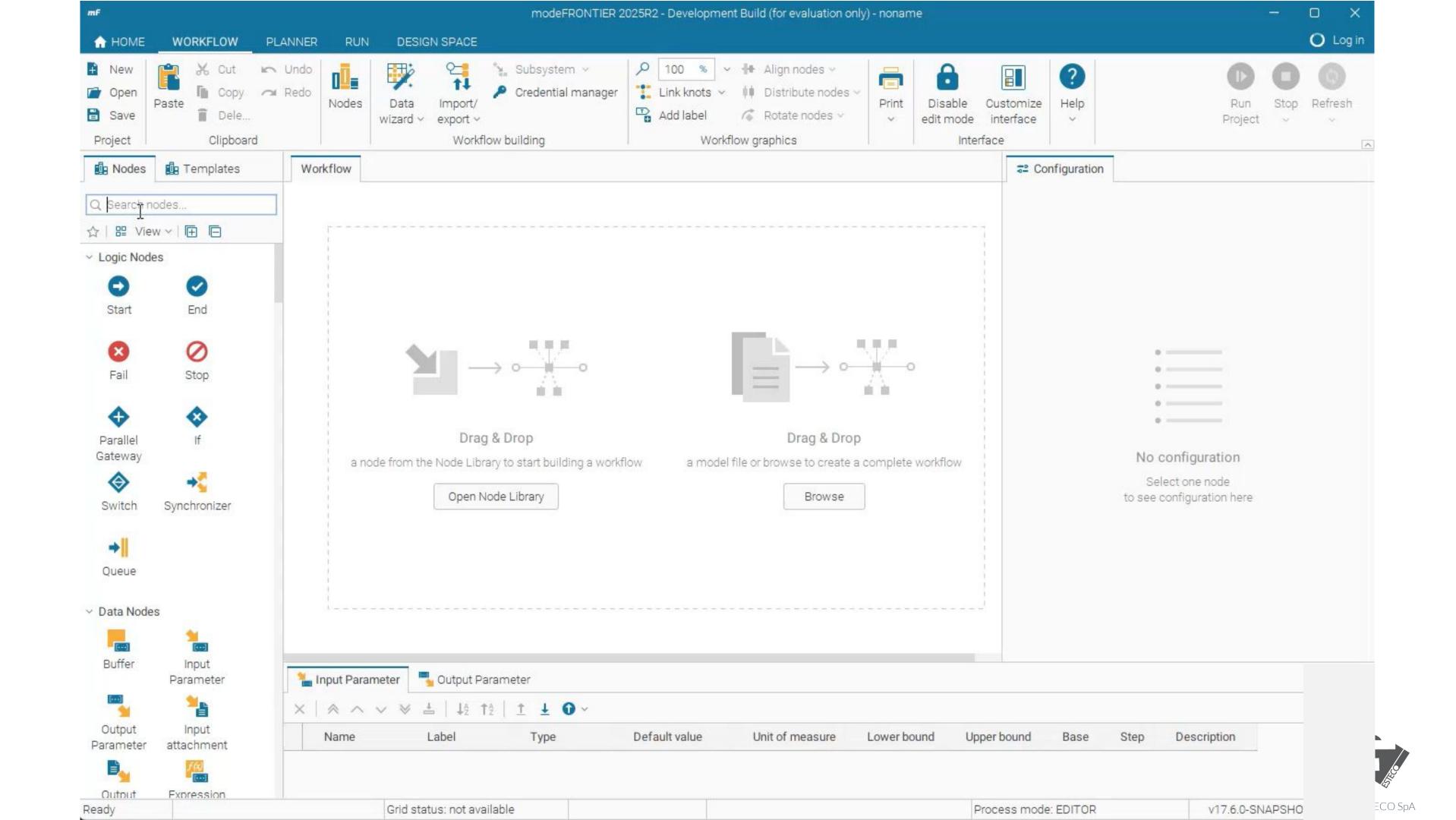


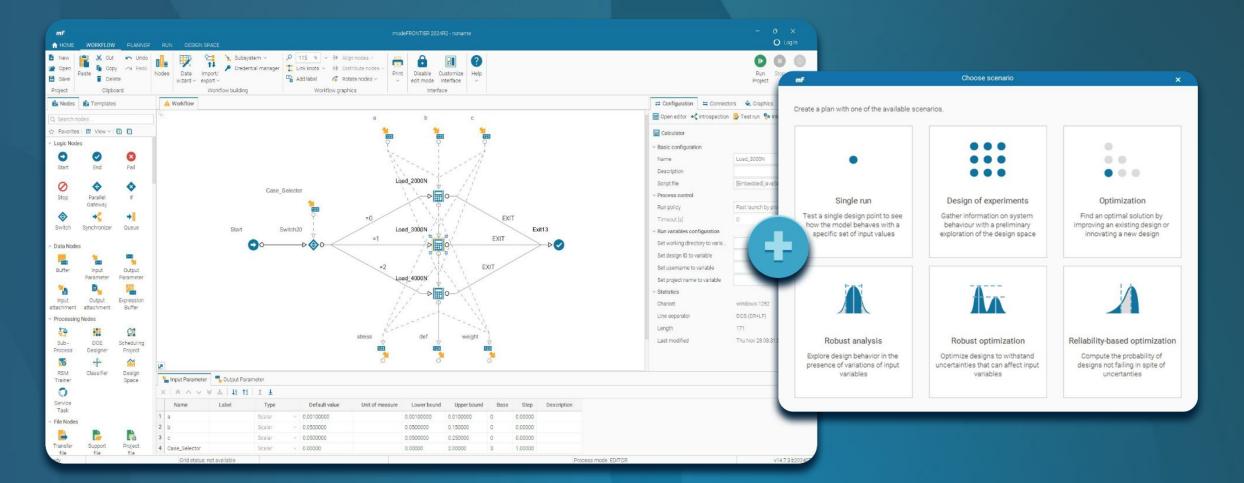
Constraints

A limit you want to respect









Usability: workflow and Planner environment

- . Improved Planner functionality (Process mode vs Scheduling node)
- Plan services
- Service task node





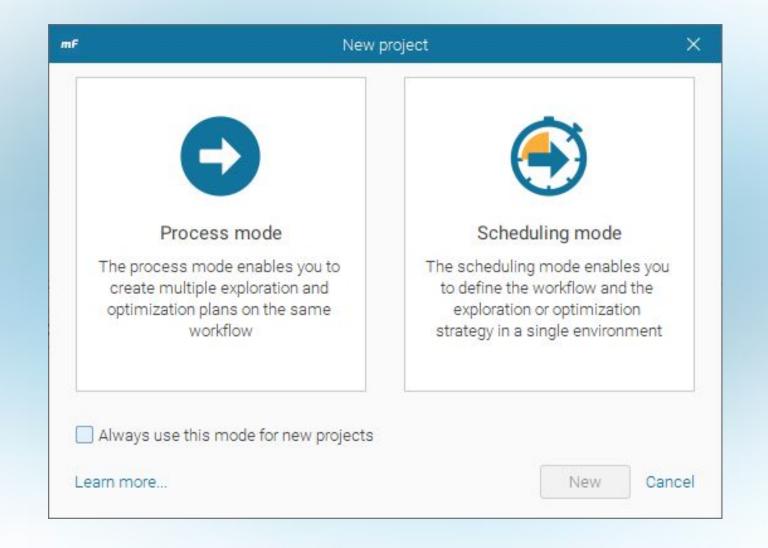
Process node vs Scheduling node

Process node

Make it easy to reuse and compare different plans for the same process.

Scheduling node

For those who want detailed control over every aspect of the process.

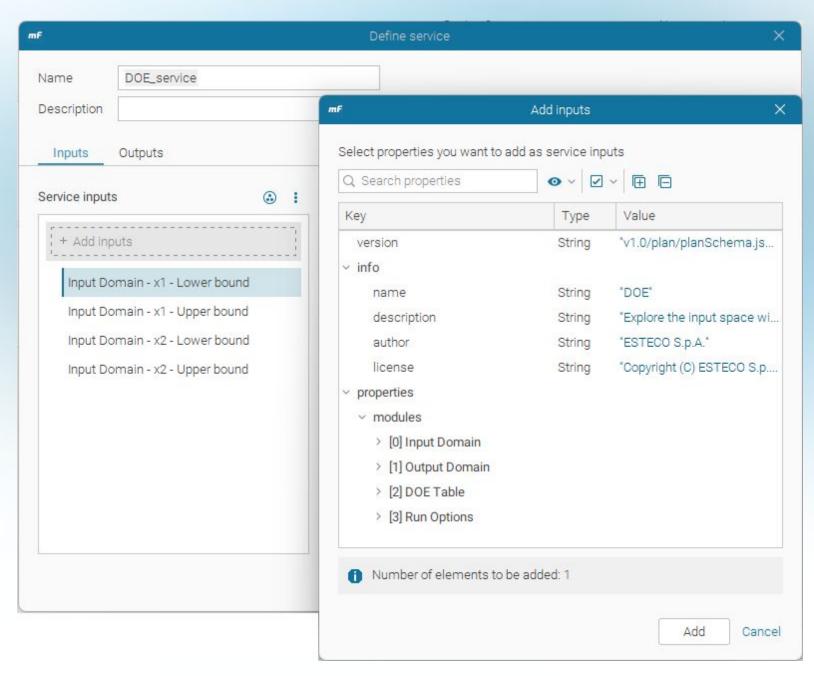






Plan services

modeFRONTIER Plan Services are subsets of plan elements that allow nesting and running multiple projects within a single workflow.



Reduced complexity

Projets can have many inputs/outputs, sometimes making them difficult to manage. Services can focus on a limited set chosen by the creator.

Fully parametric

Services can include properties of inputs (bounds, defaults), outputs, objectives/constraints (expressions, types), and algorithm settings.





Service task node

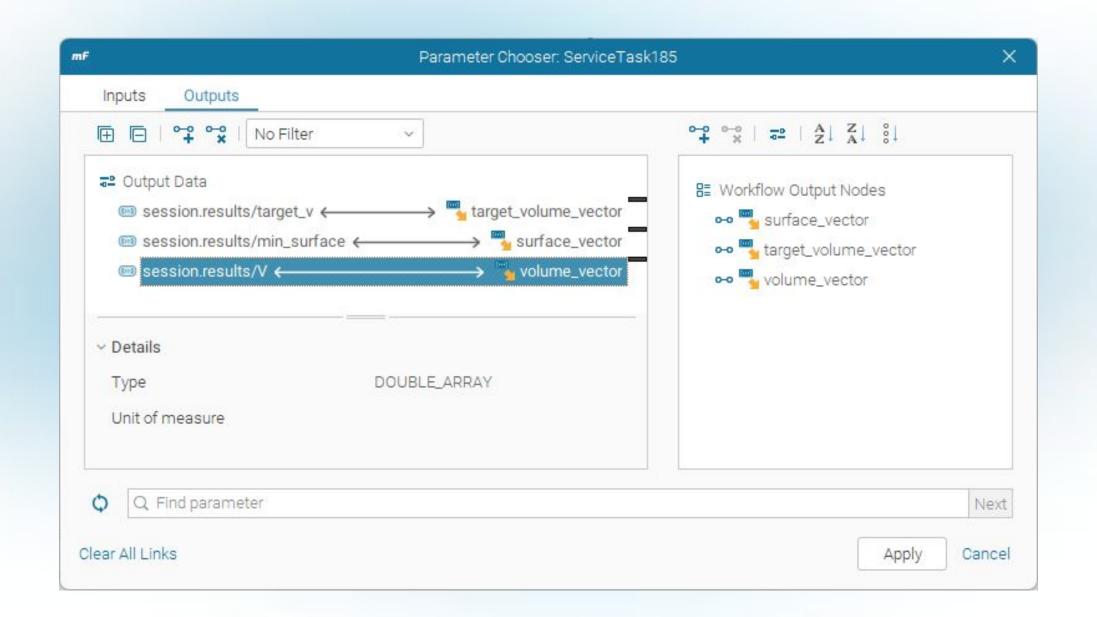
It lets you run a smaller project (service) within a larger project (main).

Service Task Parameters

Service properties become parameters in the Service Task node's Parameter Chooser.

Data handling

The main project parses data and handles data exchange with the inner service.

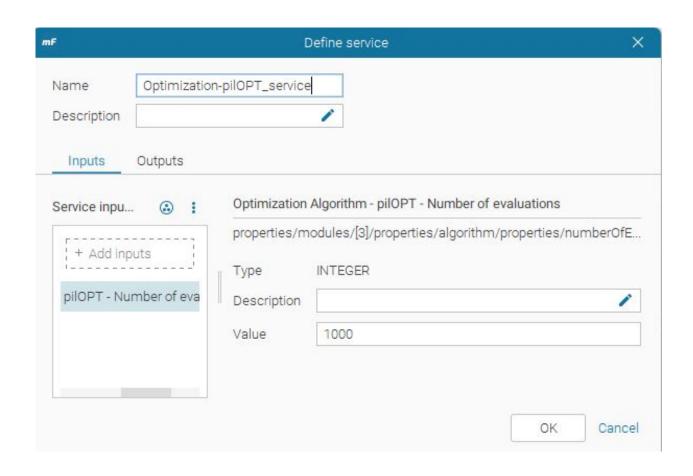






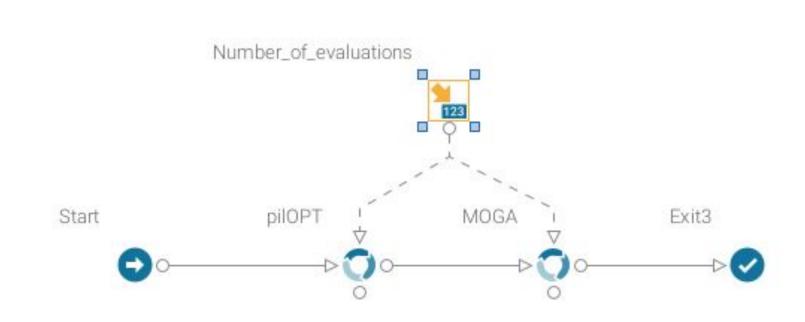
Example: run modeFRONTIER projects in batches

You can easily configure and run different optimization strategies on the same workflow.



Step 1

Define two services that expose the number of evaluations for pilOPT and MOGA-II, both in self-initializing mode.



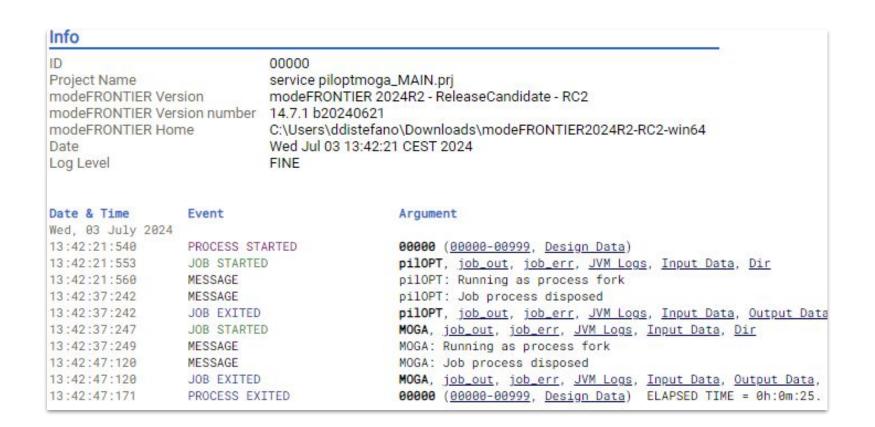
Step 2

Create the main workflow using two service task nodes.





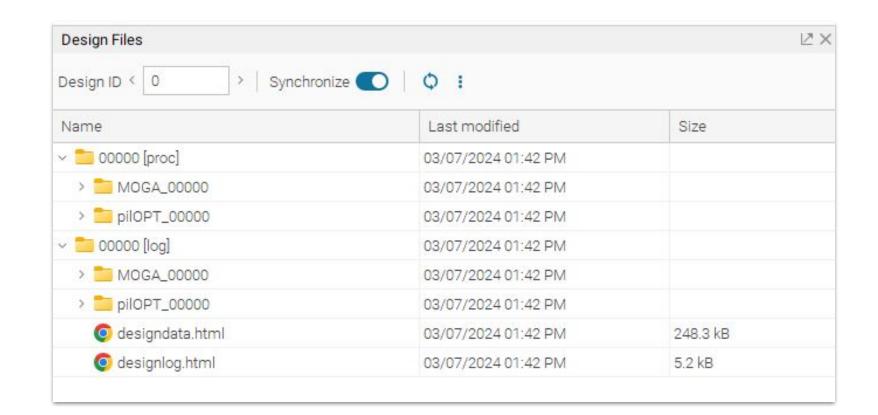
Example: run modeFRONTIER projects in batches



Step 3

Run the main workflow.





Step 4

Full results of both sessions are available in dedicated folders.



Enhanced integration with engineering applications







Adams/Car



Adams/View



Ansa



ANSYS Workbench



ASMI



AVL AST



COMSOL



CREO Elements/Pro



Microwave



















CATIA

CATIA





designer



Keysight **ADS**











Madymo



Maplesim



Mathcad Prime



Metapost

图



MSC Nastran



Python



Romax Designer



ROMbox ESTECO



SFE Concept



Simcenter 3D



Simcenter Amesim



Simcenter **Flowmaster**



MATLAB

Simcenter StarCCM+



SimulationX



SolidWorks



SolidWorks flowsimulatio n



SpaceClaim

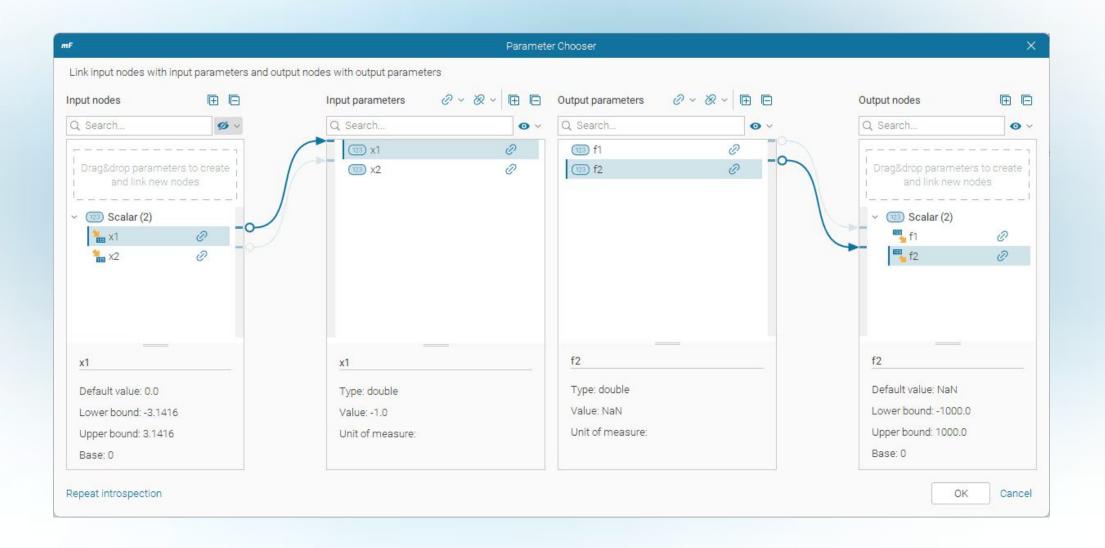






New parameter chooser

Gain efficiency and control in linking your model parameters to the modeFRONTIER workflow.



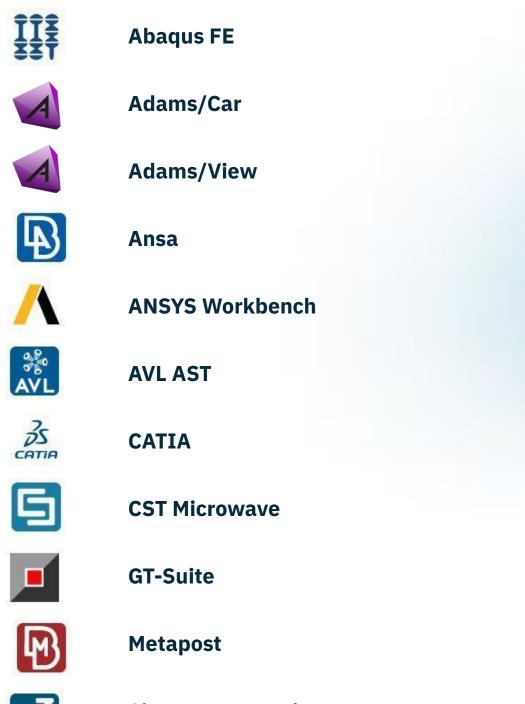
- Side-by-side visualization: easily compare inputs and outputs.
- Flexible workflow building: view and link any data node in the workflow.
- Streamlined navigation and filtering: quickly find and connect nodes and parameters.
- Automatic linking of parameters and nodes: reduce time and effort.
- Enhanced visualization of linked items.

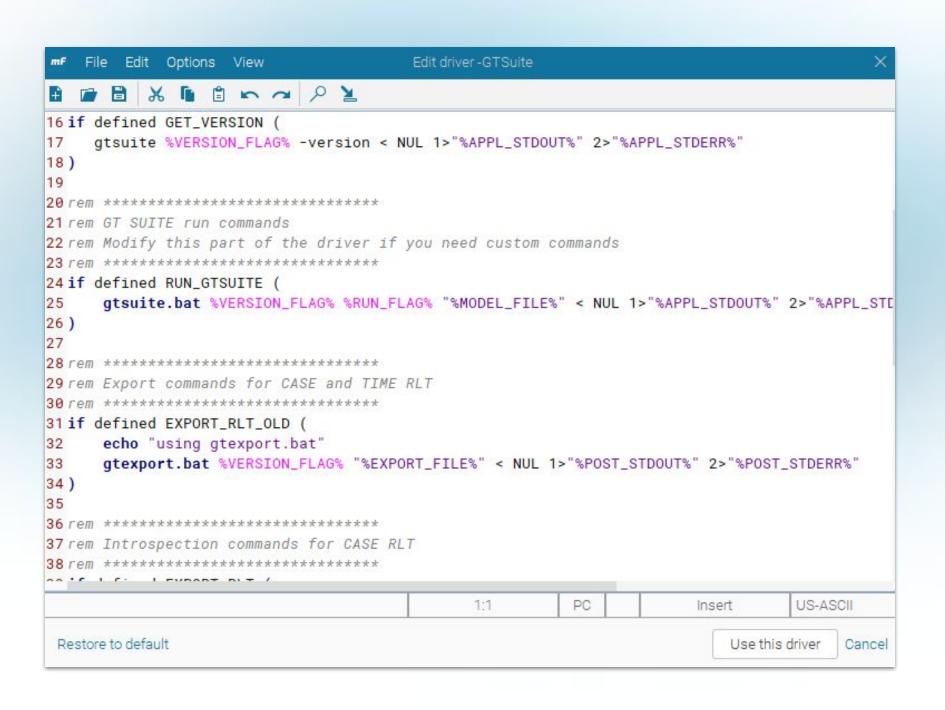




Test run feature extended to CAD/CAE connectors

Verifying integrated software communication and custom driver by running a design with default/custom values and validating outputs.



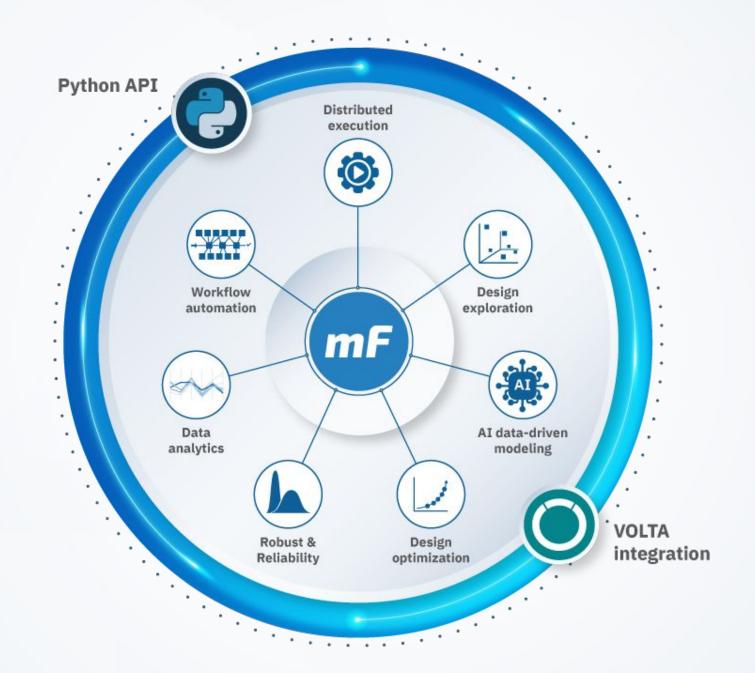






Simcenter Amesim



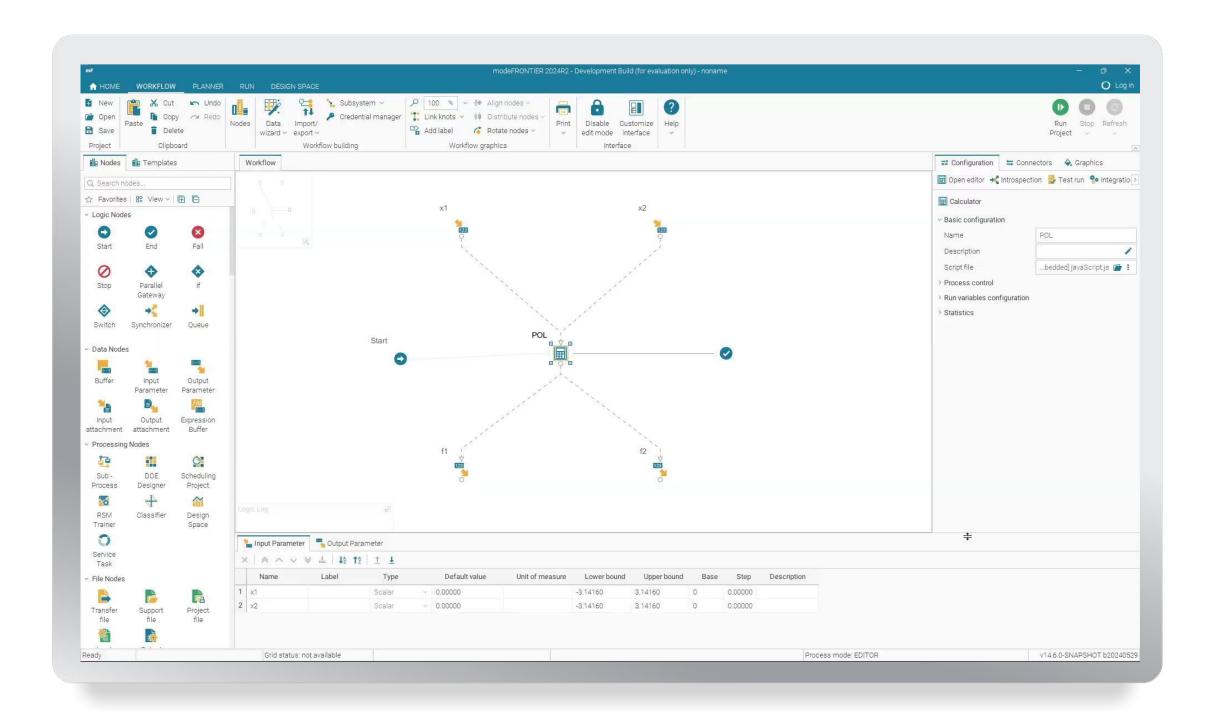


Deeper python integration (pyFRONTIER, pyCONSOLE, pyDOE, pyRSM and pySCHEDULER) for scripting and customization.



pyFRONTIER

Interactive access to modeFRONTIER from an external Python interpreter (IDE or external solver) through a dedicated set of Python APIs.



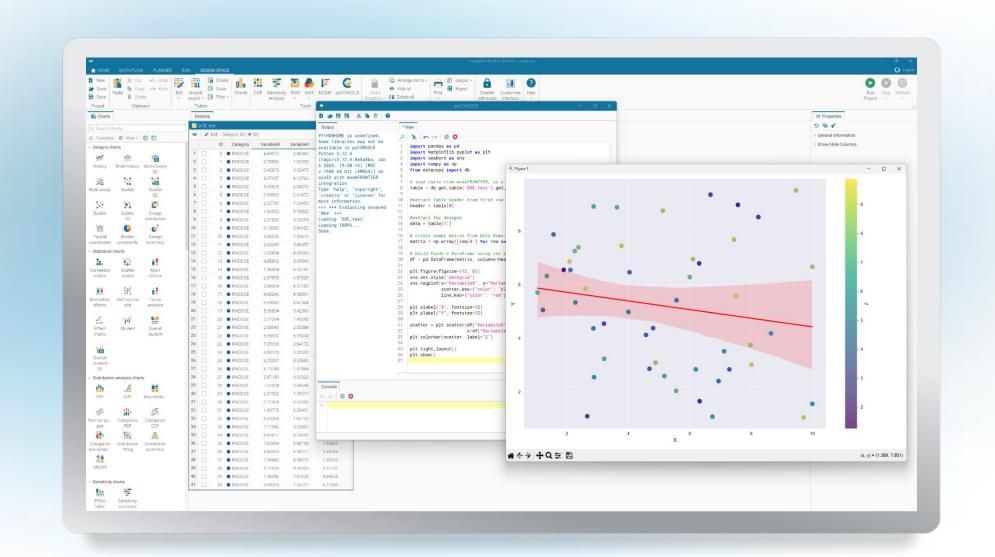




pyCONSOLE

It extends modeFRONTIER's capabilities, adding the power of Python for more flexible and efficient design exploration.

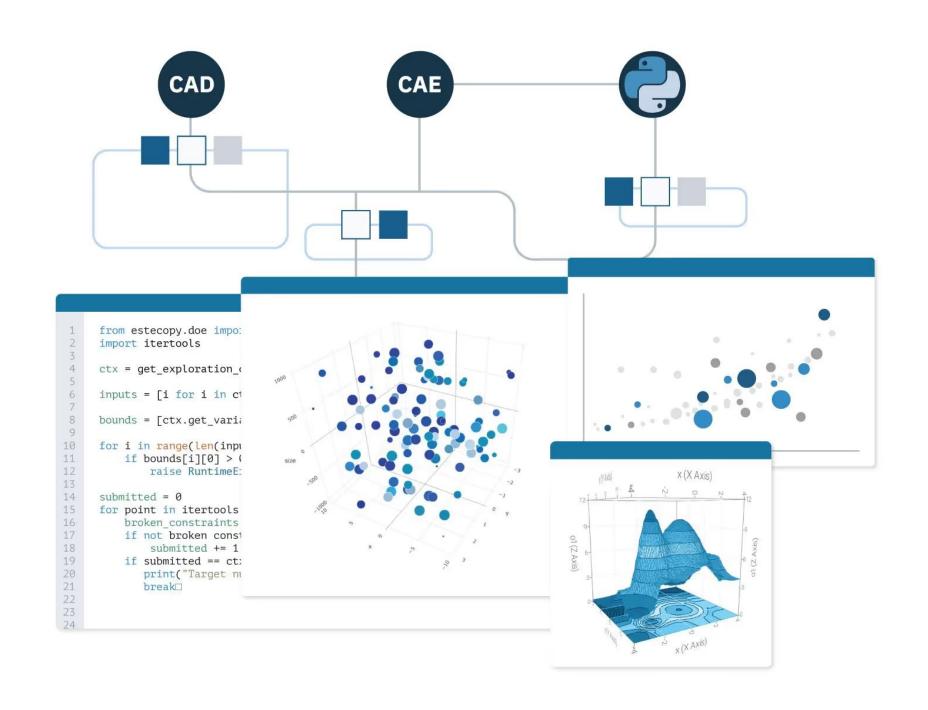
- Seamlessly integrated Python shell within modeFRONTIER's design space, for direct access to design variables, objectives, and constraints
- Enabling real-time data manipulation and analysis, by leveraging Python's extensive libraries for enhanced functionality
- Ideal for quick prototyping and debugging







Import and train your Python algorithms



- pyDOE: write your own or use built-in Python algorithms to create a set of DOE design configurations
- pyRSM: use external Python ML libraries (i.e. scikit-learn) to perform RSM analyses
- pySCHEDULER: leverage external Python Scientific and ML libraries (i.e. SciPy optimize) to perform design optimization

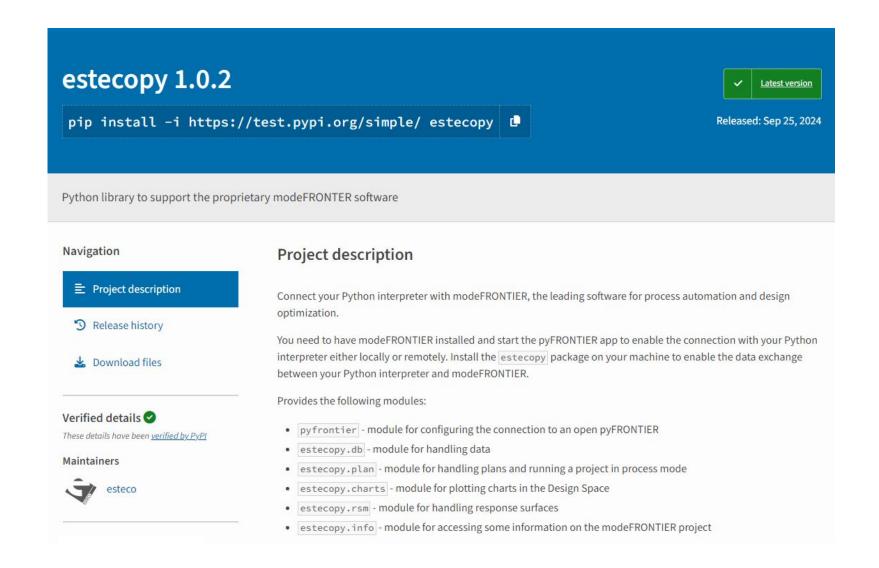




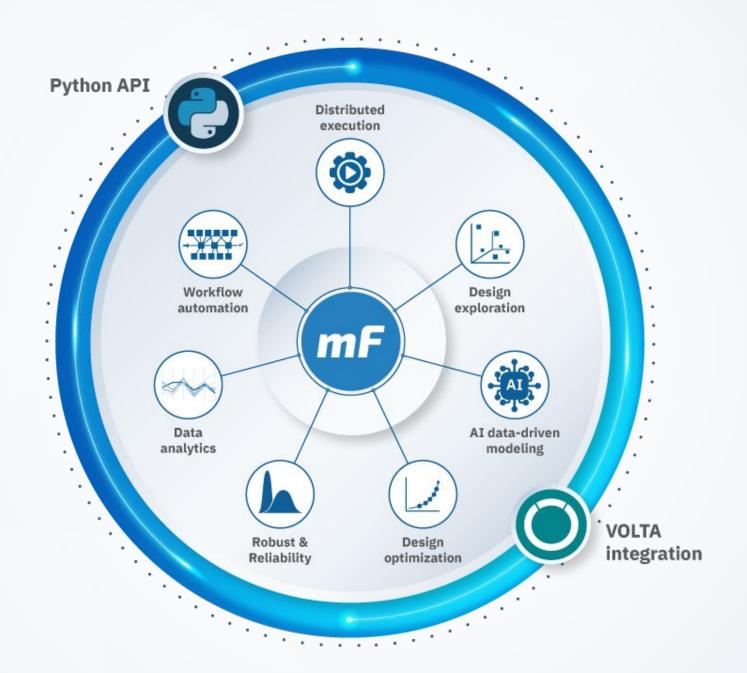
Enhanced accessibility with estecopy on pip

The estecopy library is now accessible on pip, the Python package manager. This makes it easier for you to install and use the library.

- Increased accessibility: easily access and utilize the estecopy library.
- Improved efficiency: the simplified integration process allows for quicker and more efficient interaction with modeFRONTIER.





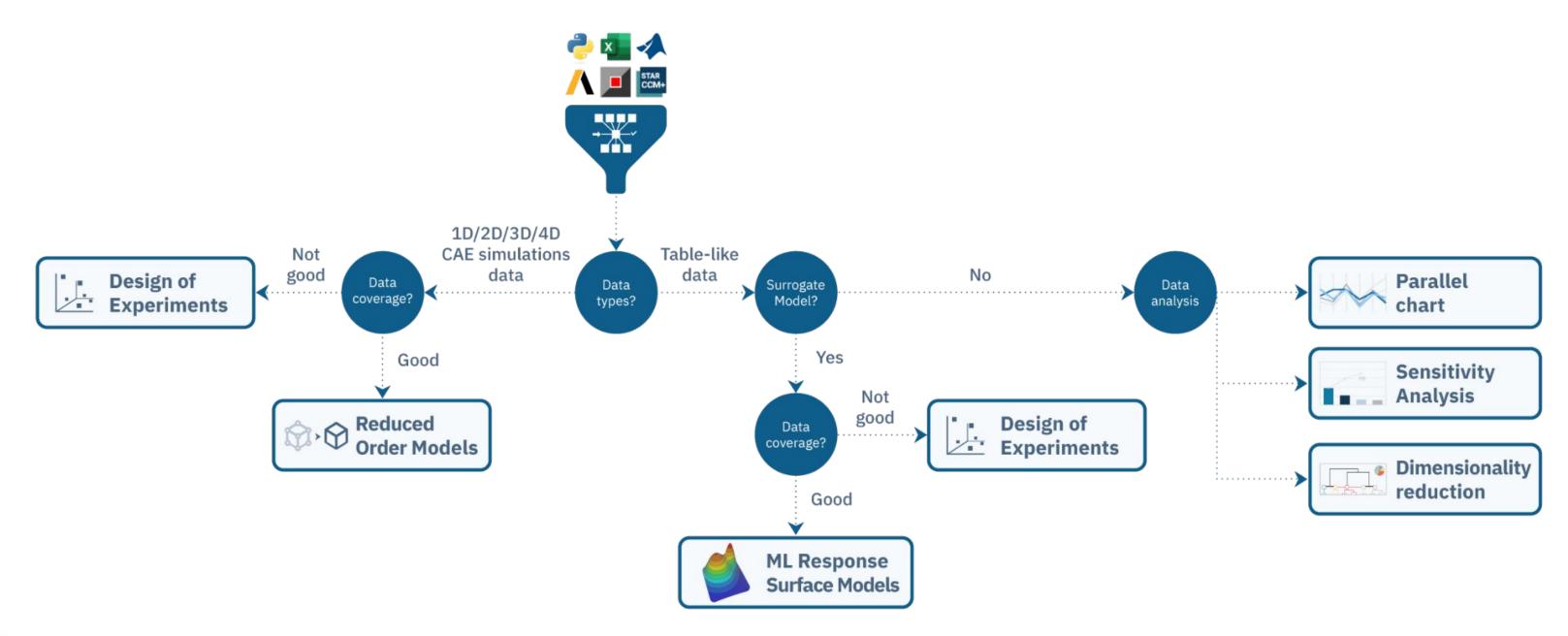


- Reduced Order Models(ROM) capabilities
- New algorithms:
 Multi-fidelity RSM and
 MUSA optimizer



The modeFRONTIER AI/ML process data flow

Enabling the development of computationally efficient surrogate models that expedite the exploration of complex designs spaces.

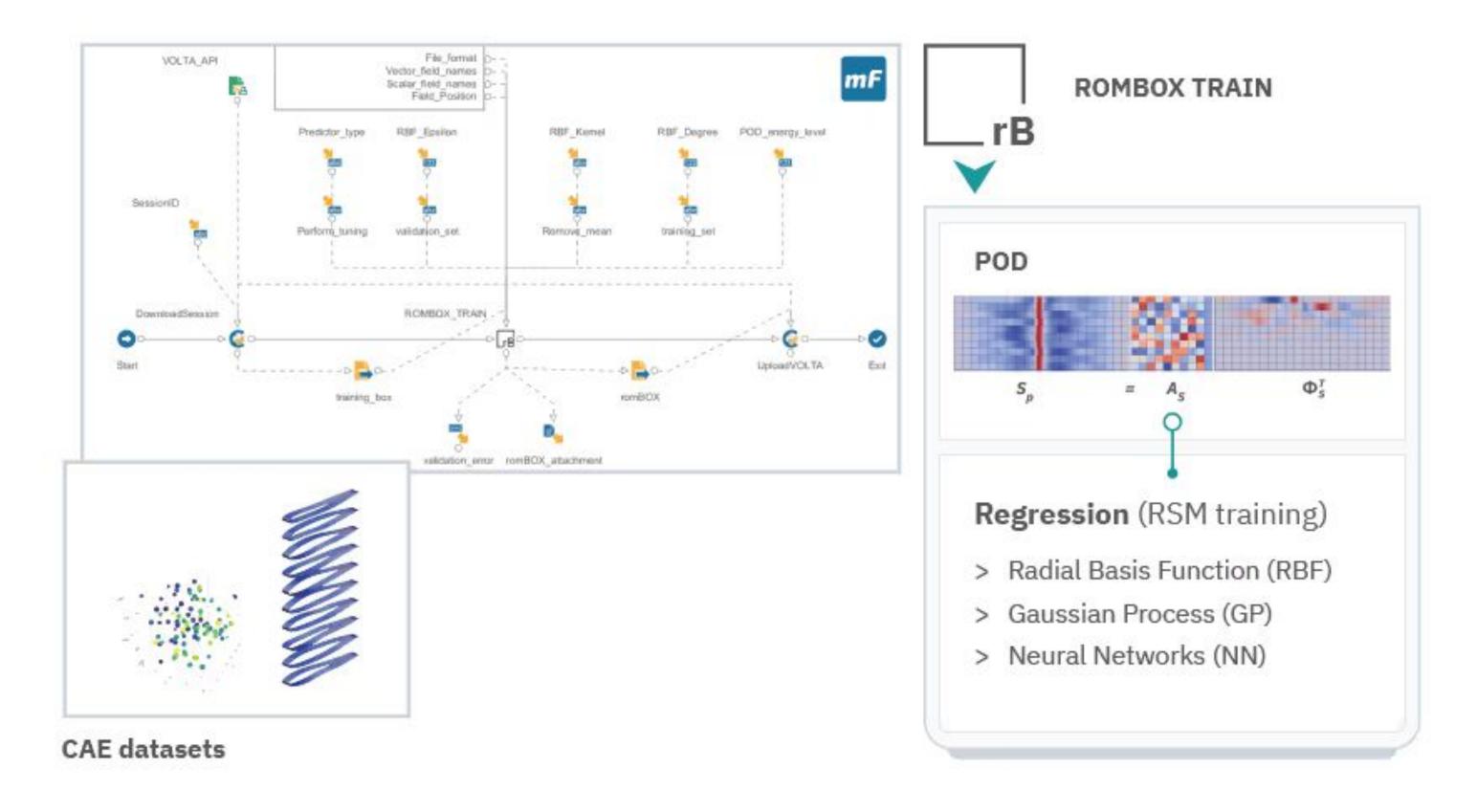






Explainable AI for Reduced Order Models (ROM)

Build CAE datasets and automate ROM training with modeFRONTIER.

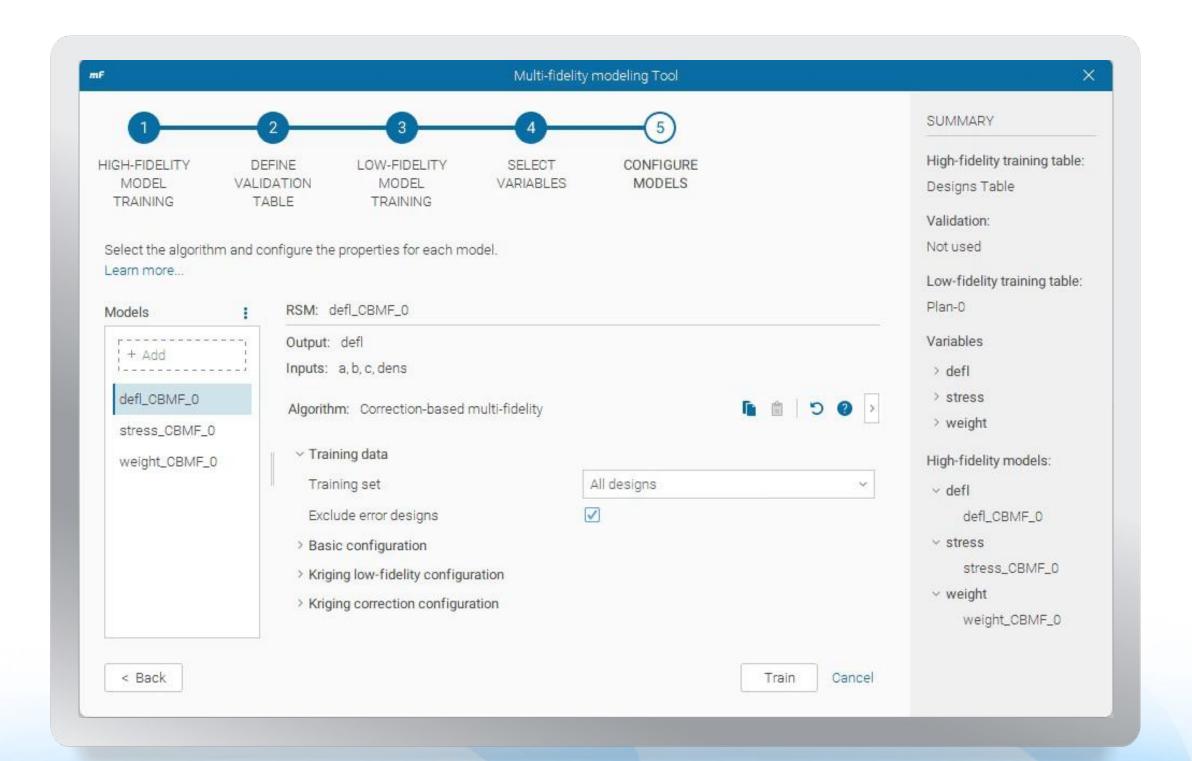






New Multi-fidelity RSM algorithm

Bridge the gap between rapid computation and high accuracy. Integrate knowledge extracted from high-fidelity simulations with low-fidelity data.







Real-world impact

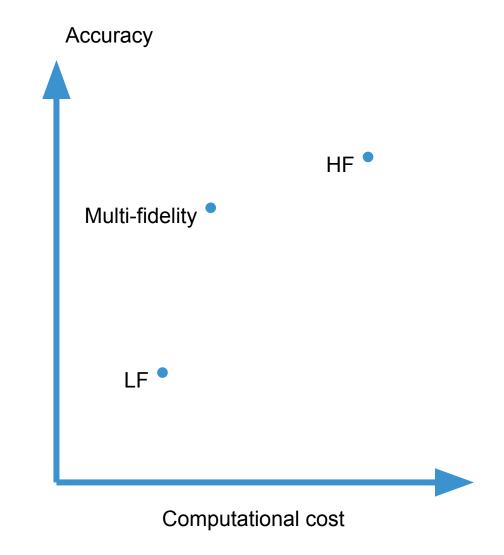
Benefits and applications in real design problems.

Benefits:

- Overall affordable computational cost.
- Improved accuracy over LF alone.
- Better coverage than HF alone.

Applications:

- Aerodynamic design optimization.
- Structural analysis.
- Any field with expensive simulations.







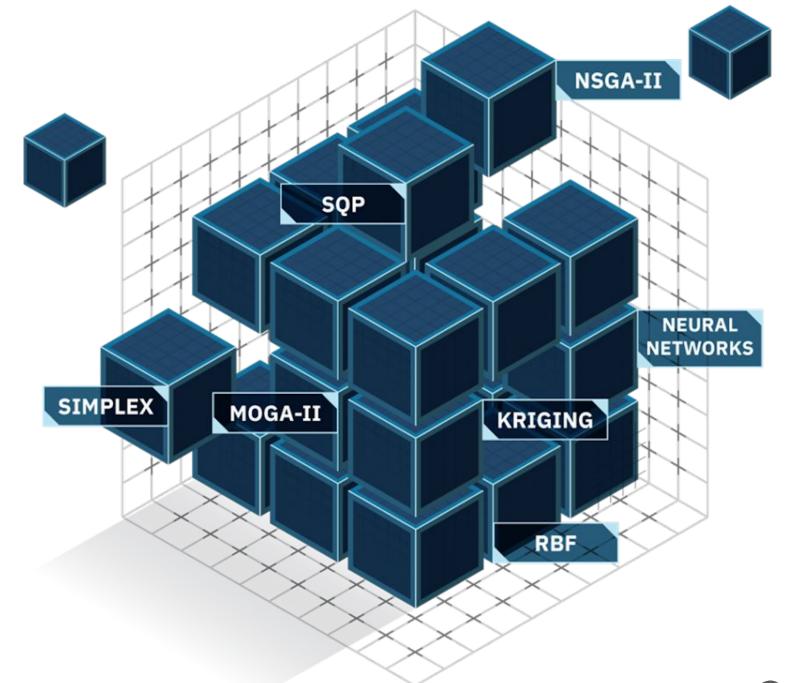
A new optimization algorithm: MUSA

Unique island-based architecture

MUSA enhances population diversity, reduces computational time, and supports both continuous and discrete variables.

Leveraging RSM-based optimization

MUSA integrates global search algorithms (MOGA-II/NSGA-II) with local refinement techniques (Simplex/AFilterSQP/NBI-AFSQP), delivering robust performance in complex design spaces.







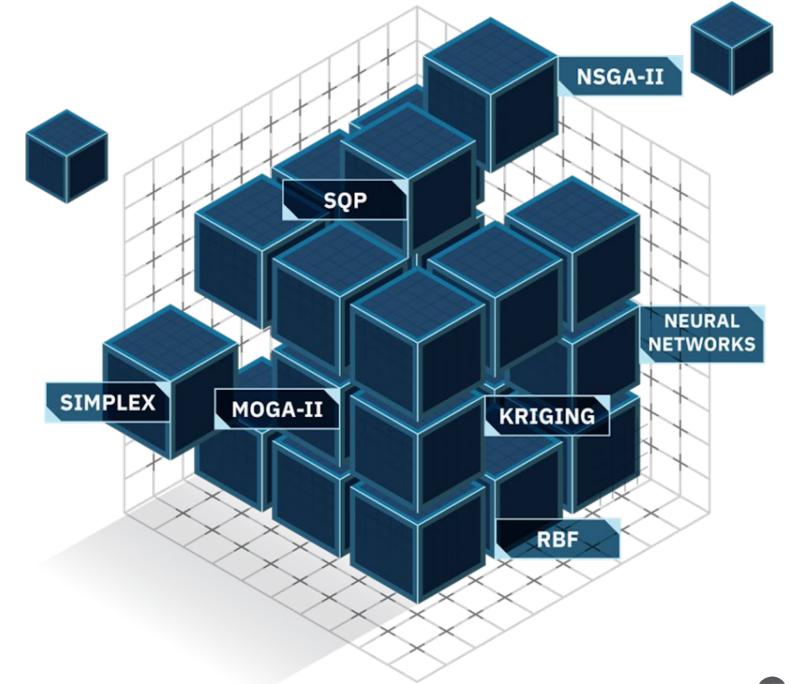
MUSA: how it works

1. Hybrid Global-Local Search Architecture

MUSA integrates MOGA-II/NSGA-II for global exploration with Simplex/AFilterSQP/NBI-AFSQP for local refinement, balancing exploration (diverse solutions) and exploitation (precision).

2. Island-based parallelism

By partitioning populations into semi-isolated "islands", MUSA enables parallel computation, each island operating independently.







MUSA: how it works

3. Dynamic RSM competition

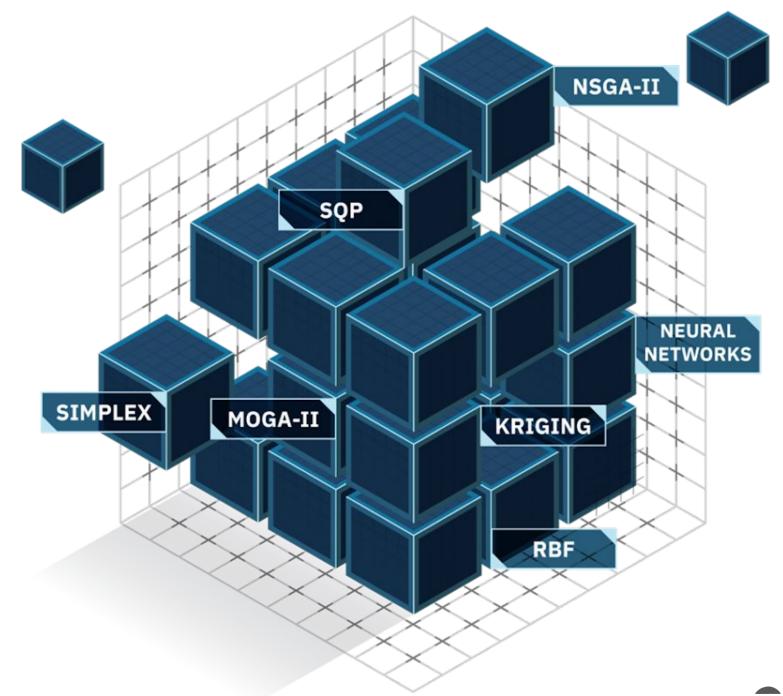
MUSA simultaneously trains RSM (Kriging, RBF, etc.) and selects the most accurate model per iteration.

4. Constraints Handling

Native support for gradient-based constraints.

5. Parameters

MUSA dynamically adapts to user-defined probabilities for: global/local search ratio, virtual optimization probability, RSM training set.

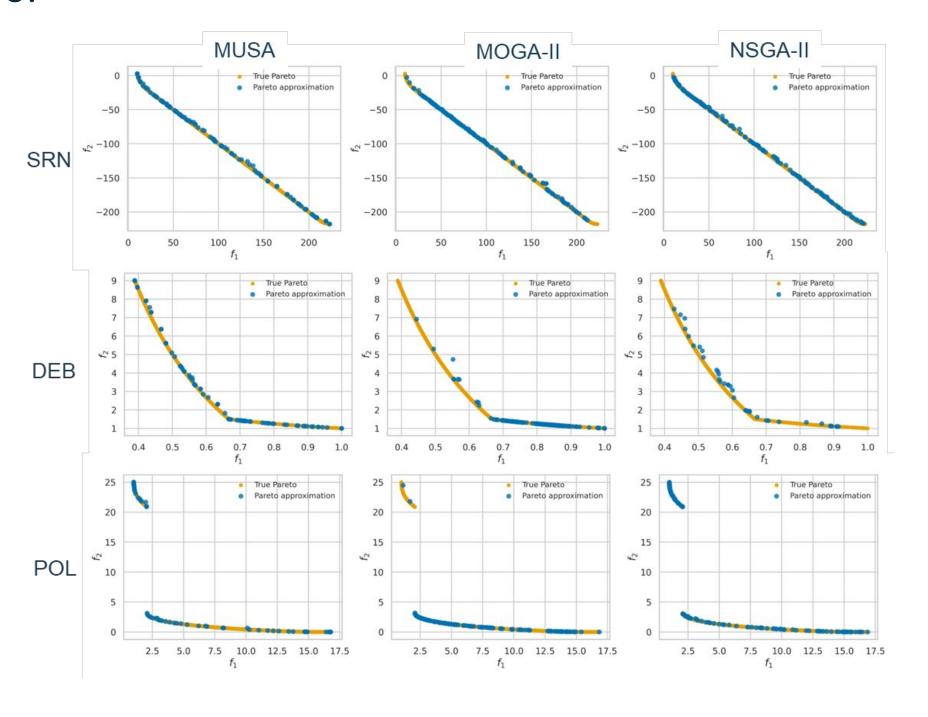






MUSA: 2-dim problem with 2 objectives

MUSA is able to leverage the best behaviors from both MOGA-II and NSGA-II at the same time.

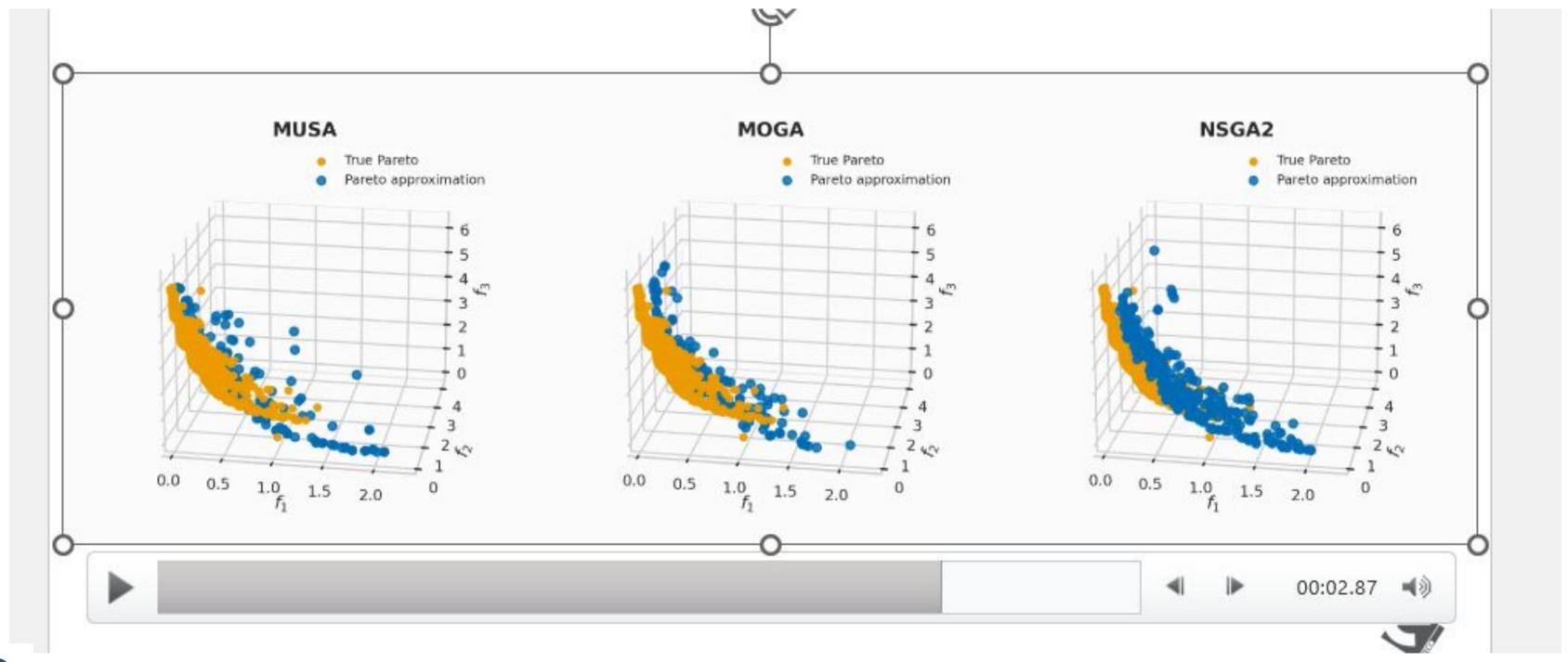






MUSA: 2-dim problem with 3 objectives

This capability is even more evident in high-dimensional problems.







modeFRONTIER Roadmap 2026/2027



Multi-fidelity RSM export

Python and FMU

Parameters in Session tables

Visualize parameters in post processing

Error and exception events in the wf

Workflow-level events management

Automated tagging for script nodes

Matlab and Octave

RSM training in the Planner

Schedule RSM training in your Plan

Connectors manager

Create your own direct interfaces

myRSM

Save, distribute and reuse your Python RSMs inside modeFRONTIER

ROM Training module

Train ROM from modeFRONTIER tables

Multi-fidelity RSM export

Excel



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