



**Intro to resilience modeling, simulation, and
visualization in Python with fmdtools.**

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Overview

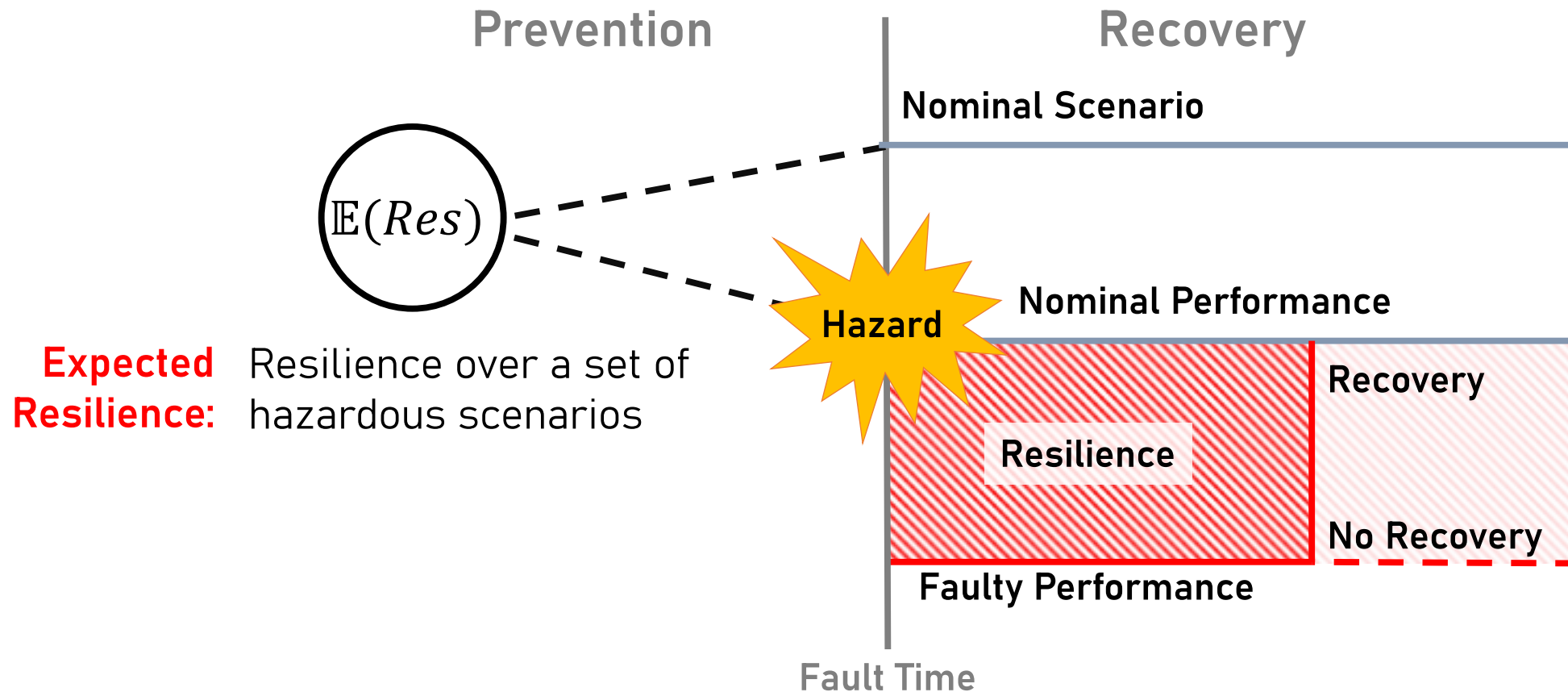
- **Overview of fmdtools**
 - Purpose
 - Project Structure
 - Common Classes/Functions
 - Basic Syntax
- **Coding Activity**
 - Example model: `examples/pump/ex_pump.py`
 - Workbook: `examples/pump/Tutorial_unfilled.ipynb`
 - Model Instantiation
 - Simulation
 - Visualization/Analysis

Prerequisites

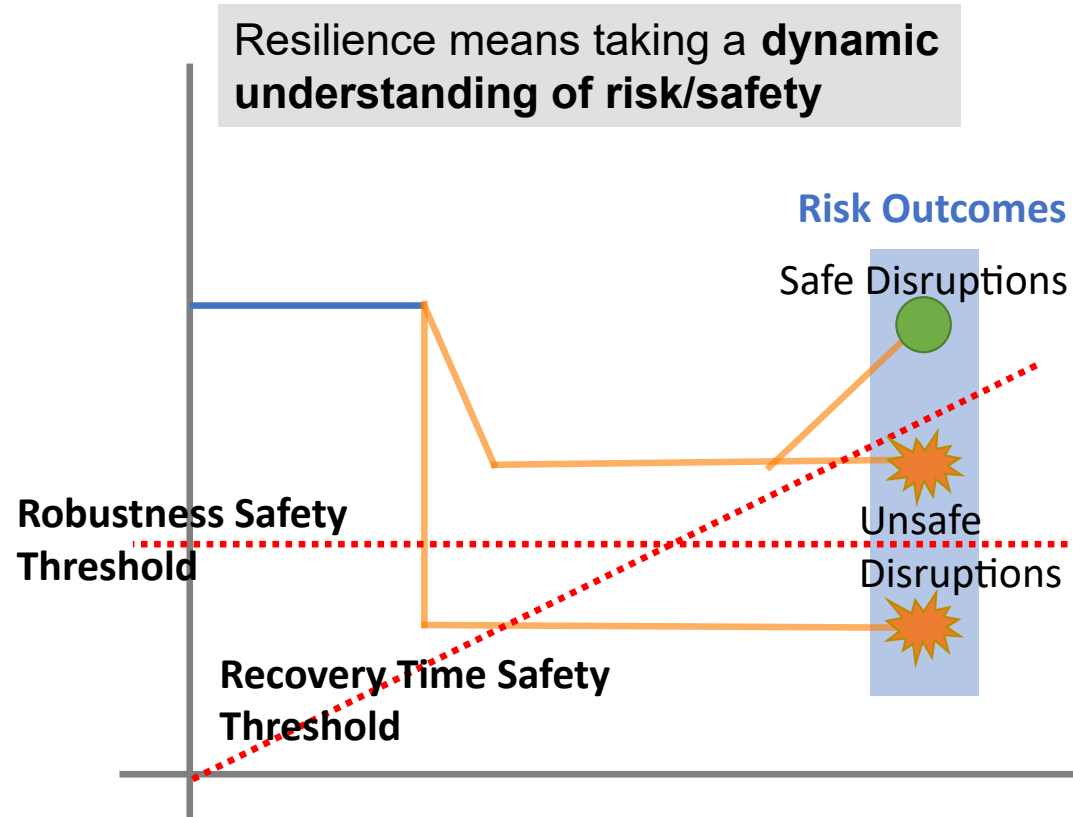
- Ideally, some pre-existing Python and Git knowledge
- Python distribution (anaconda or uv)
 - Ideally this is already set up!
 - Download/install from:
 - <https://www.anaconda.com/products/individual>
 - <https://github.com/astral-sh/uv>
- A git interface
 - [Github Desktop](#) (graphical git environment)
 - [git-scm](#) (stand-alone CLI)

Motivation: Modeling System Resilience

Resilience means taking a dynamic understanding of risk and safety



Why is Resilience Important?



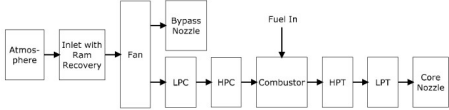
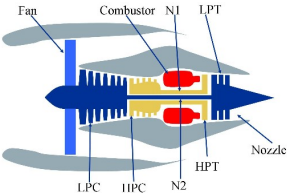
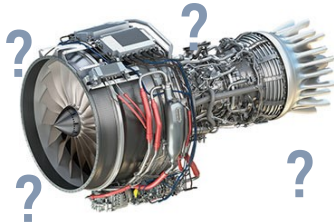
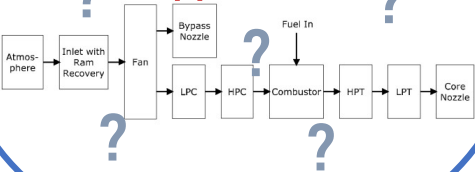
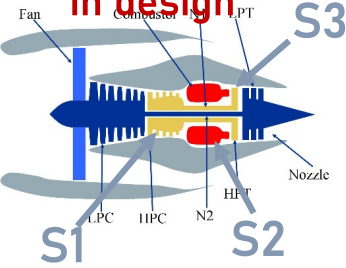
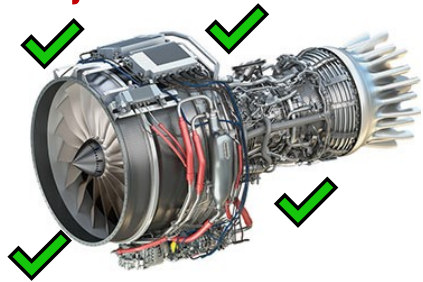
Considering resilience is important when our system has dynamic attributes, e.g.:

- **The system state changes over time**
 - (e.g., position, velocity, etc)
- **We can control this state**
 - (e.g., operators, autopilot)

Because we can use it to determine **how to control the system to a safe outcome** in unsafe circumstances and what **design/operational features** we need to enable this control

Enabling a proactive design process - especially when we don't have data

Idea: the system should be **resilient-by-design**

	Concept Design	Embodiment Design	Implementation
Reactive Design	<p>Choose concept</p> 	<p>Design system</p> 	<p>Retrofit for resilience</p> 
Proactive Design	<p>Establish resilience approach</p> 	<p>Integrate resilient features in design</p> 	<p>Verify resilient function</p> 

Why fmdtools? Possible Competitors:

- Uncertainty Quantification tools: (e.g. OpenCossan)
 - Does not incorporate fault modeling/propagation/visualization aspects
- MATLAB/modelica/etc. Fault Simulation tools
 - Rely on pre-existing model/software stack--Useful, but often difficult to hack/extend (**not open-source**)
- Safety Assessment tools: (e.g. Alyrica, Hip-Hops)
 - Focused on quantifying safety, not necessarily resilience
 - As a result, use **different model formalisms!**

Why fmdtools? Pros:

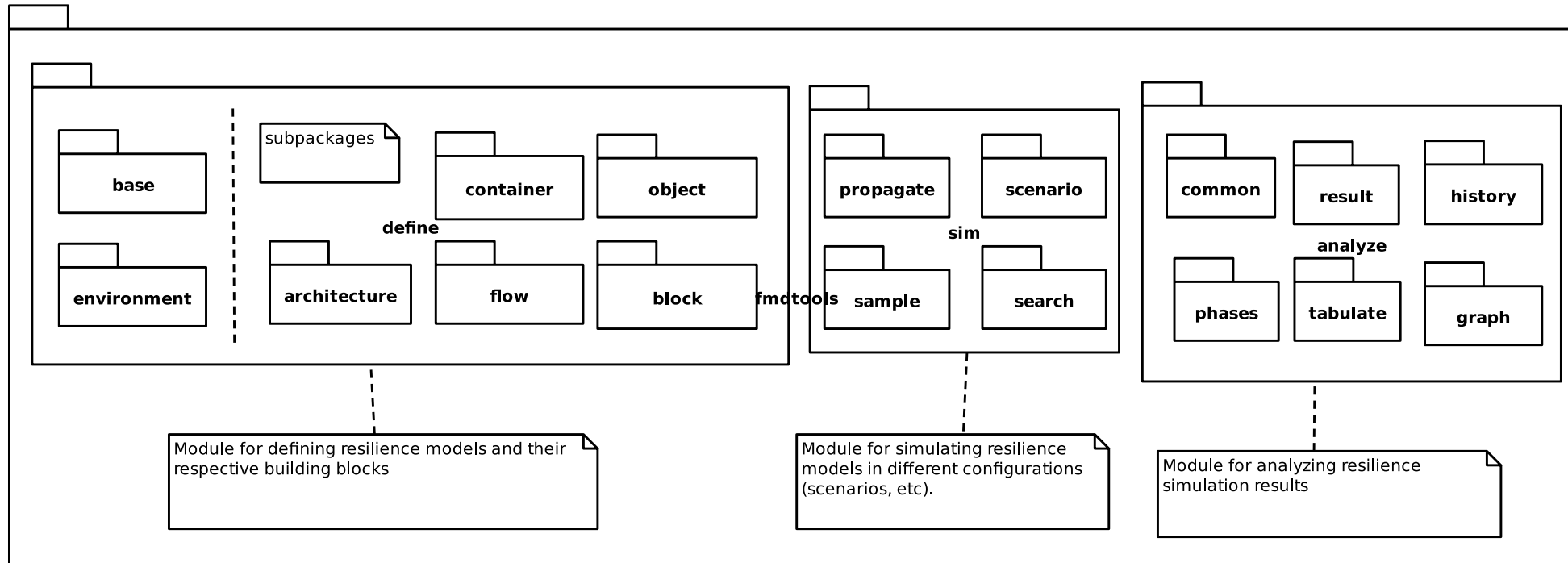
- Highly Expressive, modular model representation.
 - faults from any component can propagate to any other connected component
 - highly-extensible code-based behavior representation
 - modularity enables **complex models** and modelling use-cases
- Research-oriented:
 - Written in/relies on the Python stack
 - Open source/free software
- Enables design:
 - Models can be parameterized and optimized!
 - Plug-and-play analyses and visualizations

Why **not** fmdtools? Cons:

- You already have a pre-existing system model
 - fmdtools models are built in fmdtools
 - if you have a simulink/modelica model, you may just want to use built-in tools
- You want to use this in production
 - fmdtools is Class E Software and thus mainly suitable for research (or, at least, we don't guarantee it)
 - Somewhat dynamic development history

What is fmdtools? A Python package for design, simulation, and analysis of resilience.

pkg module organization



What is fmdtools? Repo Structure

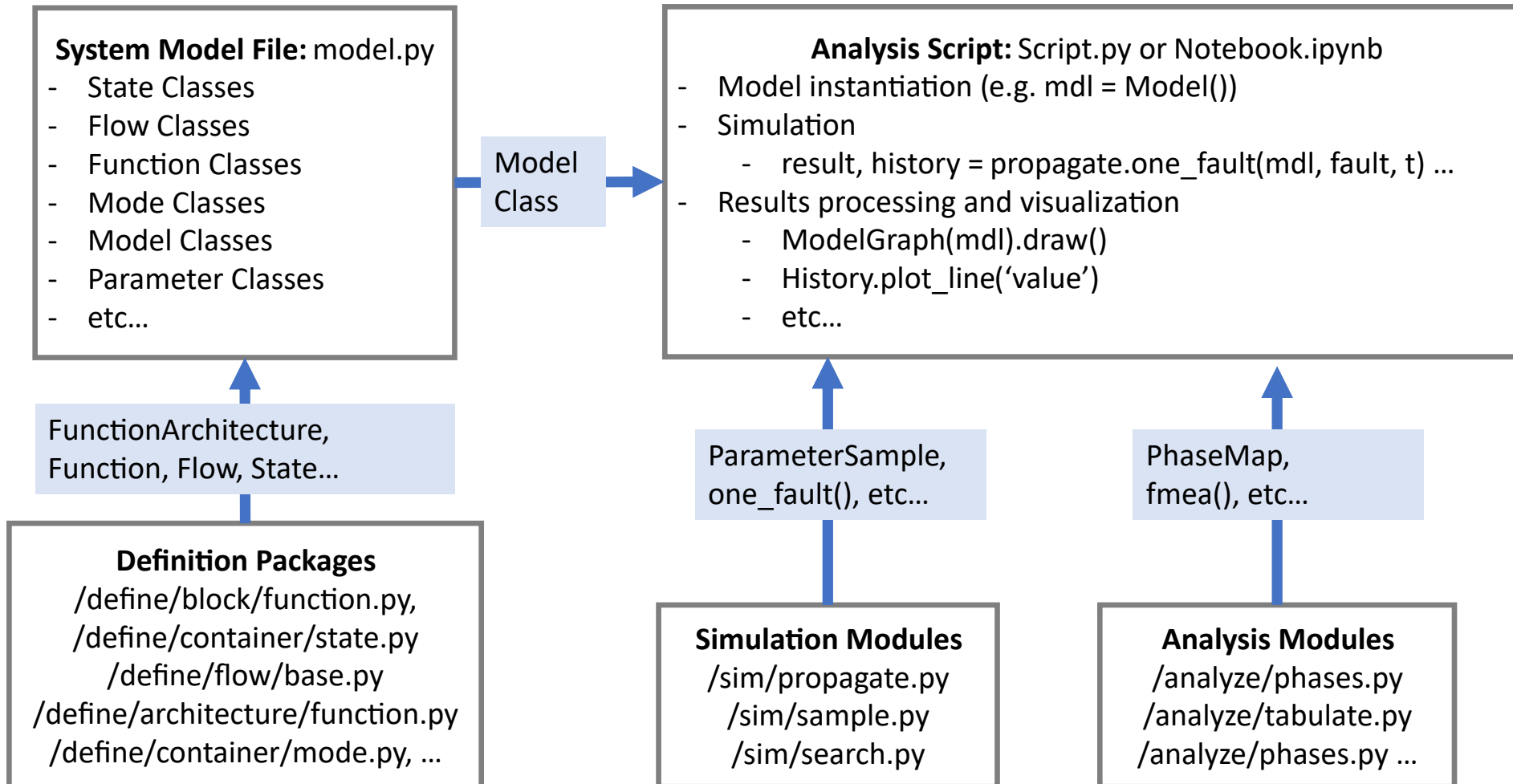
[Repository] (<https://github.com/nasa/fmdtools/>)

- `/fmdtools` : installable package directory
- `/examples` : example models with demonstrative notebooks and tests
- `/docs` : [HTML Documentation](#) (source files at `/docs-source`)
- `/tests` : stand-alone tests (and testing rigs)
- Basic information: `README.md` , `CONTRIBUTORS.md` , `PUBLICATIONS.md` , `LICENSE` , `fmdtools_Individual_CLA.pdf` , etc.
- Config/test files: `requirements.txt` , `pyproject.toml` , `conf.py` , `index.rst` , etc.

Activity: Download and Install fmdtools

- repo link: <https://github.com/nasa/fmdtools/>
- set up repo:
 - create `path/to/fmdtools` folder for repo
 - (usually in `/documents/GitHub`)
 - clone git into folder:
 - `git clone https://github.com/nasa/fmdtools.git`
 - can also use webpage
- package installation:
 - anaconda: Open Python from anaconda (e.g., open Spyder) and install with `pip install -e /path/to/fmdtools`
 - uv: run `uv pip install .` from fmdtools repository

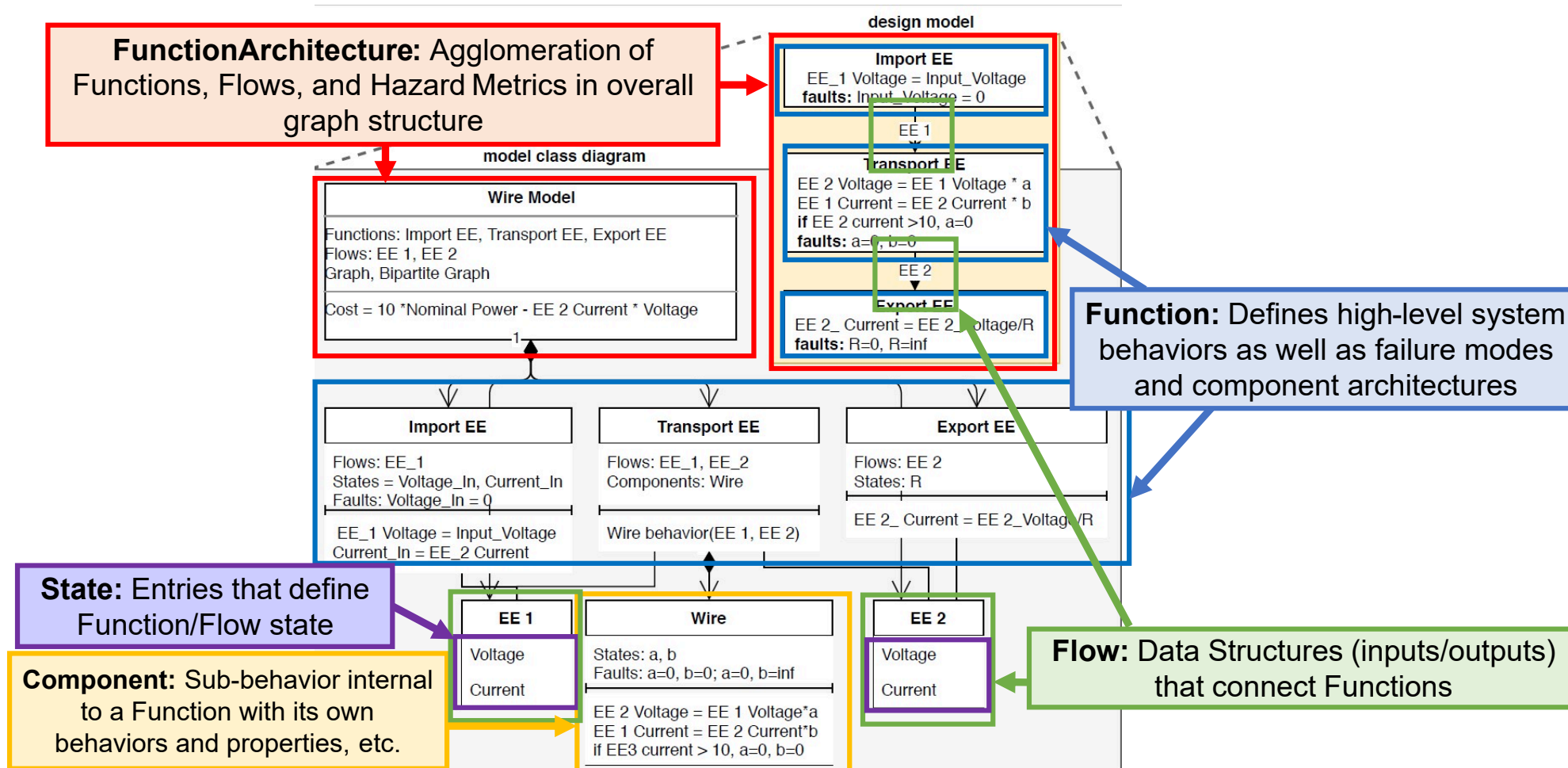
Analysis Workflow/Structure



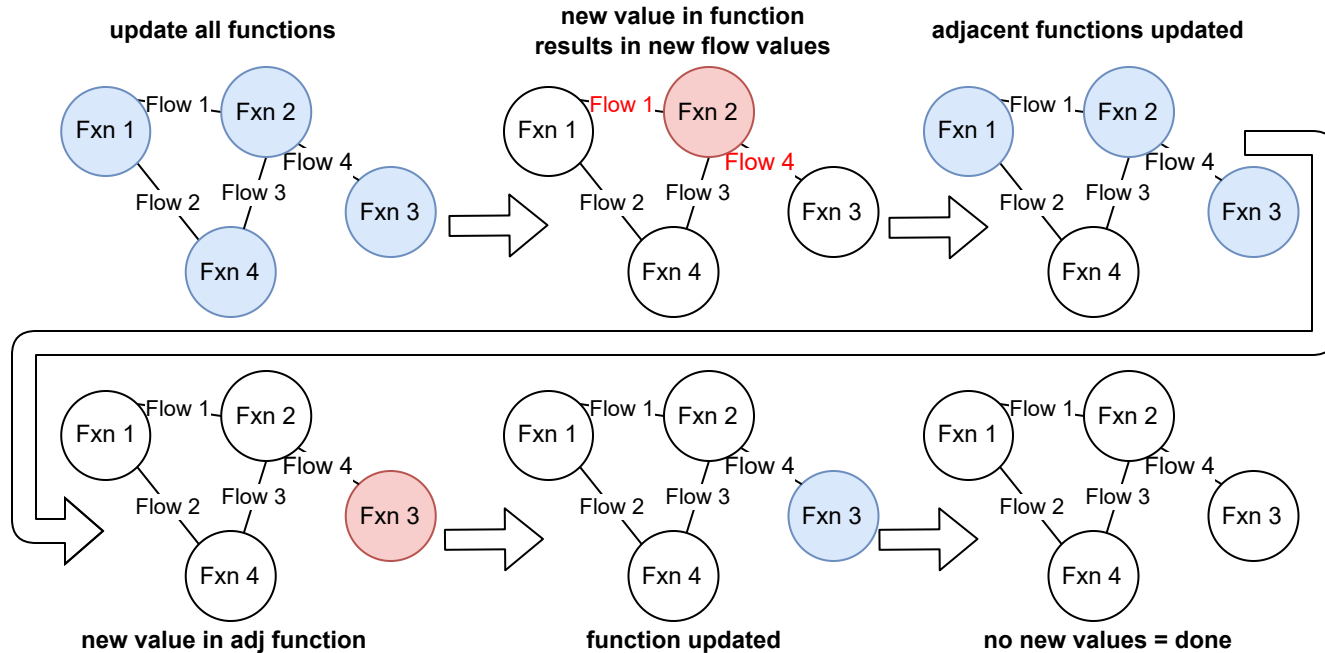
Defining a Model

- What do we want out of a model?
 - What behaviors and how much fidelity do we need?
 - What **functions/components** and interactions make up the system?
 - One function or multiple functions?
 - Is it a **controlled system**? Are there multiple **agents**?
- What type of simulation do we want to run?
 - Single-timestep vs multi-timestep vs network
- What scenarios do we want to study and how?
 - **Failure modes** and faulty behaviors
 - **Disturbances** and changes in parameters
 - What are the possible effects of hazards and how bad are they?
 - By what **metrics**?

Defining a Model



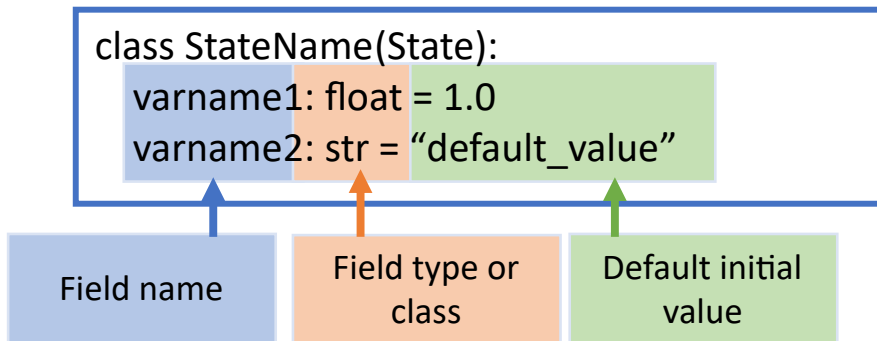
Concept: Static Propagation



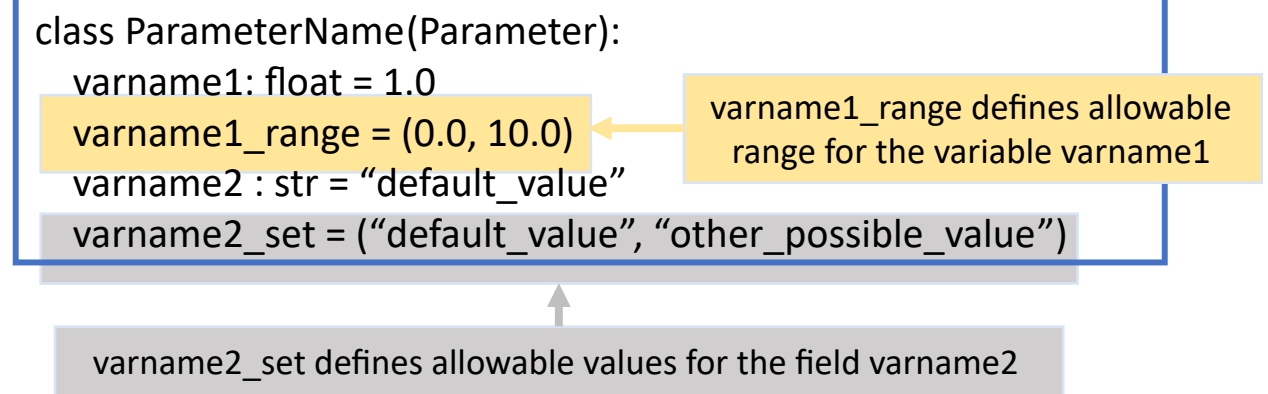
In a single timestep, functions with `static_behavior()` methods simulate until behaviors converge (i.e., no new state values)

Containers - The building blocks of simulations

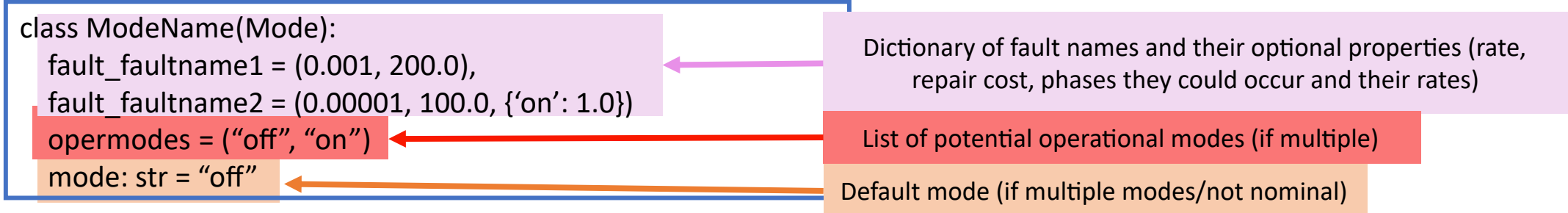
State classes are used to represent variables (called fields) that change over time



Parameter classes are used to represent variables that don't change over time, with similar syntax to States



Mode classes are used to represent modes (faults and operational modes) that could occur in the system



Flow Code Template

```
class FlowName(Flow):
```

```
    container_s = StateName
```

```
    container_p = ParameterName
```

```
    default_track = ['s','m']
```

```
    def indicate_XXX(self):
```

```
        Conditional statement (e.g., self.s.state>threshold) which  
        is logged in the history and may be used to terminate  
        simulations
```

Specifies which container classes play specific Flow **roles** (e.g., s corresponds to the state role, p corresponds to a parameter role, m corresponds to the mode role, etc)

Specifies **what should be tracked** in the FxnBlock history (fxnname.h) by default. May be a dict ({role:value}), list ([role1, role2]), or string ("all","none", etc). Overwritten by track parameter at model instantiation.

These methods define Flow **indicators** and are called/tracked during simulation (in flow.i)

- Flows represent connections or shared variables between different functions. Think of them as Function inputs/outputs.
- Flows are build from container classes like states, along with their own methods/variables.

Function Code Template

flow_XXX is used to append a **flow** of given type that is named XXX to the function class. If the flow(s) has a different name outside the function (optional), flownames matches the external name to the internal name

Specifies **what should be tracked** in the FxnBlock history (fxnname.h) by default. May be a dict ({role:value}), list ([role1, role2]), or string ("all","none", etc). Overwritten by the track parameter during model instantiation.

These methods define the **behavior** of the FxnBlock and thus simulate at each time-step of the simulation.

These methods define FxnBlock **indicators** and are called/tracked during simulation (in fxnname.i)

This method defines the Result to be returned **when simulated individually**

```
class FunctionName(Function):
```

```
    container_s = FunctionState
    container_m = FunctionMode
    container_t = FunctionTime
```

Specifies which classes play specific FxnBlock **roles** (e.g., s corresponds to the state role, m corresponds to the mode role, etc)

```
    flow_flowname1 = FlowClass1
```

```
    flownames = {"outsideflowname": "flowname1"}
```

```
    default_sp = {'end_time': 100}
```

Default **keyword arguments for SimParam**. Only necessary when the functionblock will be simulated individually.

```
    default_track = ['s','m']
```

```
    def init_block(self, **kwargs):
```

```
        <e.g., self.s.x = 2.0>
```

Optional method to call to set up FxnBlock in ways not already defined by role initialization (e.g., attaching local MultiFlows or setting initial values for States from Parameter)

```
    def static_behavior(self):
```

```
        Runs only in static propagation steps
```

```
    def dynamic_behavior(self):
```

```
        Runs only in dynamic propagation steps
```

```
    def indicate_XXX(self):
```

```
        Conditional statement (e.g., self.s.state>threshold) which is logged in the history and may be used to terminate simulations
```

```
    def find_classification(self, scen, mdlhists):
```

```
        Returns a Result dictionary (calculated at completion)
```

Function Architecture Code Template

Default **keyword arguments for SimParam**.
Defines max time of the simulation, along with phases, timestep, units, etc.

Method to **instantiate the model and define its structure**:

- **.add_flow** is used to instantiate a flow
- **.add_fxn** is used to instantiate a function and attach connected flows

These methods define Model **indicators** and are called/tracked during simulation (in modelname.i)

This method defines the Result to be returned by the model.

```
class ArchitectureName(FunctionArchitecture):
```

```
    container_p= ModelParam
```

```
    default_sp = {'end_time': 100}
```

```
    default_track = ["fxns", "flows"]
```

```
    def init_architecture(self, **kwargs):
```

```
        self.add_flow("flowname", FlowClass)
```

```
        ...
```

```
        self.add_fxn("functionname", FunctionClass, "flowname1", "flowname2")
```

```
        ...
```

```
    def indicate_XXX(self):
```

```
        Conditional statement
```

```
        (e.g., self.fxn["functionname"].s.state>threshold)
```

```
        which is logged in the history and may be used to terminate simulations
```

```
    def find_classification(self, scen, mdlhists):
```

```
        Returns a Result dictionary (calculated at completion)
```

Points to a Parameter representing immutable model characteristics instantiated at the start of the simulation

Specifies **what should be tracked** in the Model history by default. May be a dict ({role:value}), list ([role1, role2]), or string ("all","none", etc). Overwritten by the track option in propagate.

Demo Model Activity: examples/pump/ex_pump.py

Notice the definitions and structure:

- **States:** WaterStates , EEStates , SignalStates
- **Flows:** Water , EE , Signal
- **Functions:** ImportEE , ImportWater , ExportWater , MoveWater , ImportSignal
 - **Modes** (e.g., ImportEEMode , ImportSigMode)
 - Mode probability model
 - Actual modes in fm_args entry
 - others attributes, e.g., Timer
- **Model:** Pump connects functions, flows, and defines end_classification
- **Parameter:** PumpParam defines values we can change in the simulation

More Resources for Model Definition

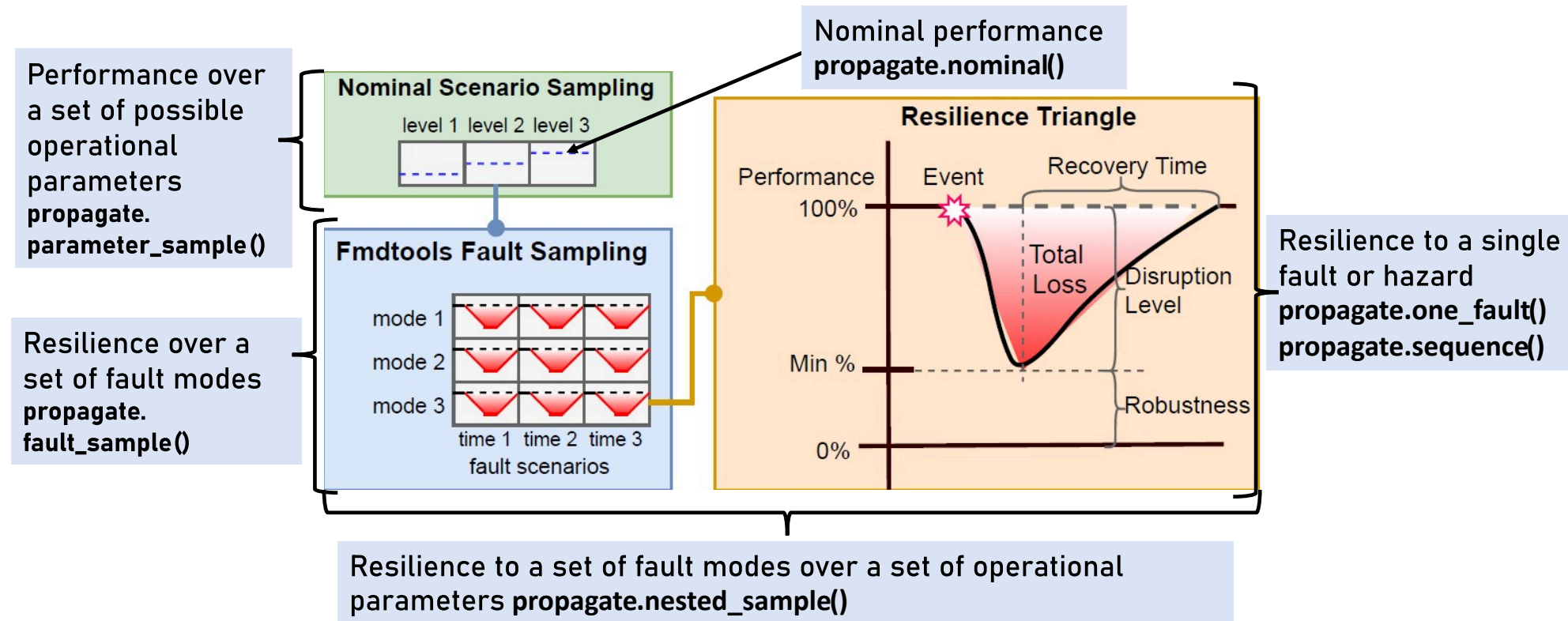
- Note the docs for model definition are in <https://nasa.github.io/fmdtools/docs-source/fmdtools.define.html>
- Other examples also can be helpful:
<https://nasa.github.io/fmdtools/examples/Examples.html>

Notebook Activity:

Open `/examples/pump/Tutorial_unfilled.ipynb` :

- Instantiate the model
 - `mdl = Pump()`
- Explore structure
 - Try different parameters!
 - Change things!
What does the model directory look like?
 - `dir(mdl)`

Simulation Concepts: Types of Simulations



For more info on syntax/arguments, see documentation for [fmdtools.sim.propagate](#).

Simulation Concepts: Sampling Approaches

These classes define **multi-run simulations** which can be used to quantify uncertain performance/resiliences:

- **SampleApproach/FaultSample**: Which faults to sample and when
 - Relies on **mode** information encoded in the model
 - Simulated using `propagate.fault_sample()`
- **ParameterSample**: Nominal parameters or random seeds to sample
 - Can be simulated in `propagate.parameter_sample()`
 - Can be simulated in conjunction with faults using `propagate.nested_sample`

See docs for: `fmdtools.sim.fault_sample`

Simulation Concepts: Things to Consider

Static/Dynamic propagation: How function states propagate to each other in a single time-step and multiple time-steps?

Stochastic Propagation: Whether and how stochastic states are instantiated over time

- e.g. do we run with the “default” values of parameters, or do we sample from a random number generator?

Breadth of Scenarios: How hazards are represented as discrete scenarios to simulate

- What set of joint faults do we use? How many times are sampled?
- Operational scenarios and joint operational/fault scenarios

Activity: Simulate the Model

Run fault propagation methods:

- `propagate.nominal()` , `propagate.one_fault()` , `propagate.fault_sample()`

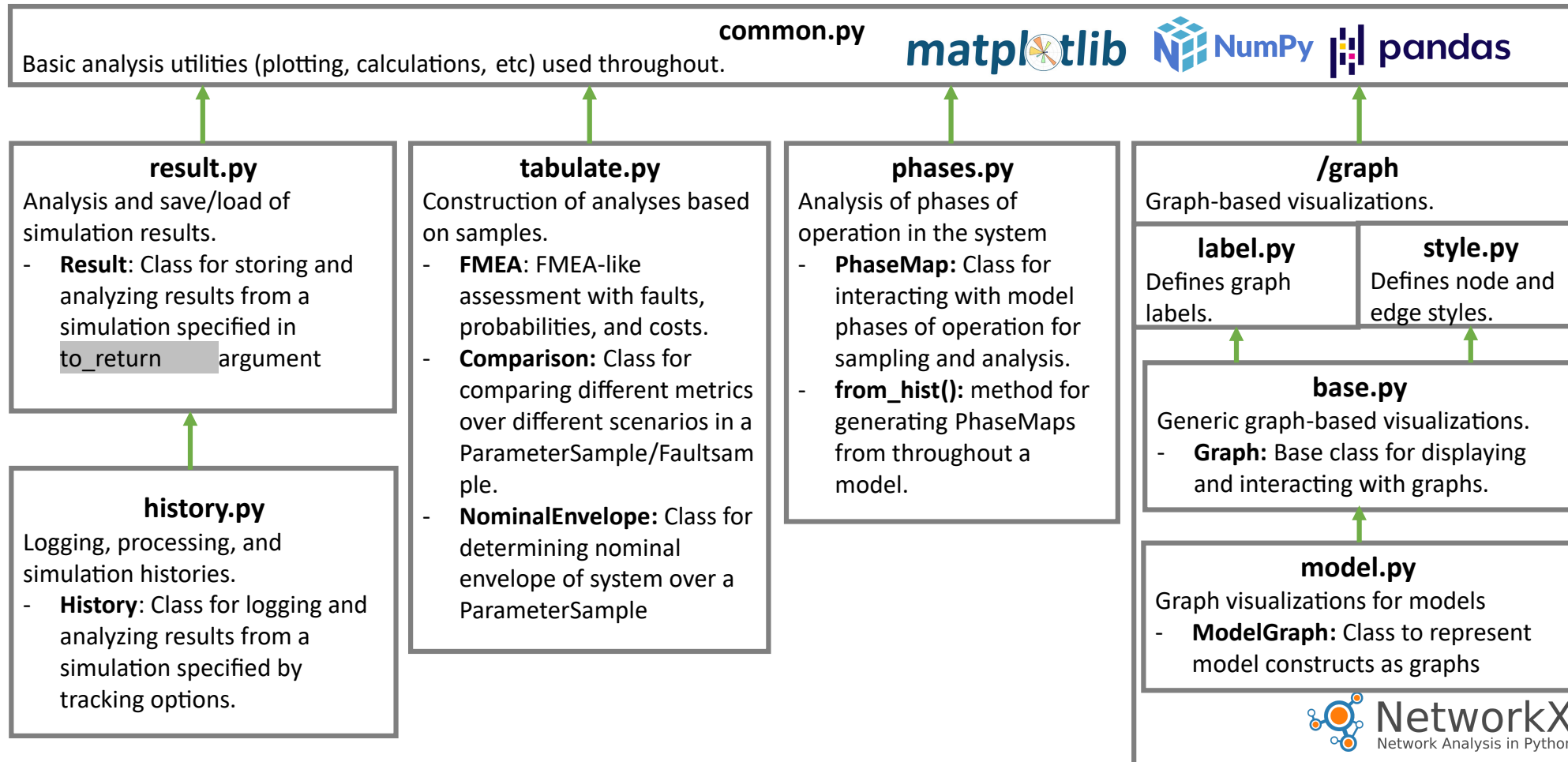
What do the results look like? Explore data structures:

- `analyze.result.Result` , `analyze.result.History`

Explore:

- What happens when you change `FaultSample` parameters?
- What happens when you change `Model` parameters?
- How do these methods compare in terms of computational time?

Analysis Modules - see docs for `fmdtools.analyze`



Analysis Activity

Visualize the results:

- Show model graph
- Show nominal performances
- Show performances in a nominal scenario
- Make a scenario-based FMEA

Explore:

- How can you show only the parameters you want? Or change the formatting?
- What does the behavior under other faults look like?
- What other analyses can you perform with these results?

Conclusions/Summary

- **fmdtools** is an environment for designing resilient systems
 - `/define` enables model definition
 - `/sim` is used to define simulations
 - `/analyze` is used to analyze and visualize simulation results
- I hope you agree that it has some powerful features!
 - Modeling expressiveness and clarity
 - Types of simulations that can be run
 - Powerful but easy-to-leverage plug-and-play analyses

Further Reading/Links

- More advanced topics ([see examples](#)), including Search and optimization, Human, Systems-of-Systems modeling, and Modeling Stochastic Behavior
- [Model Development Guide](#): Has best practices for developing models in a strategic way (especially helpful for complex models)
- Overview Paper:
 - [Hulse, D., Walsh, H., Dong, A., Hoyle, C., Tumer, I., Kulkarni, C., & Goebel, K. \(2021\). fmdtools: A fault propagation toolkit for resilience assessment in early design. International Journal of Prognostics and Health Management, 12\(3\).](#)