

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

**Author: Daniel Hulse** 

Version: 2.1.3

#### **Overview**

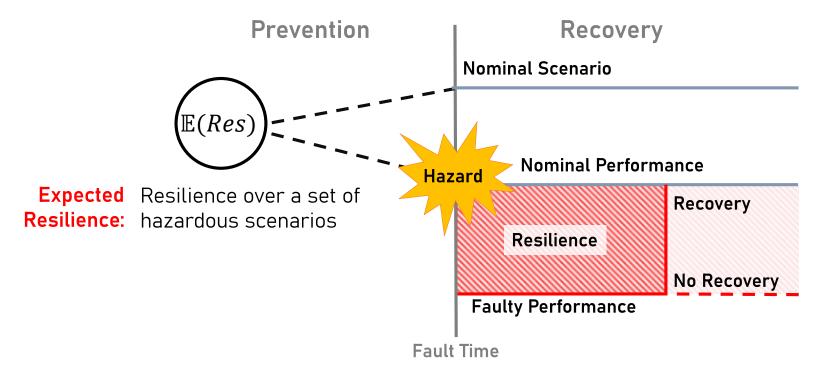
- Overview of fmdtools
  - Purpose
  - Project Structure
  - Common Classes/Functions
  - Basic Syntax
- Coding Activity
  - o Example model: examples/pump/ex\_pump.py
  - O Workbook: examples/pump/Tutorial\_unfilled.ipynb
    - Model Instantiation
    - Simulation
    - Visualization/Analysis

#### **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)

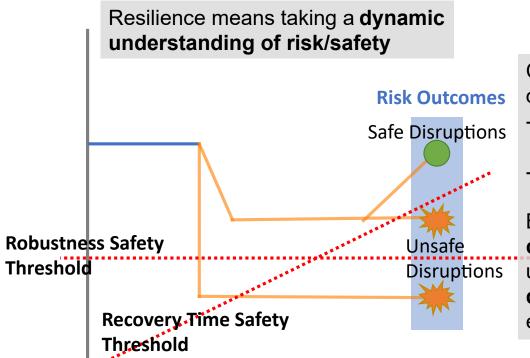
# Motivation: Modeling System Resilience

Resilience means taking a dynamic understanding of risk and safety



Yodo N., & Wang, P. (2016).

## Why is Resilience Important?

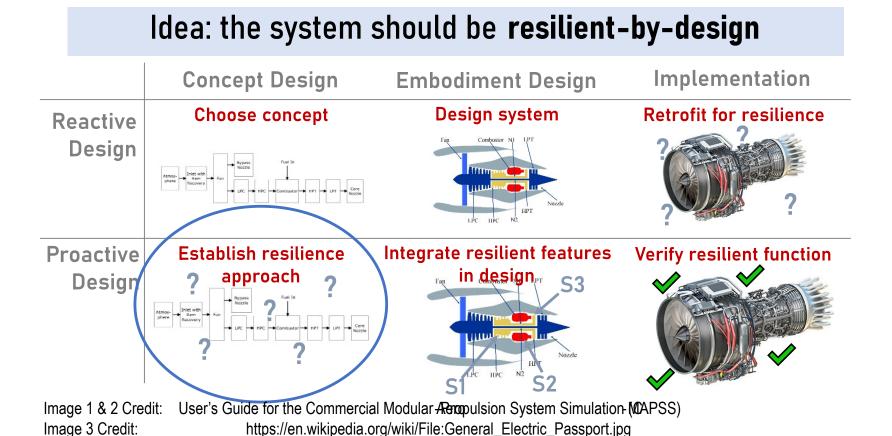


Considering resilience is important when our system has dynamic attributes, e.g.:

- 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 outcomen unsafe circumstances and what design/operational features we need to enable this control

### **Enabling proactive design process**



- Capacially relayant to have exeterned when we don't have date

# 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:

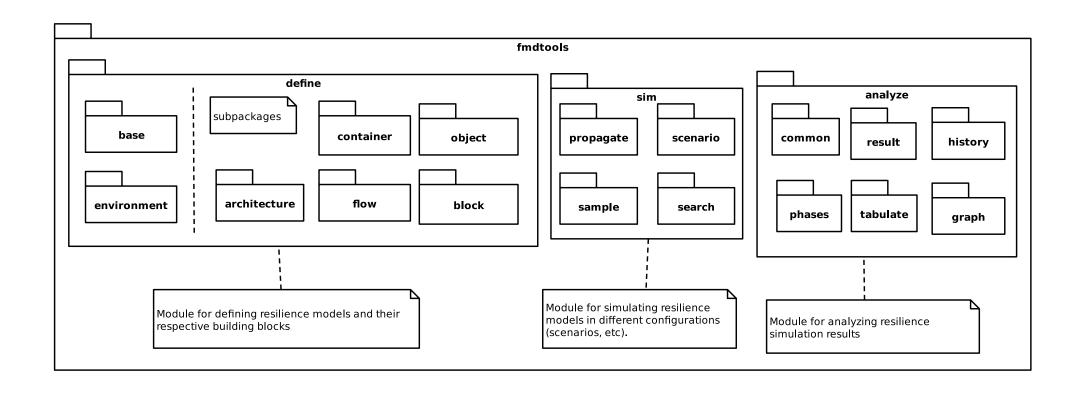
- Highly Expressive, modular model representation.
  - faults from any component can propagate to any other connected component via undirected propagation
  - highly-extensible code-based behavior representation
  - class structure enables complex models representing human behavior and systems of systems
- 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
  - o 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

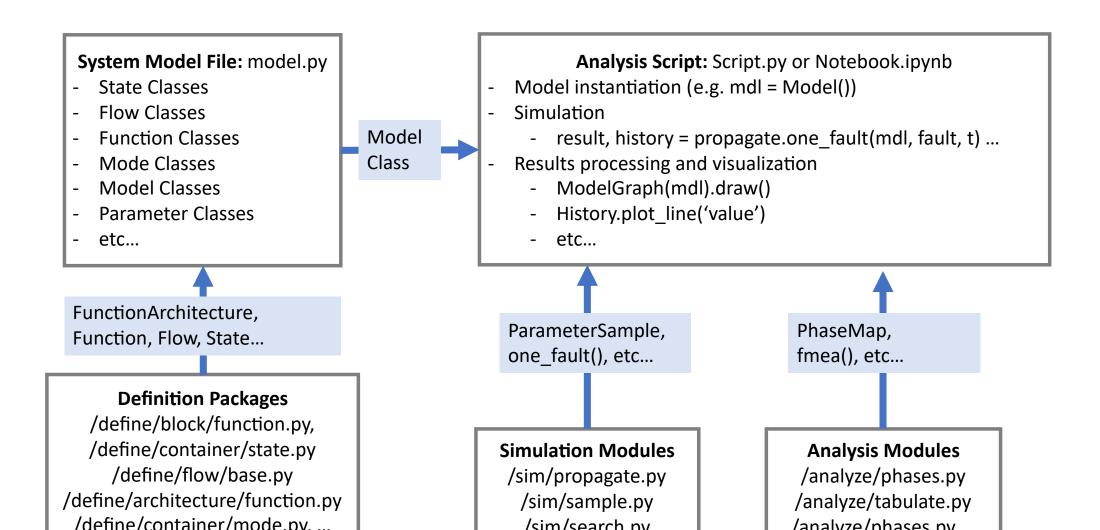
[Repository] (https://github.com/nasa/fmdtools/)

- /fmdtools : installable package
- /examples : example models with demonstrative notebooks and tests
- /docs : resources for documentation
- /tests : stand-alone tests (and testing rigs)
- README.md: Basic package description
- CONTRIBUTORS.md: Credit for contributions
- requirements.txt : List of requirements
- ... and other configuration files

#### **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
  - o uv: run uv pip install . from fmdtools repository

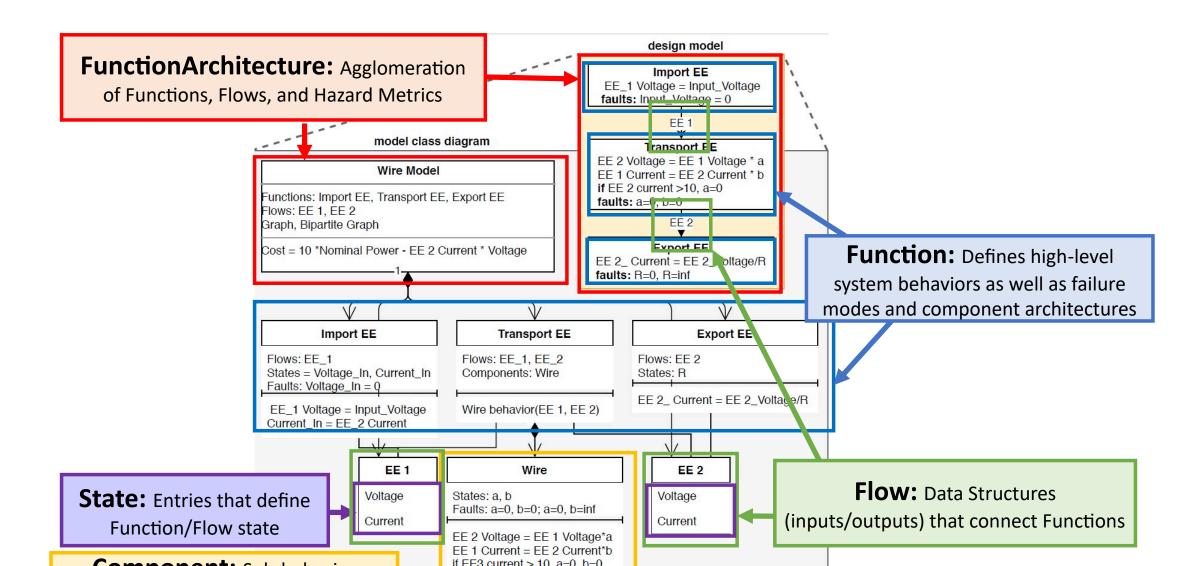
#### **Analysis Workflow/Structure**



# **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?
    - Single function or multiple functions?
    - Is it controlled? 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**



## 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

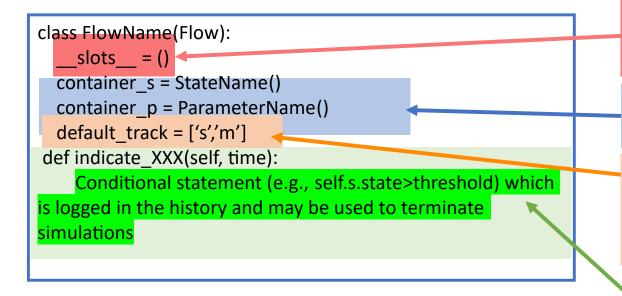
```
class ParameterName(Parameter):
    varname1: float = 1.0
    varname1_range = (0.0, 10.0)
    varname2 : str = "default_value"
    varname2_set = ("default_value", "other_possible_value")

varname2_set defines allowable values for the field varname2
```

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

# Flow Code Template

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Classes in fmdtools rely heavily on \_\_slots\_\_ for performance and type-safety. If no non-standard attributes are added, leave this blank.

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 the track parameter during 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

### **Function Code Template**

class FunctionName (Function): Tuple of flows must be specified in slots slots =('flowname1',) flow XXX is used to append a **flow** of container s = FunctionState given type that is named XXX to the Specifies which classes play specificFxnBlock roles (e.g., s container m = FunctionMode function class. If the flow(s) has a corresponds to the state role, m corresponds to the mode container t = FunctionTime different name outside the function role, etc) (optional), flownames matches the flow flowname1 = FlowClass1 external name to the internal name Default keyword arguments for SimParam. flownames = {"outsideflowname":"flowname1"} Only necessary when the function block will default sp = {'end time': 100} be simulated individually. Specifies what should be tracked in the default track = ['s','m'] FxnBlock history (fxnname.h) by default. **Optional** method to call to set up FxnBlock in def init block(self, \*\*kwargs): May be a dict ({role:value}), list ([role1, ways not already defined by roleinitalization <e.g., self.s.x = 2.0>role2]), or string ("all","none", etc). (e.g., attaching local MultiFlows or setting def static behavior (self, time): Overwritten by the track parameter during initial values for States from Parameter) Runs only in static propagation steps model instantiation. def dynamic behavior (self, time): Runs only in dynamic propagation steps These methods define the **behavior** of def behavior(self, time): the FxnBlock and thus simulate at each Runs in static propagation step (same as static behavior) time-step of the simulation. def condfaults(self, time): Runs in both static and dynamic propagation steps prior to behaviors and internal fault propagation (to components and actions) These methods define ExnBlock def indicate XXX(self, time): indicators and are called/tracked Conditional statement (e.g., self.s.state>threshold) which is logged in during simulation (infxnname.i) the history and may be used to terminate simulations def find classification (self, time):

### **Model Code Template**

This method defines the Result to be

returned by the model.

class ArchitectureName (FunctionArchitecture): Architecture classes are usually given empty slots \_\_slots\_\_ = () Default keyword arguments for SimParam. container p = ModelParam Points to a Parameter representing immutable model Defines max time of the simulation, along characteristics instantiated at the start of the simulation default sp = {'end time': 100} with phases, timestep, units, etc. default track = ["fxns", "flows] Specifies what should be tracked in the Model history by default. May be adict ({role:value}), list ([role1, role2]), or string ("all","none", etc). def init architecture(self, \*\*kwargs): Method to instantiate the model and define Overwritten by the track option in propagate. self.add flow("flowname", FlowClass) its structure:. .add flow is used to instantiate a flow self.add fxn("functionname", FunctionClass, "flowname1", "flowname2") .add fxn is used to instantiate a function and attach connected flows def indicate XXX(self, time): These methods define Model Conditional statement indicators and are called/tracked (e.g., self.fxns["functionname"].s.state>threshold) during simulation (in modelname.i) which is logged in the history and may be used to terminate simulations def find classification (self, time):

Returns a Result dictionary (calculated at completion)

## 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
  - Flows
  - O 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

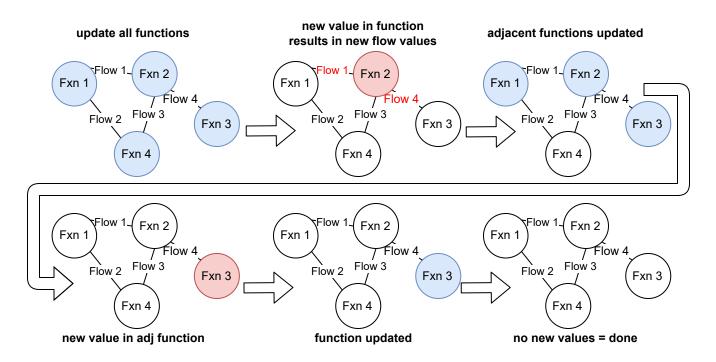
- Note the docs for model definition are in https://nasa.github.io/fmdtools/docs-source/fmdtools.define.html
- Other examples also can be helpful: https://nasa.github.io/fmdtools/examples/Examples.html

#### **Notebook Activity:**

Open /examples/pump/Tutorial\_unfilled.ipynb :

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

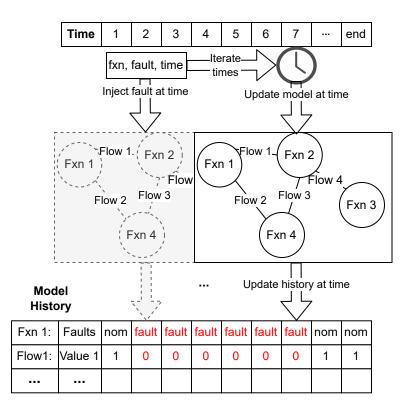
# Simulation Concepts: Static/Undirected Propagation



#### In a single timestep:

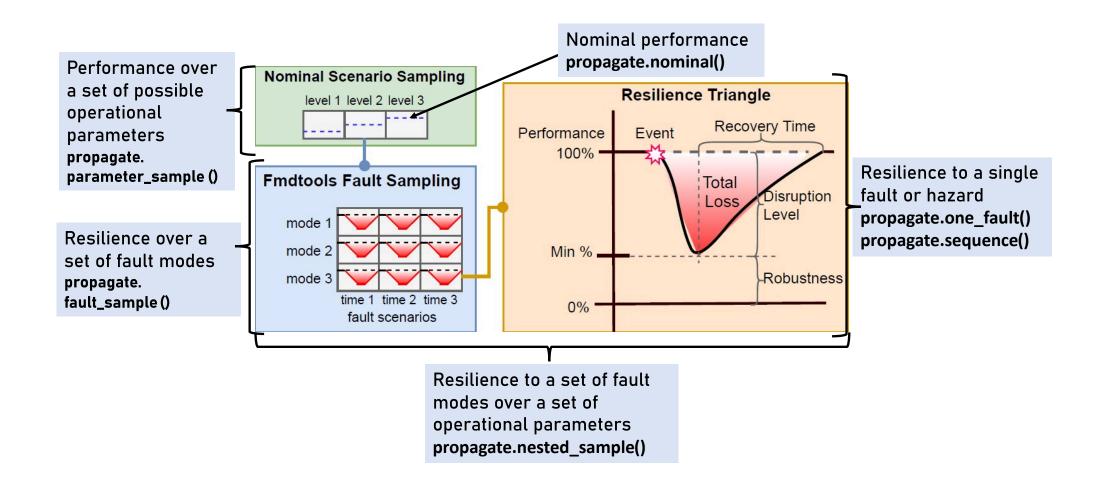
- Functions with static\_behavior() methods simulate until behaviors converge (i.e., no new state values)
- Functions with dynamic\_behavior() run once in defined order

### Simulation Concepts: Propagation over Time



 Model increments (simulated + history updated) over each time-step until a defined final time-step or specified indicator returns true.

## Simulation Concepts: Types of Simulations



# 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

 Undirected graph representation—states can effect all other connected states, and vice versa, in any order

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()

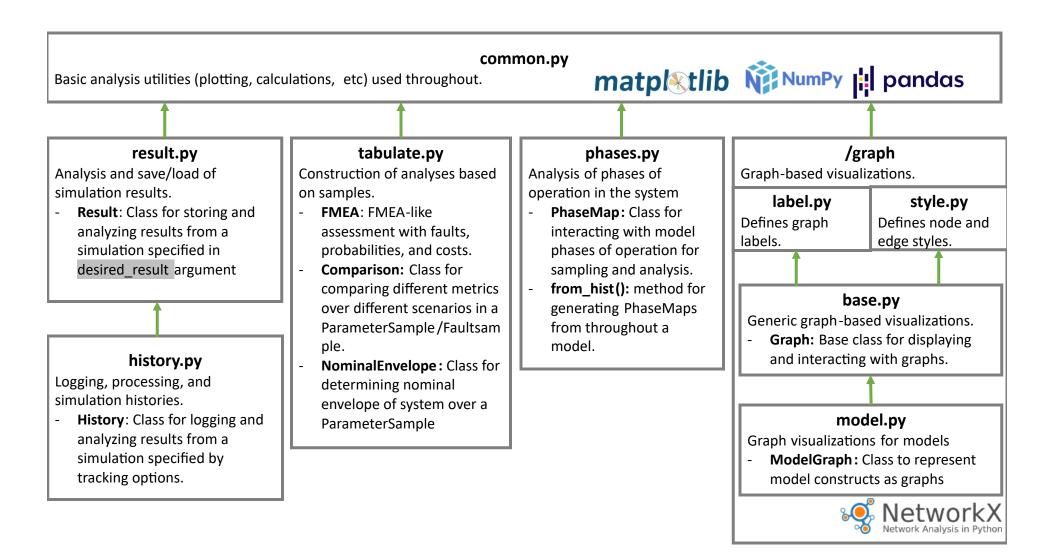
What do the results look like? Explore data structures:

- analyze.result.Result
- analyze.result.History

#### Explore:

- What happens when you change FaultSample parameters?
- What happens when you change Model parameters?

# **Analysis Modules**



# **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):
  - Search and optimization
  - Human/Al Modeling
  - Systems-of-Systems modeling
  - Modeling Stochastic Behavior
  - ... and more
- 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).