Solving Interoperability Between Digital Twins

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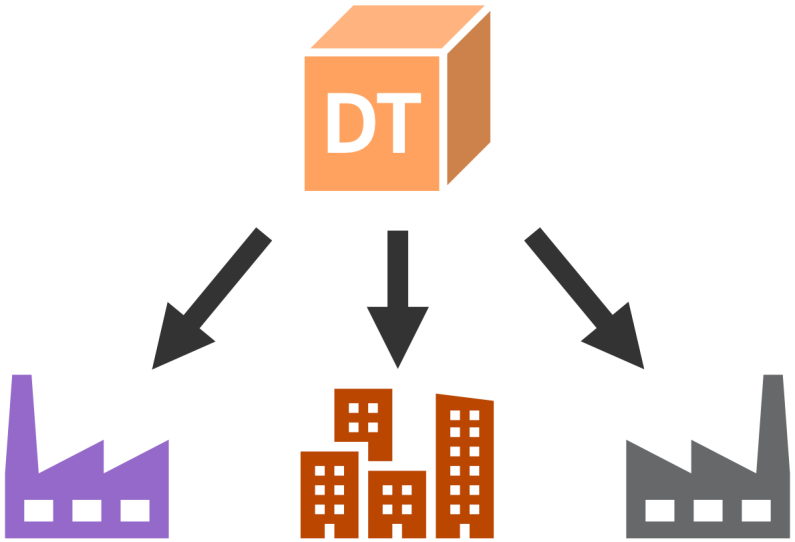
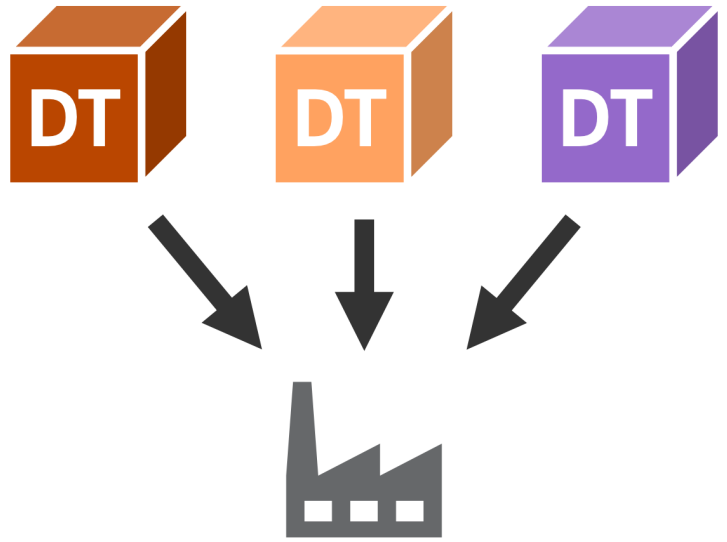
Off-the-shelf Digital Twins must be interoperable*

Maintains meaning in different contexts



Maintains meaning with different DTs

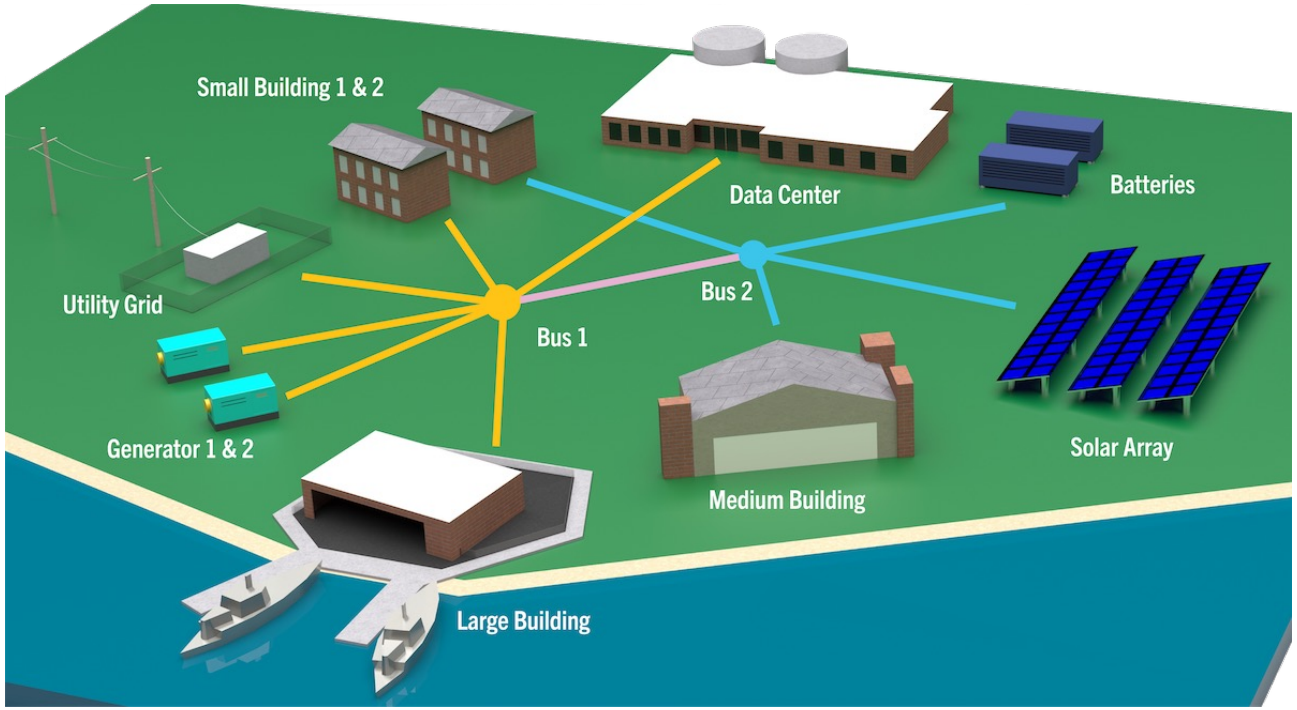


*Similar terms include composable, extensible, interactive, modular, etc.

Requires multiple-front approach to interoperability

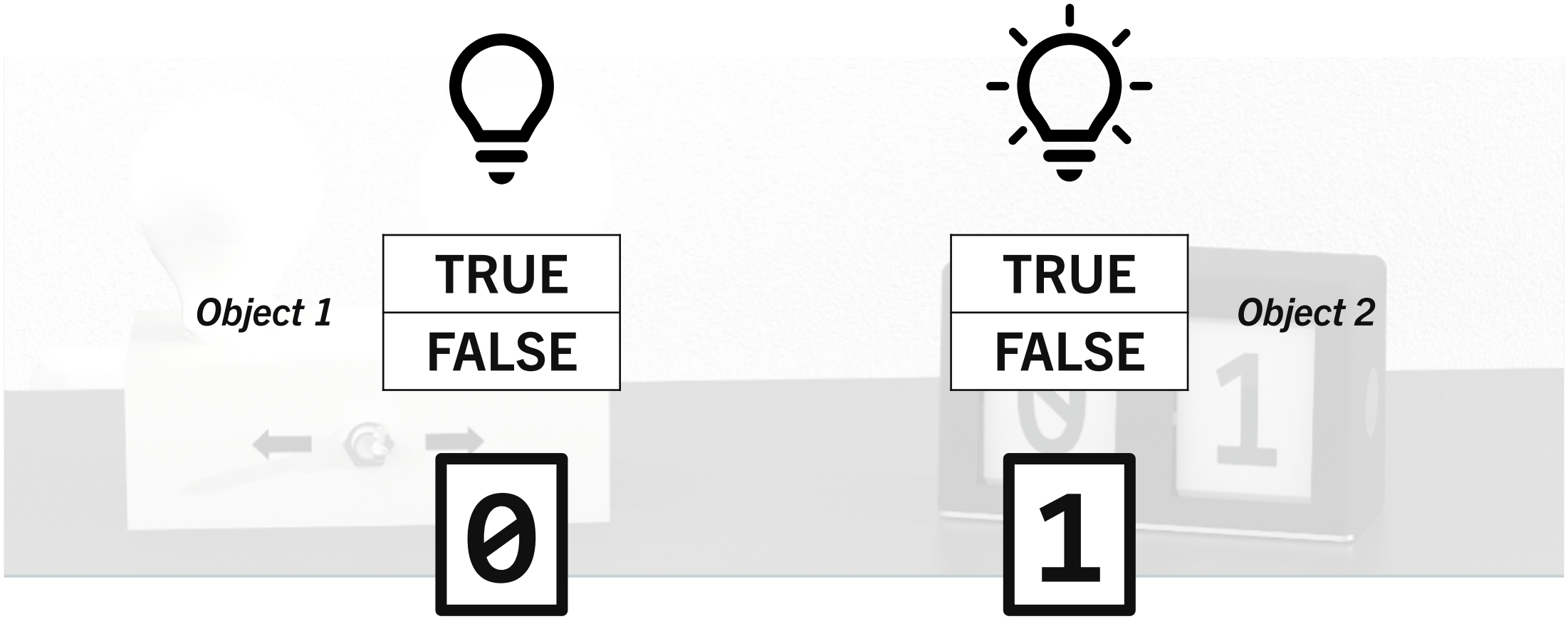
Syntactic Interoperability:
The definition of the system is truthful across all interactions

Semantic Interoperability:
The behavior of the system is truthful across all interactions



DTs are systems synchronized with another system of interest

Both systems are scoped to a similar set of variables



DTs provide observations for the system facts

System of interest

Digital Twin

Synchronized DTs provide observations for measured system facts

System of interest

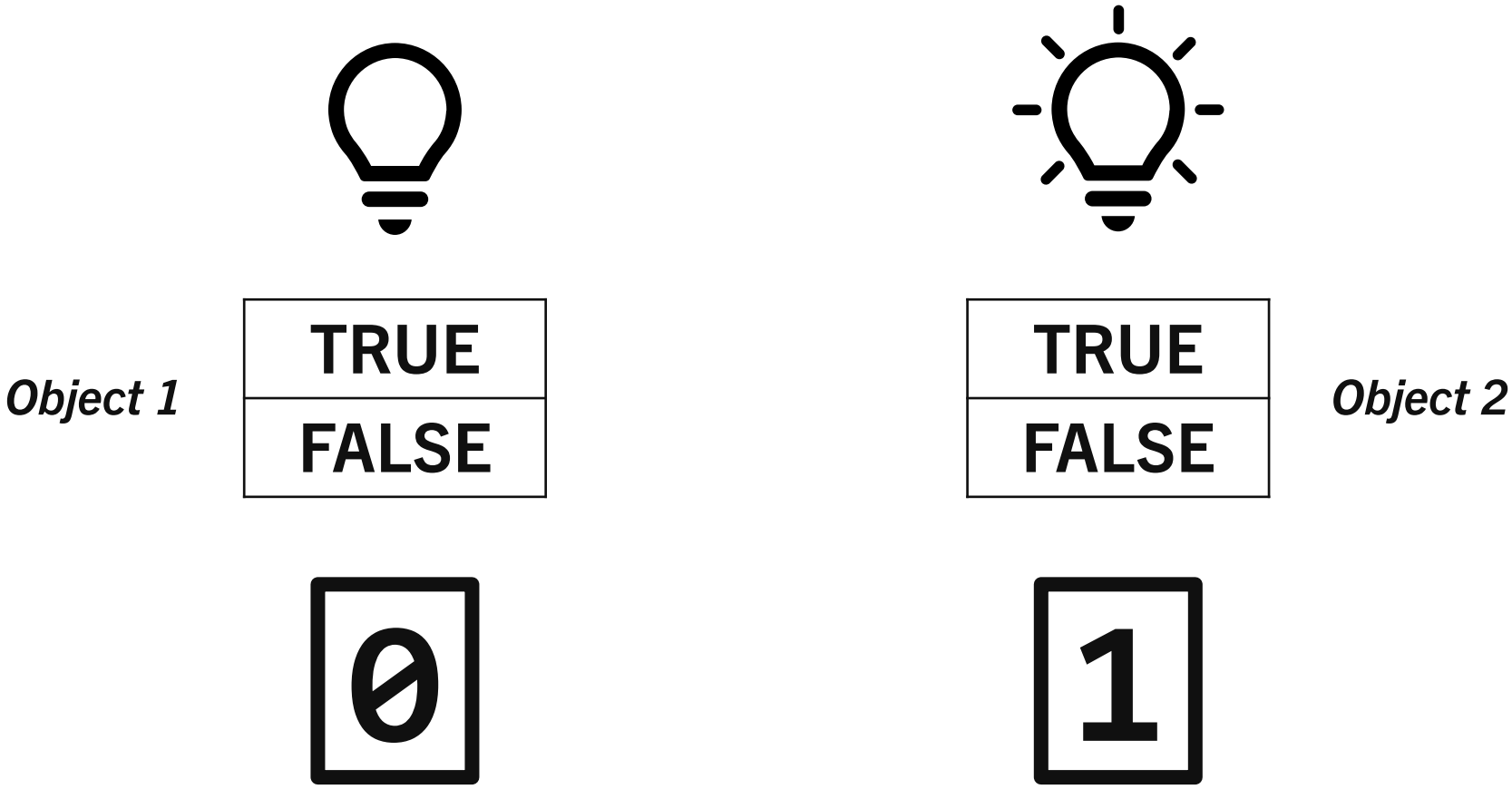
Digital Twin

How are observations provided when full synchronization is impossible?

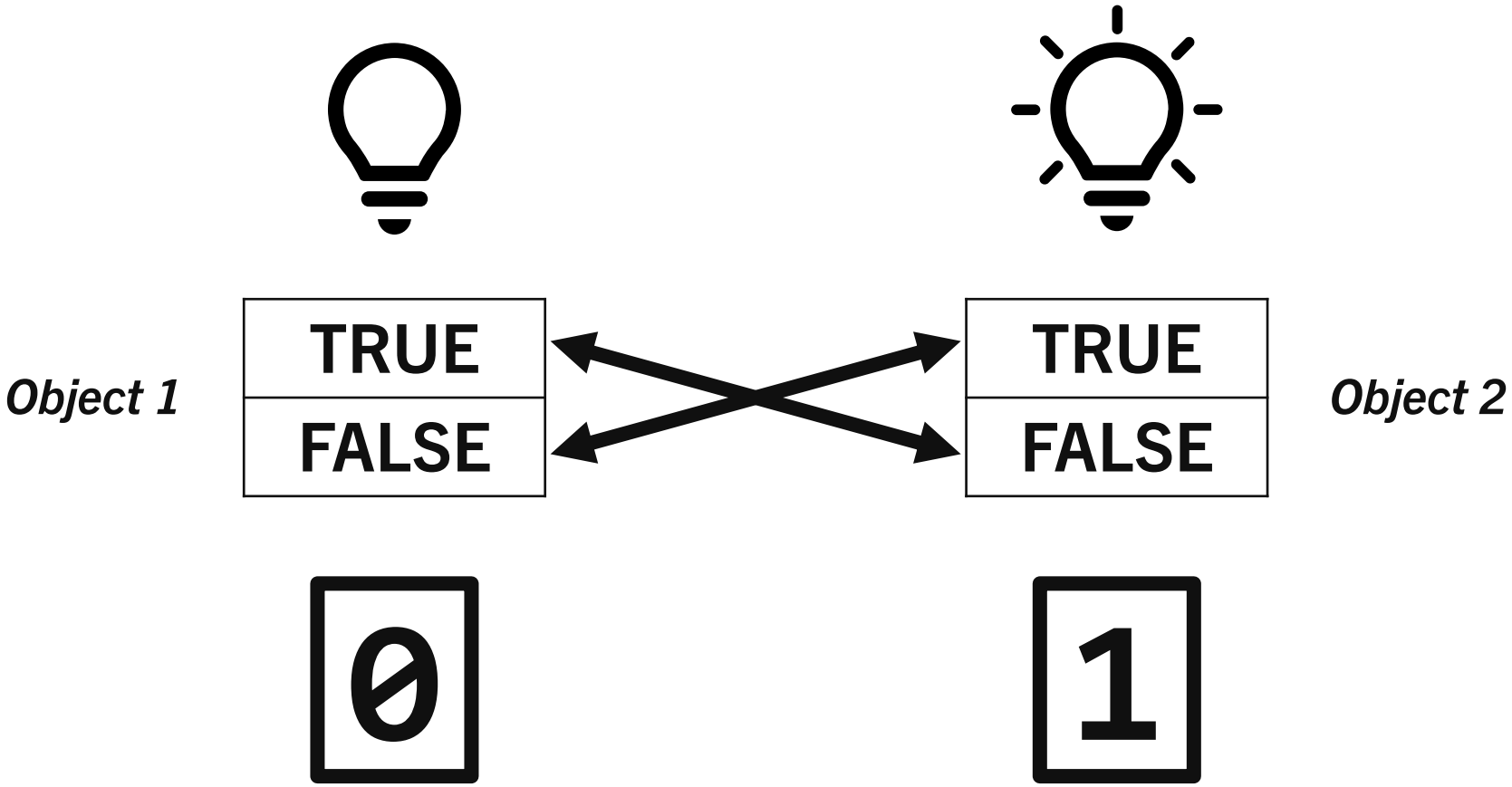
System of interest

Digital Twin

The DT shares behavior with the system of interest

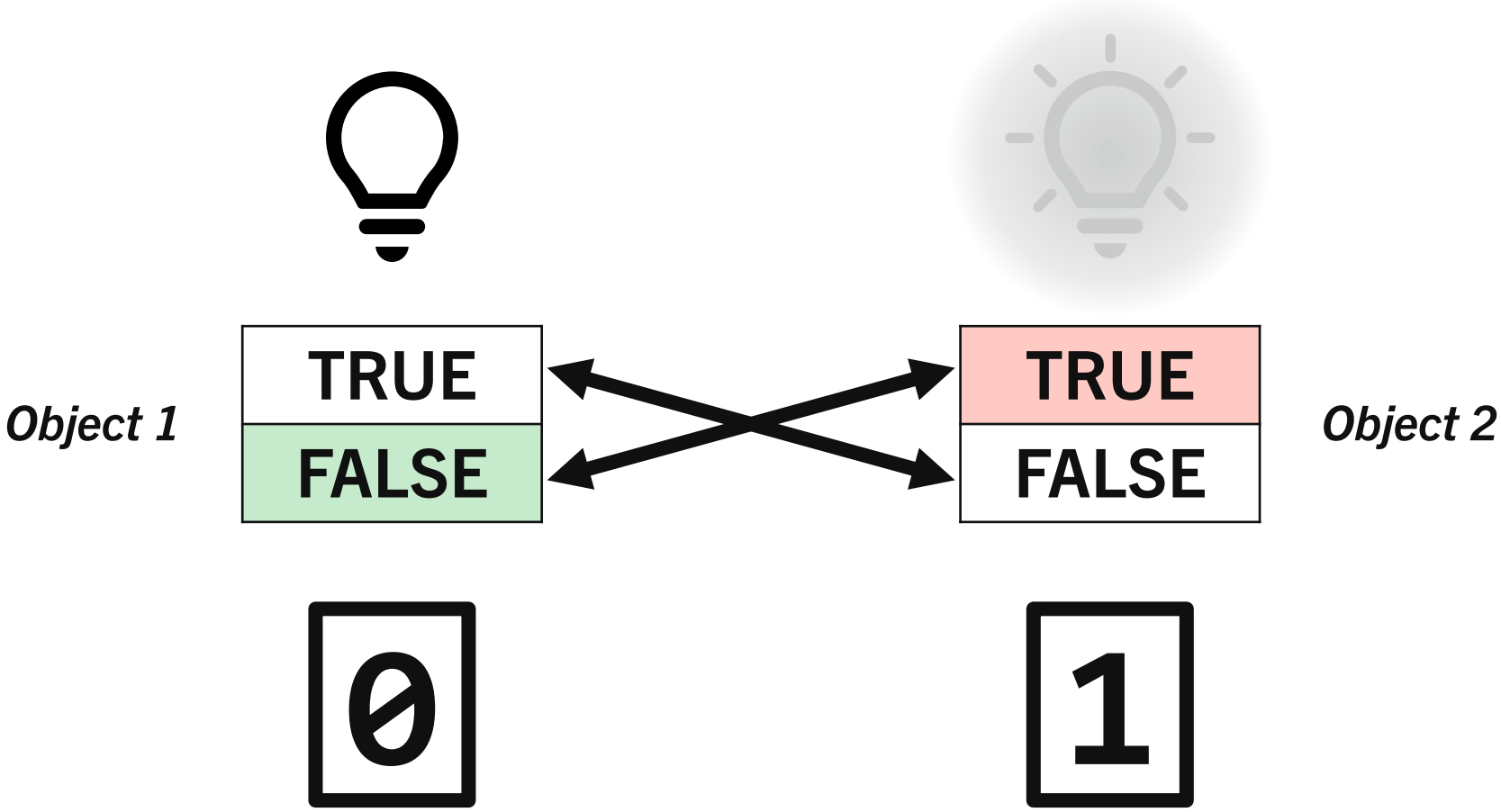


Behaviors are the constraints between system states

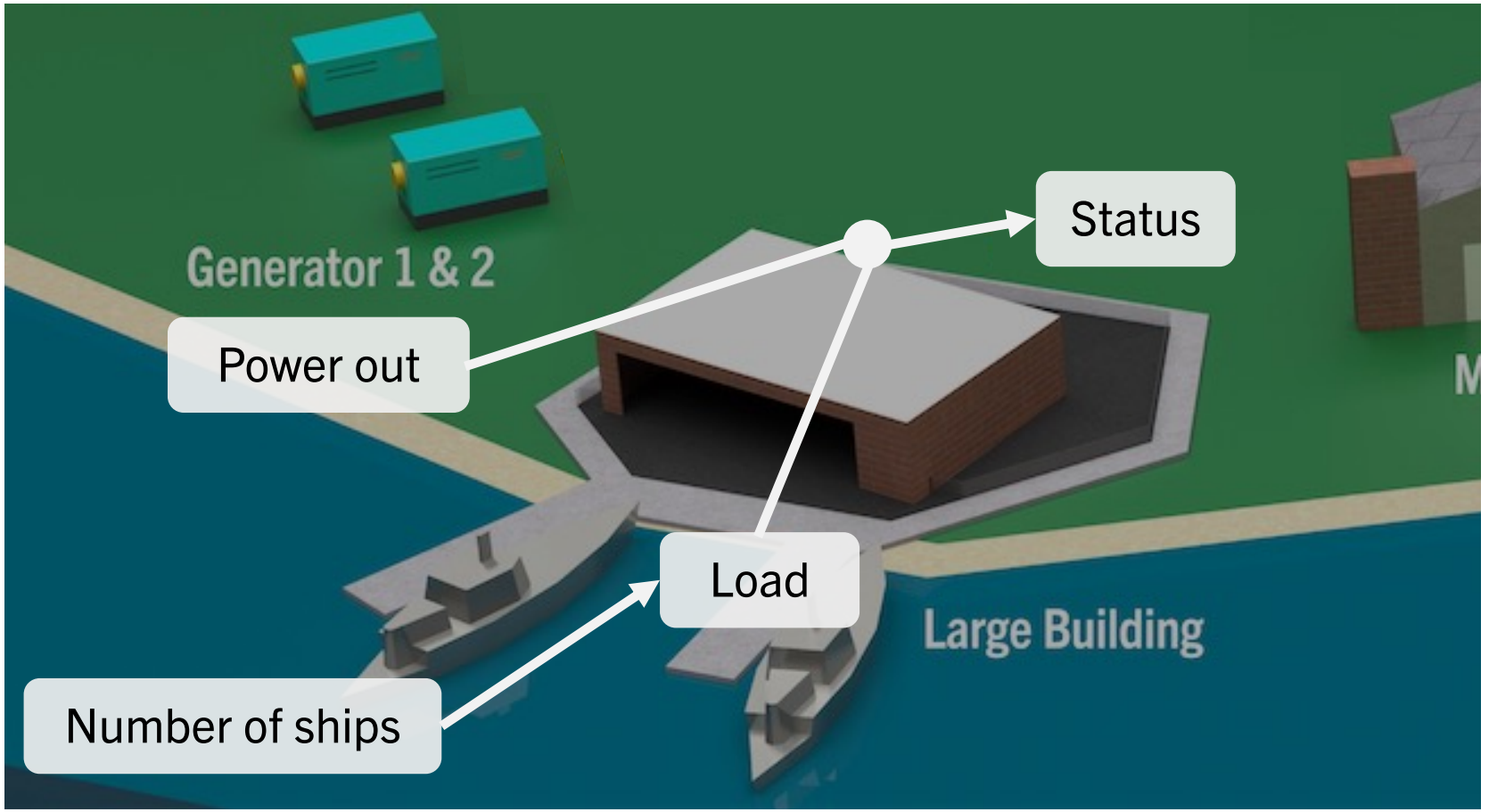


Polderman, J. W., & Willems, J. C. (1998). Dynamical Systems. In Introduction to Mathematical Systems Theory: A Behavioral Approach (Vol. 26, pp. 1–25). Springer.

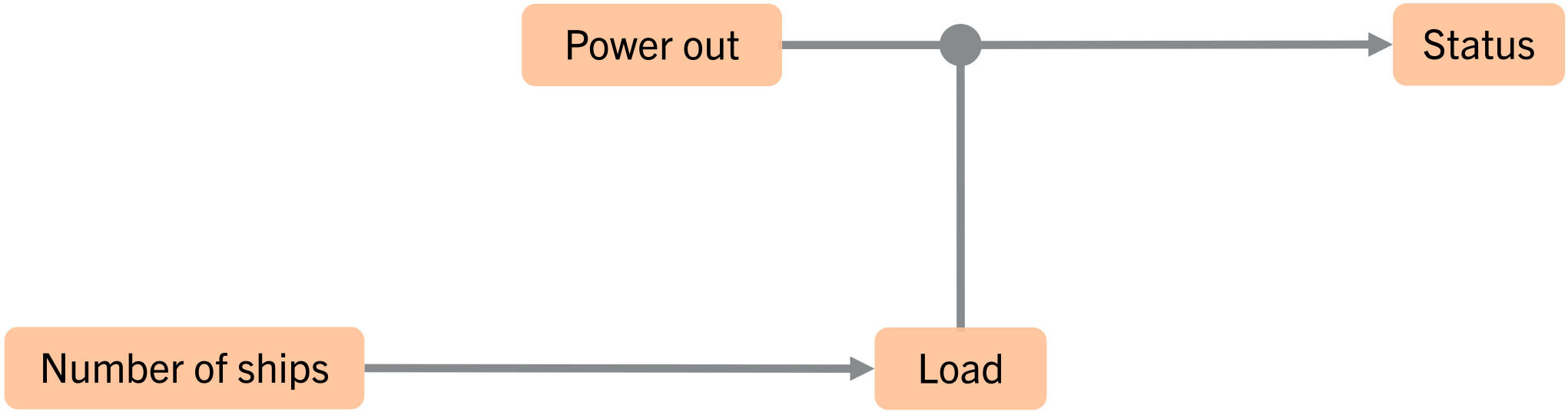
Simulations are conducted by providing inputs and outputs to models



Constraint hypergraphs reduce all system behavior to a single model

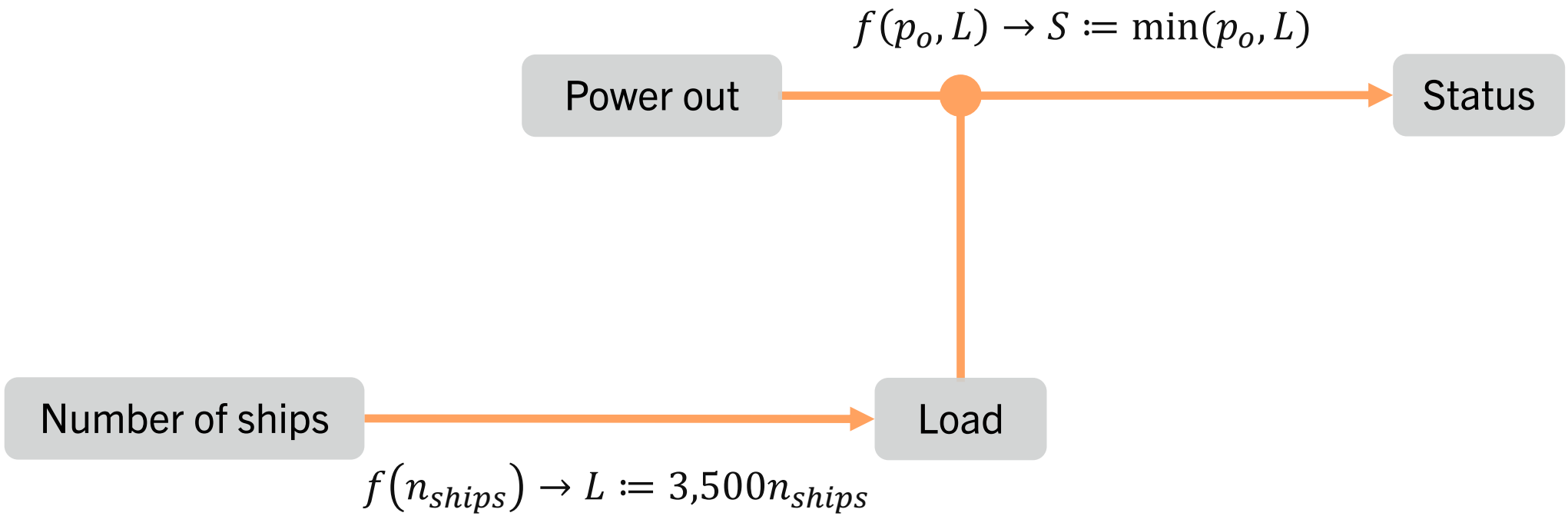


Nodes represent facts—variables—about the system



Nodes represent sets of values

Edges represents constraints on system variables

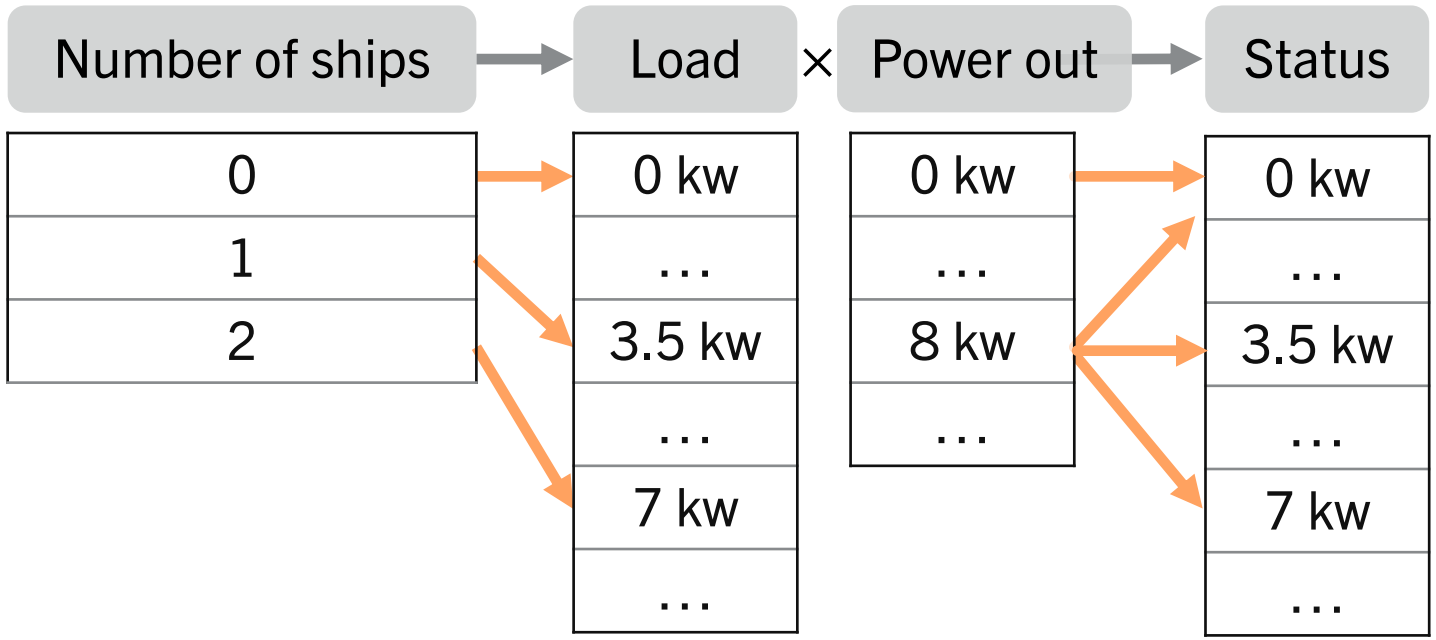


Constraints are multiple-arity, mapping combinations of multiple node values to one node

Fully constrained by definition

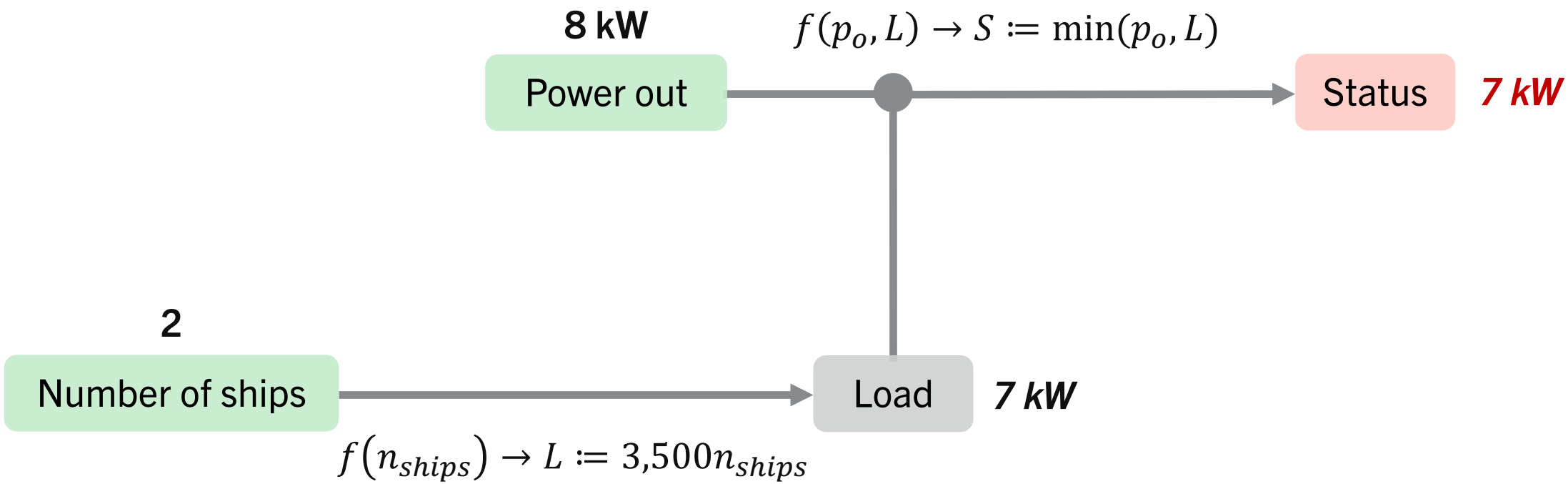
Functions map every set of nodes to a single value that is guaranteed to exist

Each path describes a valid configuration of the system



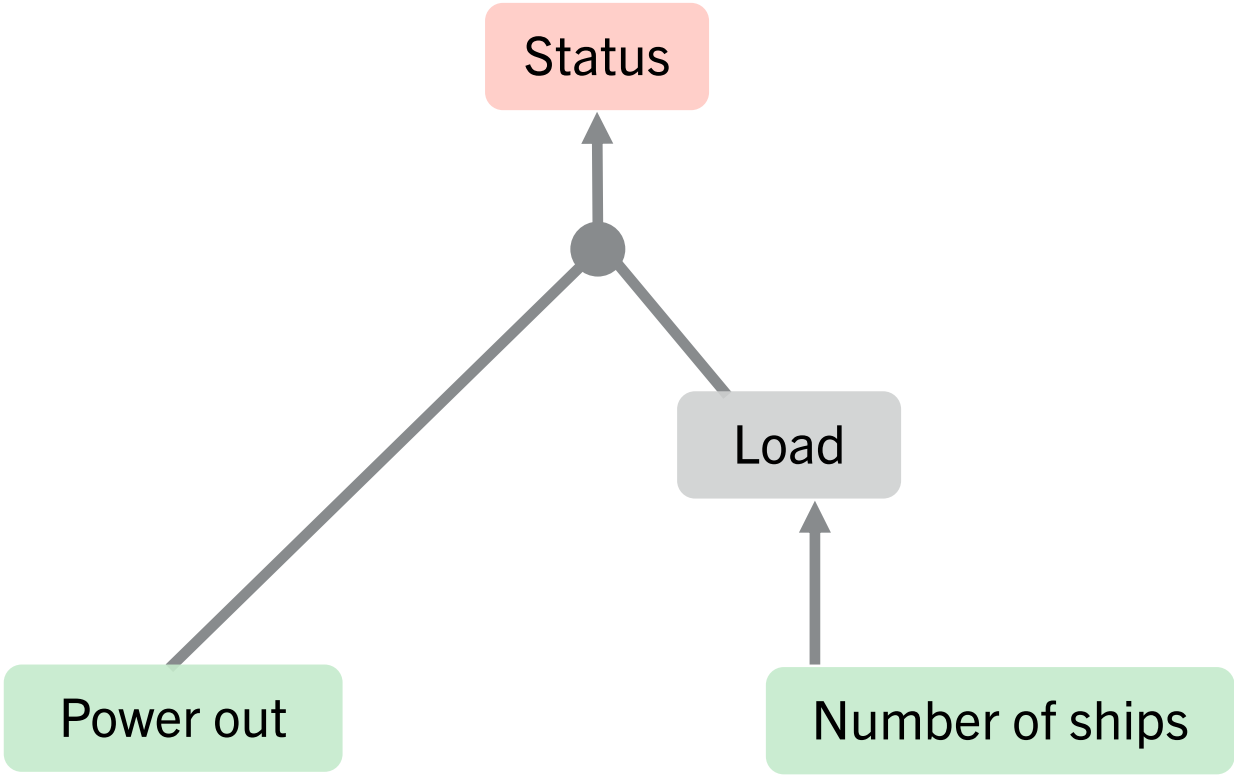
Morris, J., Mocko, G., & Wagner, J. (2025). Unified System Modeling and Simulation via Constraint Hypergraphs. Journal of Computing and Information Science in Engineering, [In Press]

Constraint Hypergraphs engender simulation through function composition



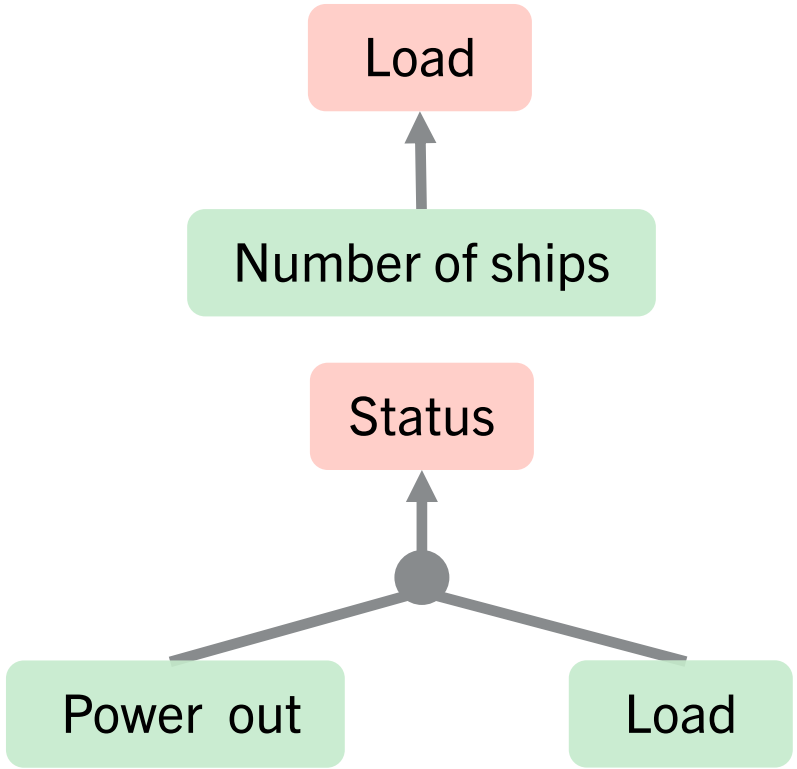
Simulation is the observation of a system's state without real-world measurements

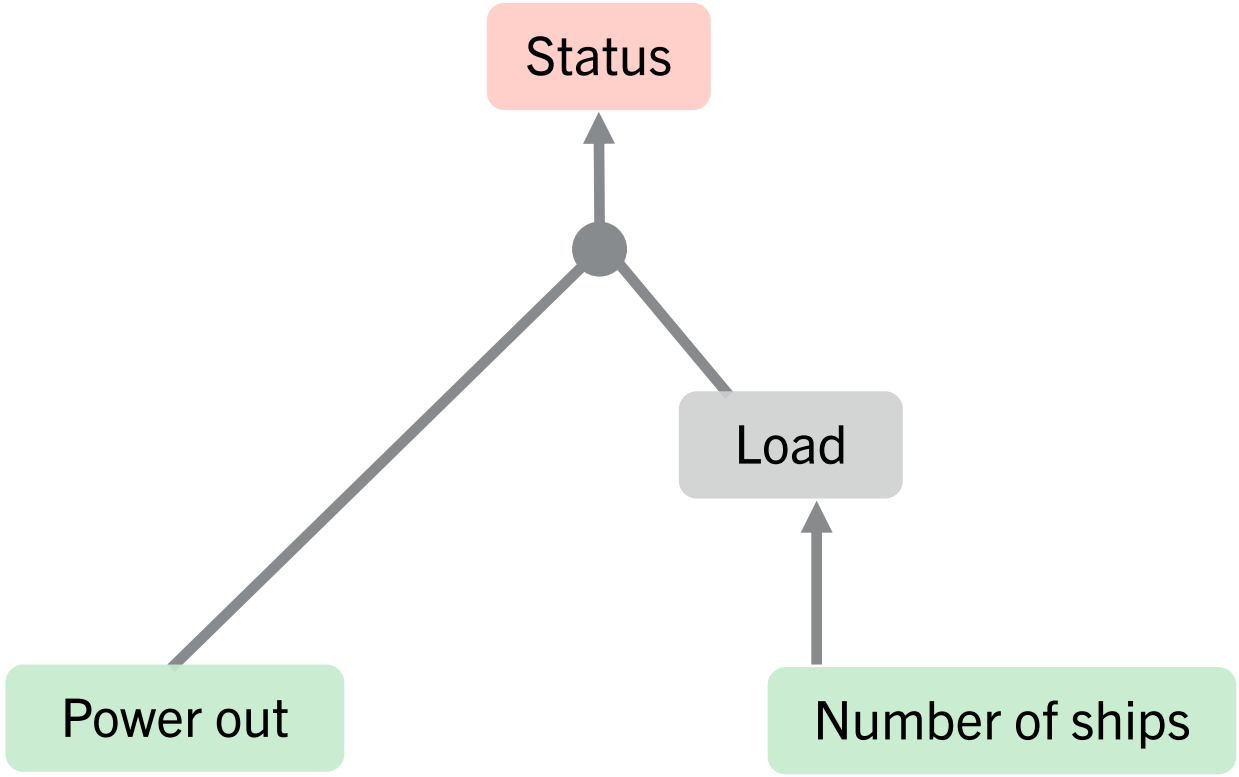
Simulation is represented as a path through the hypergraph



Paths in a hypergraph form *trees*

The set of all paths in the hypergraph show all possible simulations



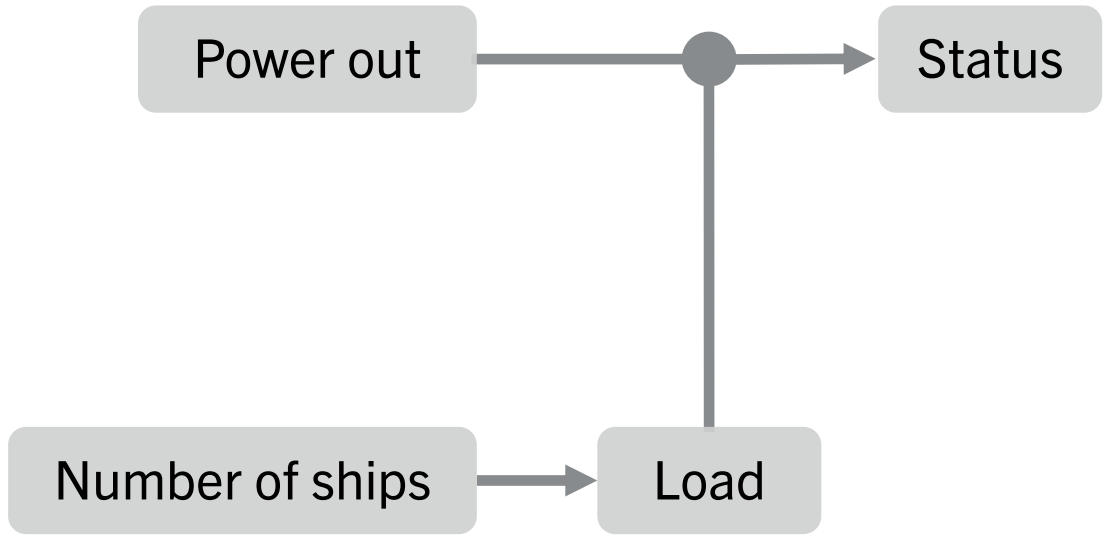


Paths in a hypergraph form *trees*

Hypergraphs expose system information to external agents

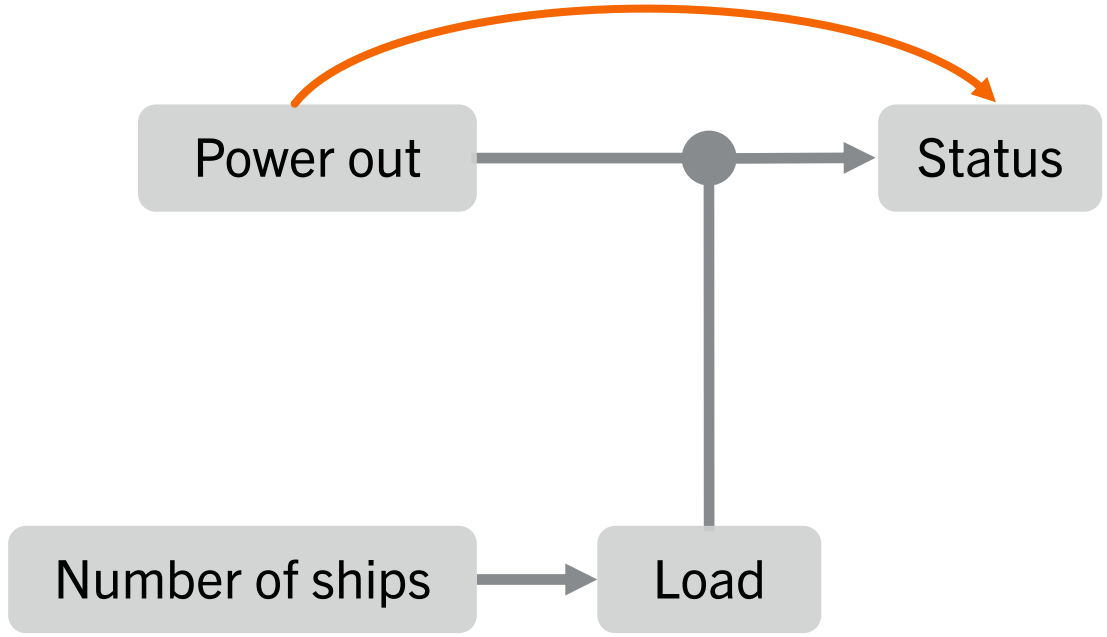
Interfacing agents—humans, computers, sensors, dashboards, etc.—can request a value for a node

If the node has an observed value, or can be simulated from a set of inputs, the hypergraph can automatically deliver the value



Independent behaviors make the hypergraphs immediately modifiable

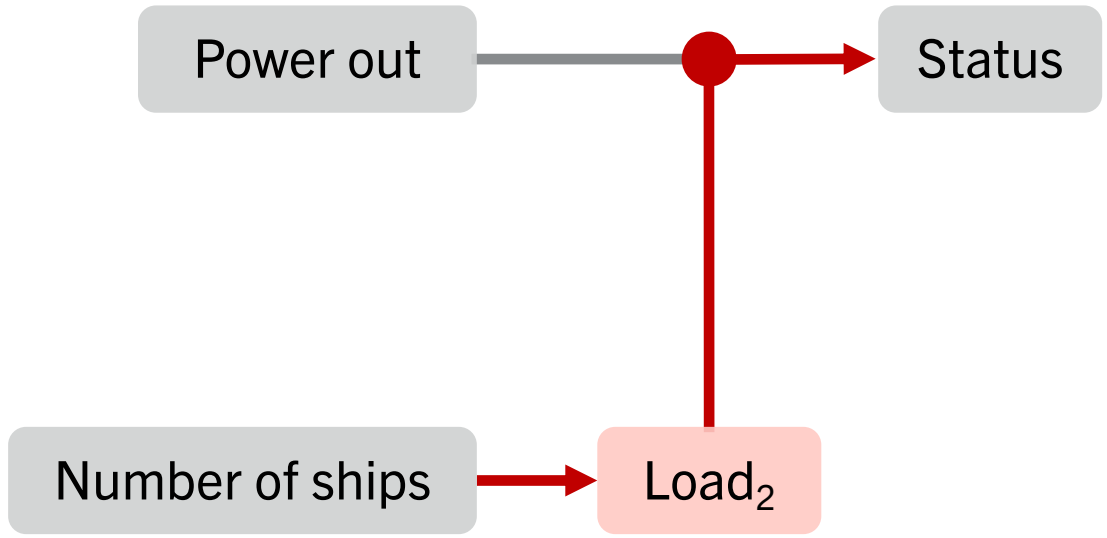
Functions can be rewritten, removed, or added without affecting the consistency of the graph *(no side effects)*



Independent behaviors make the hypergraphs immediately modifiable

Functions can be rewritten, removed, or added without affecting the consistency of the graph *(no side effects)*

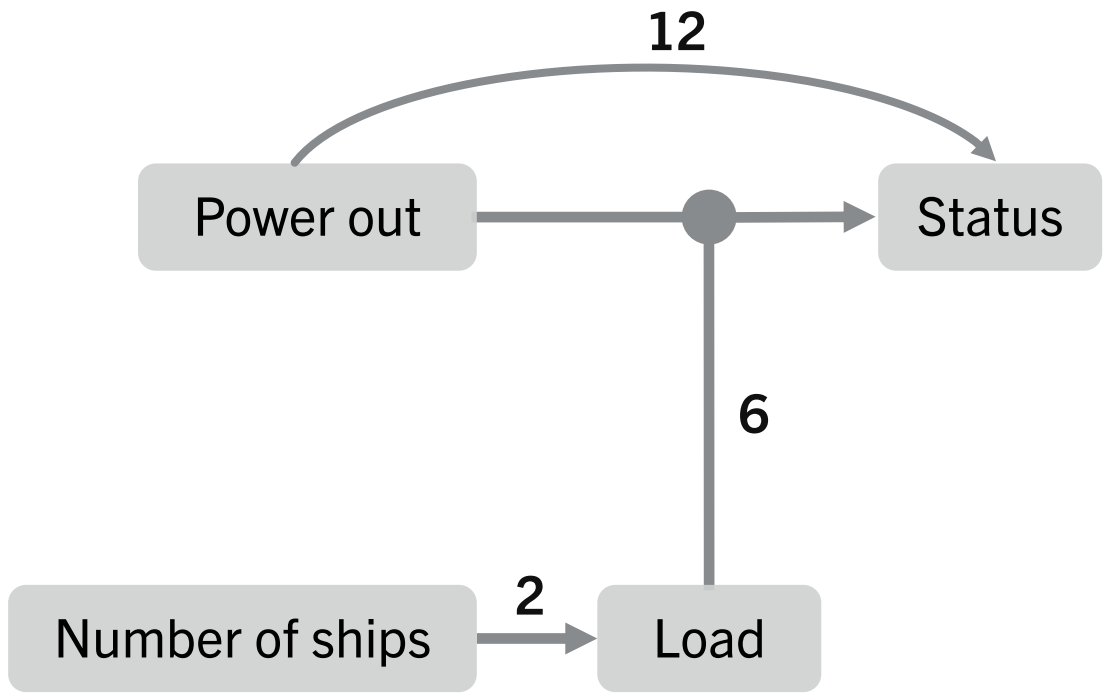
Nodes can be modified only if all connecting edges are also updated to maintain consistency *(only local side effects)*



Weights allow model characterization

Edge weights allow models to be compared (model selection)

Weights can reference a model's computation time, availability, parallelization, etc.



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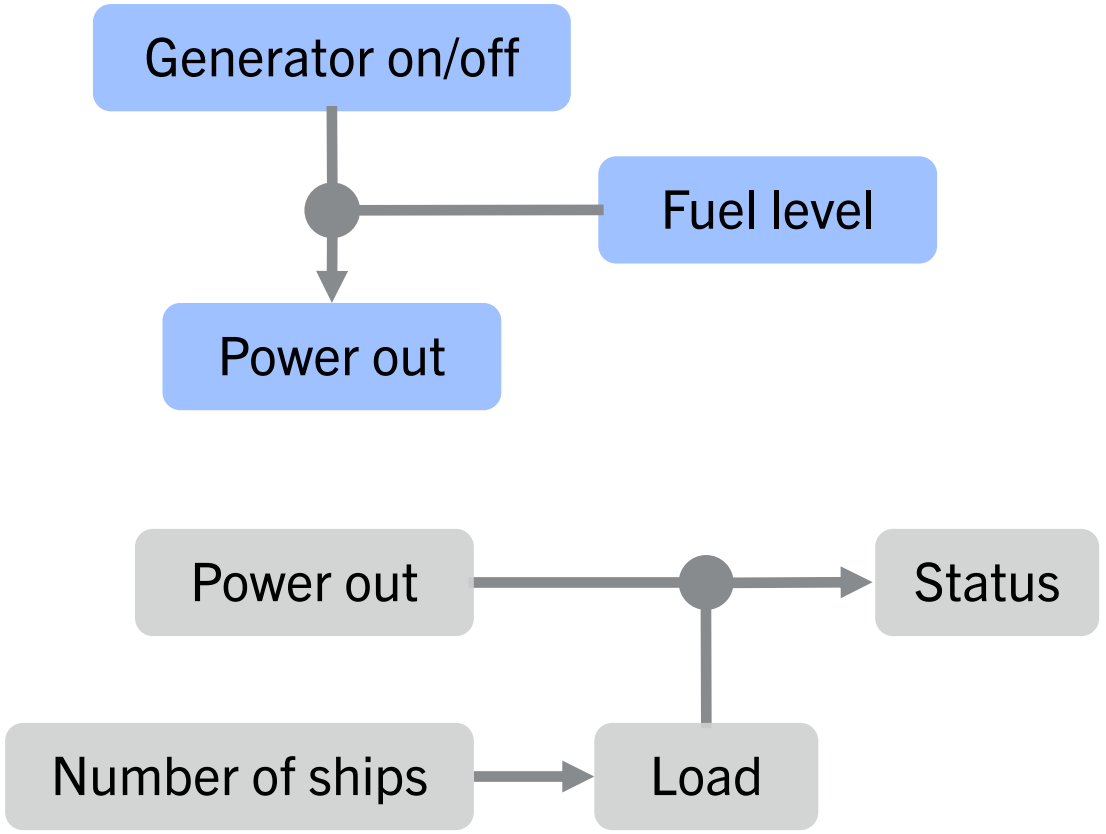
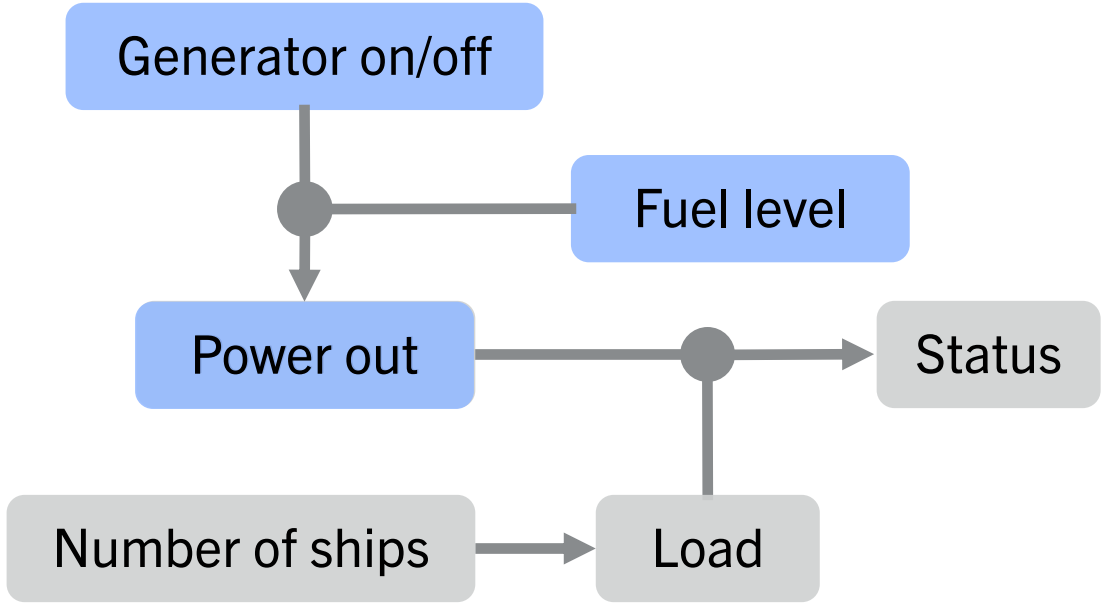
Composable with other graphs along shared nodes (union operation)

All simulation paths remain consistent

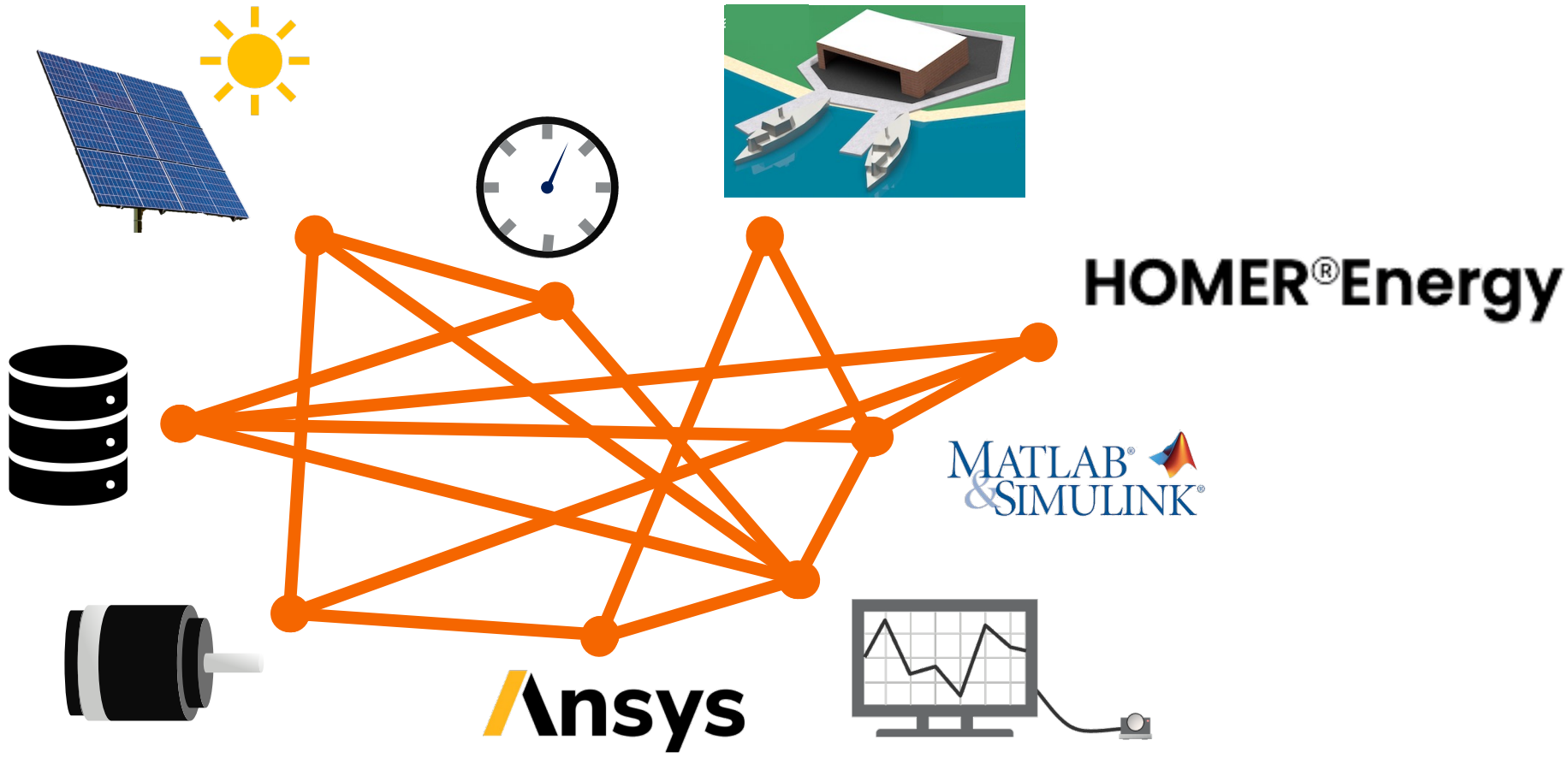


Emergent behavior is provided entirely from composition

All simulation paths remain consistent



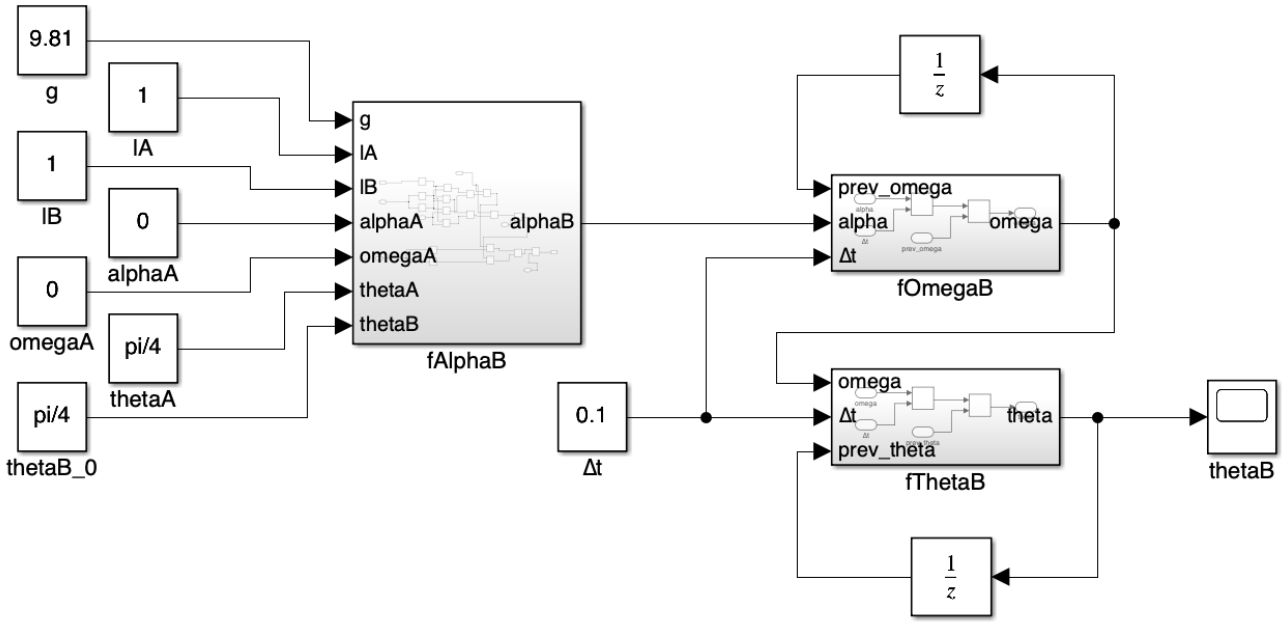
Complex systems require models to be represented and synchronized between many software applications



Imperative models provide only a single form of simulation

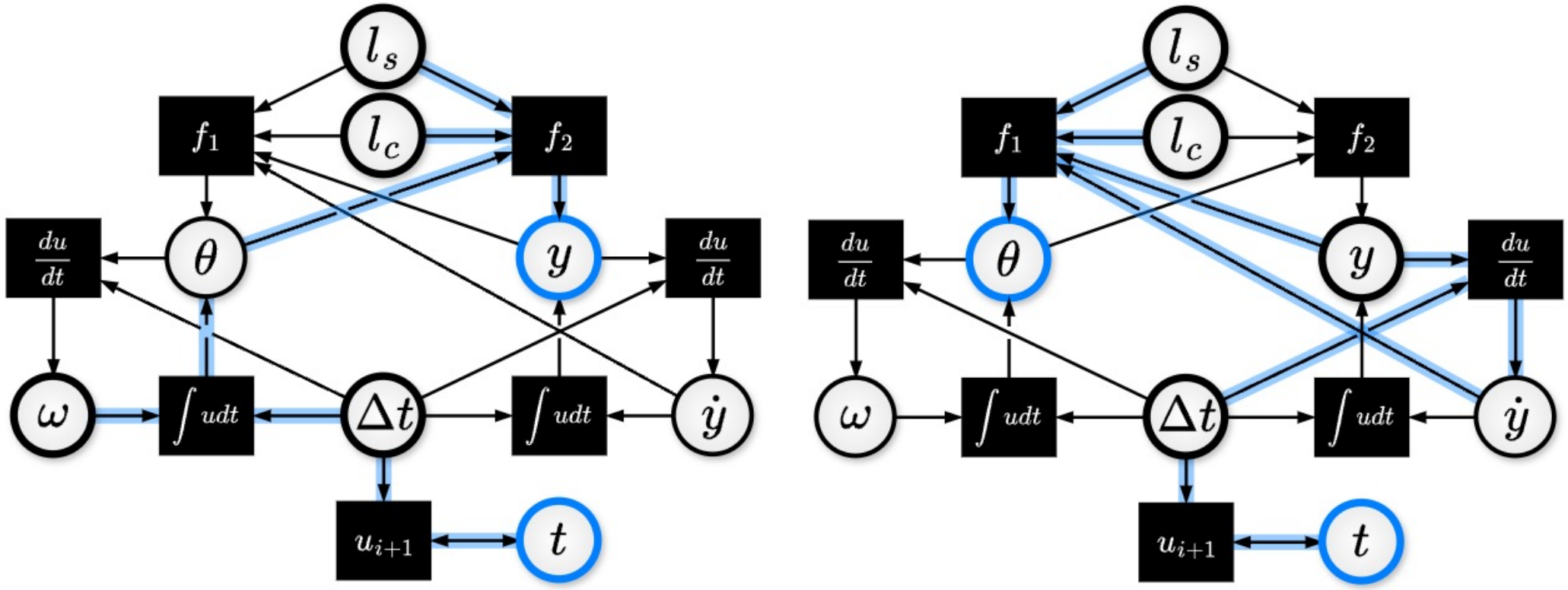
The number of imperative simulations from a system representation with n state variables is the combination of:

$$\sum_{i=1}^{n-1} n - i \binom{n}{i}$$



Morris, J., Mocko, G., Wagner, J., & Ramnath, S. (2025). Declarative Integration of CAD Software into Multi-Physics Simulation via Constraint Hypergraphs. [Submitted to the ASME IDETC-CIE 2025 conference]

Declarative models allow the modeler to focus on what the system is, rather than how it will be simulated

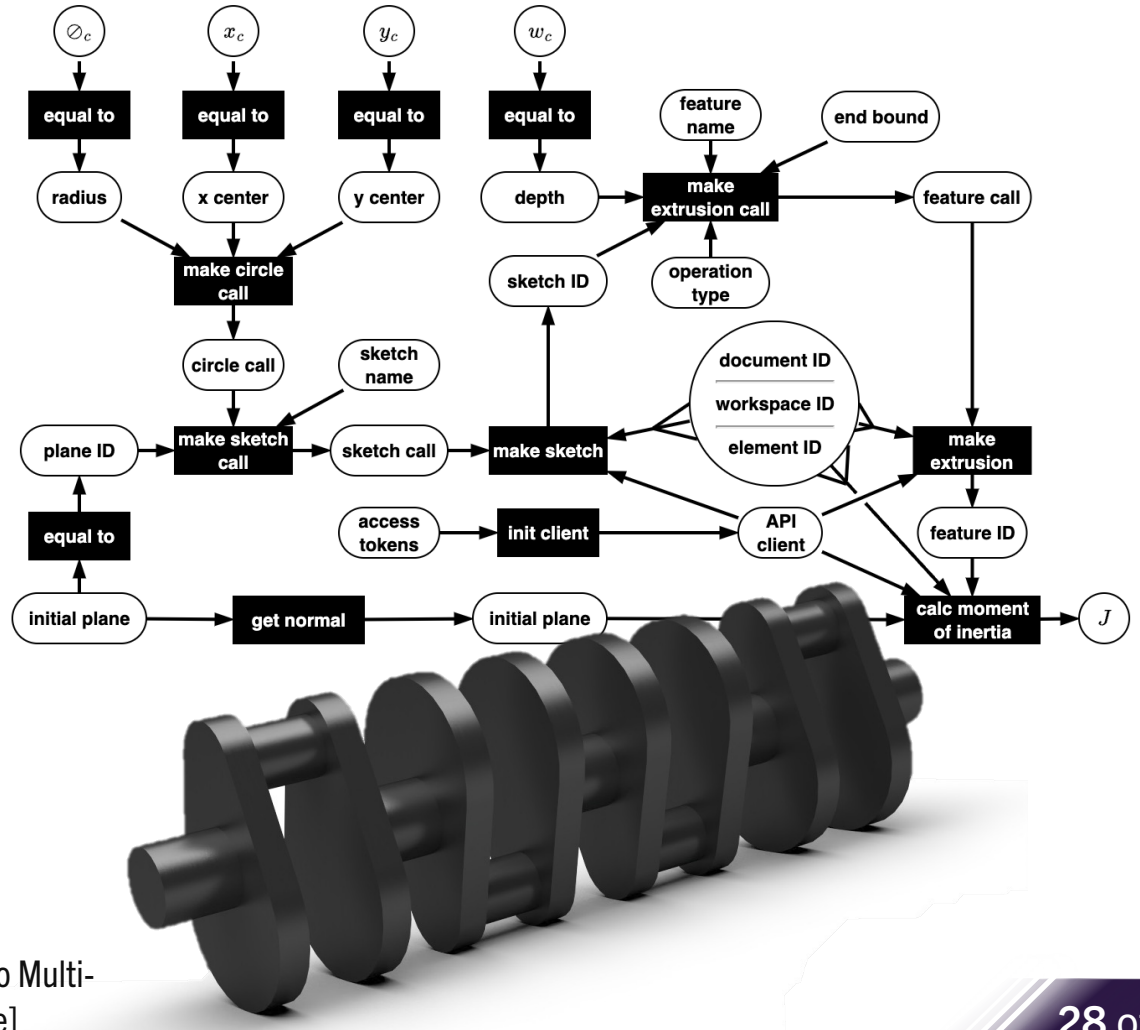


Morris, J., Mocko, G., & Wagner, J. (2025). Effects of Functional and Declarative Modeling Frameworks on System Simulation. [Manuscript Submitted for Review to the ASME Journal of Dynamic Systems, Measurement and Control].

Constraint hypergraphs tie siloed software into a single representation

Models that must be simulated in specific software (such as FE models) can be integrated into the system by wrapping API calls as edges

Avoids imperative (hard-wired) connectors between software applications



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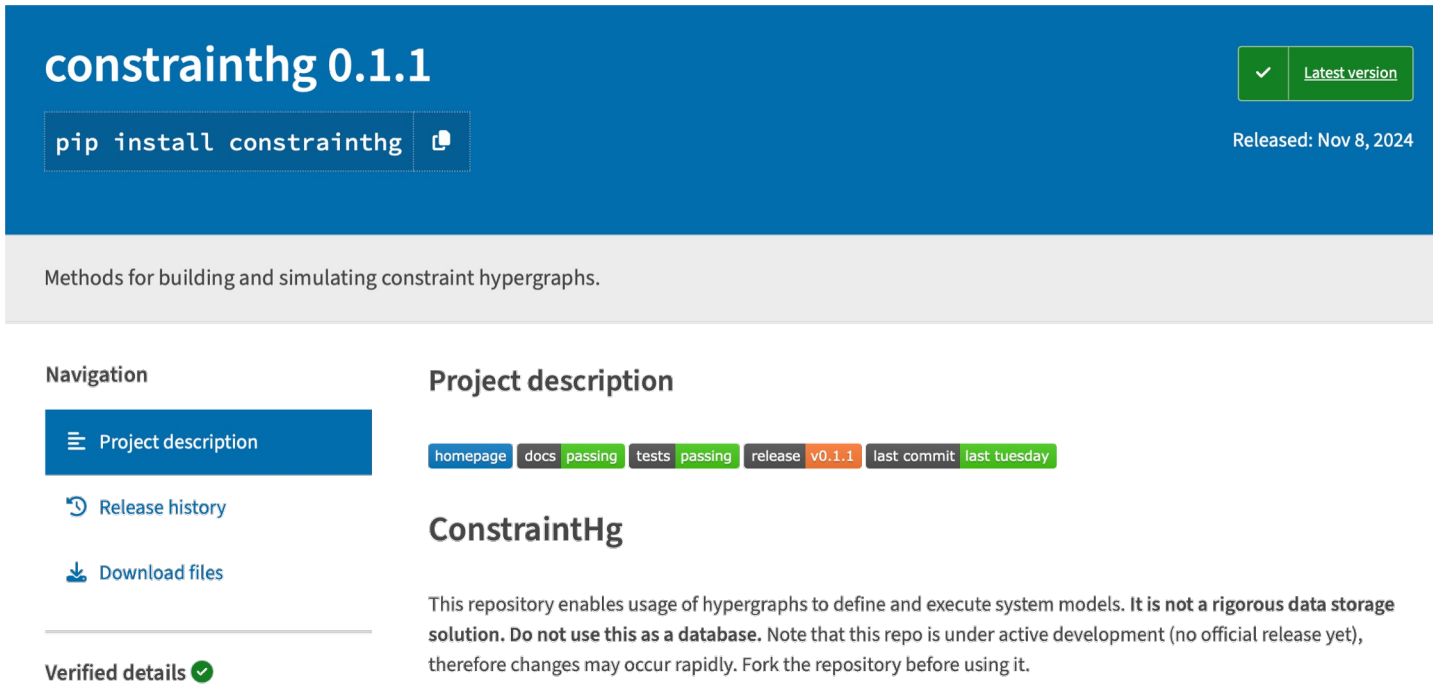
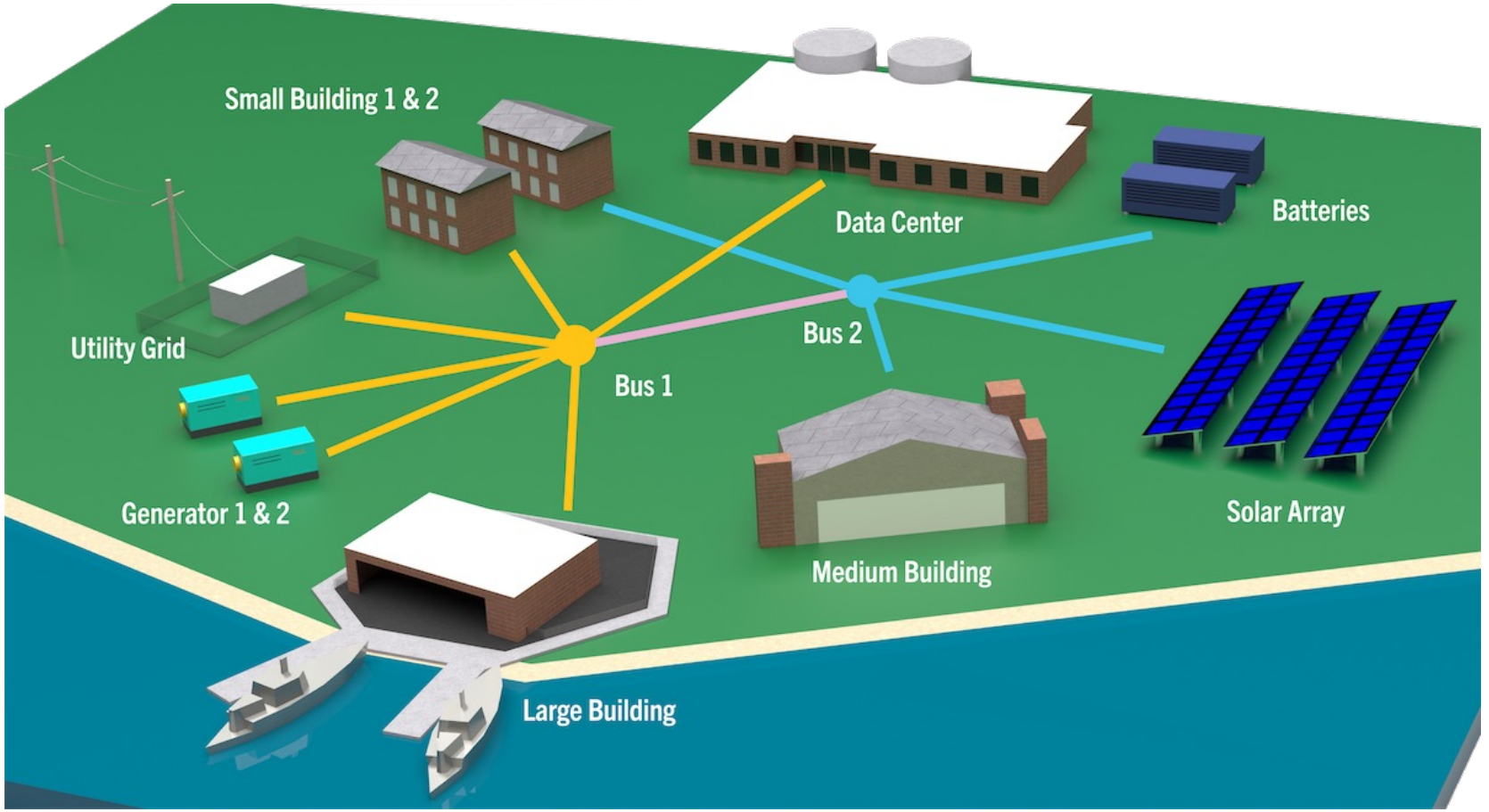
Solving engine implementing a Breadth-First Search approach available on the Python Package Index

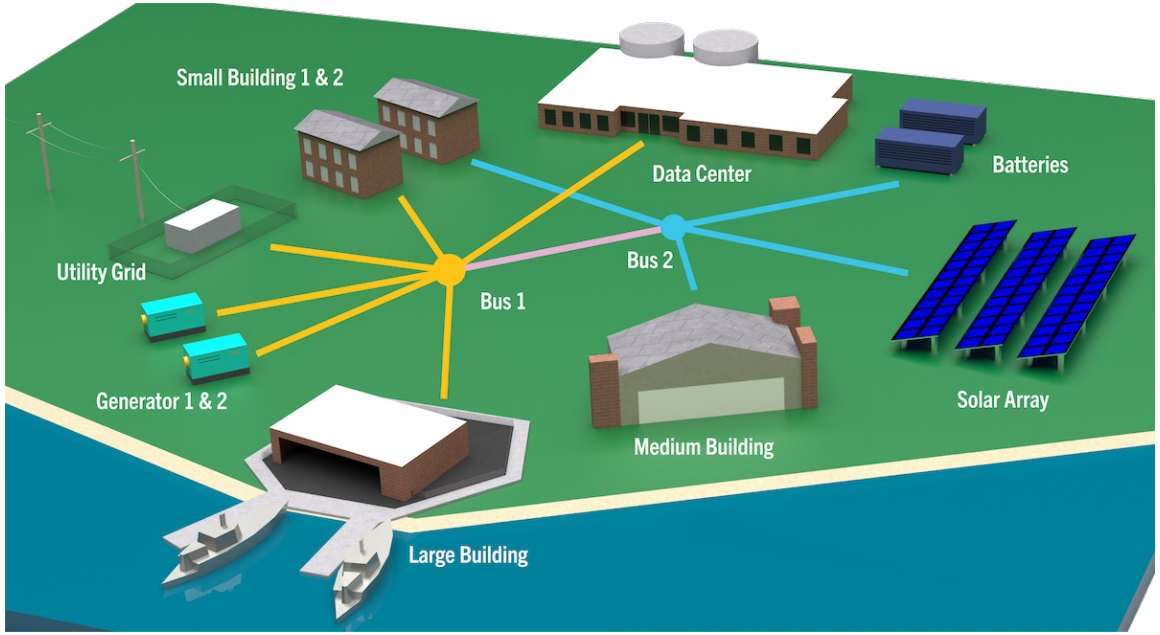


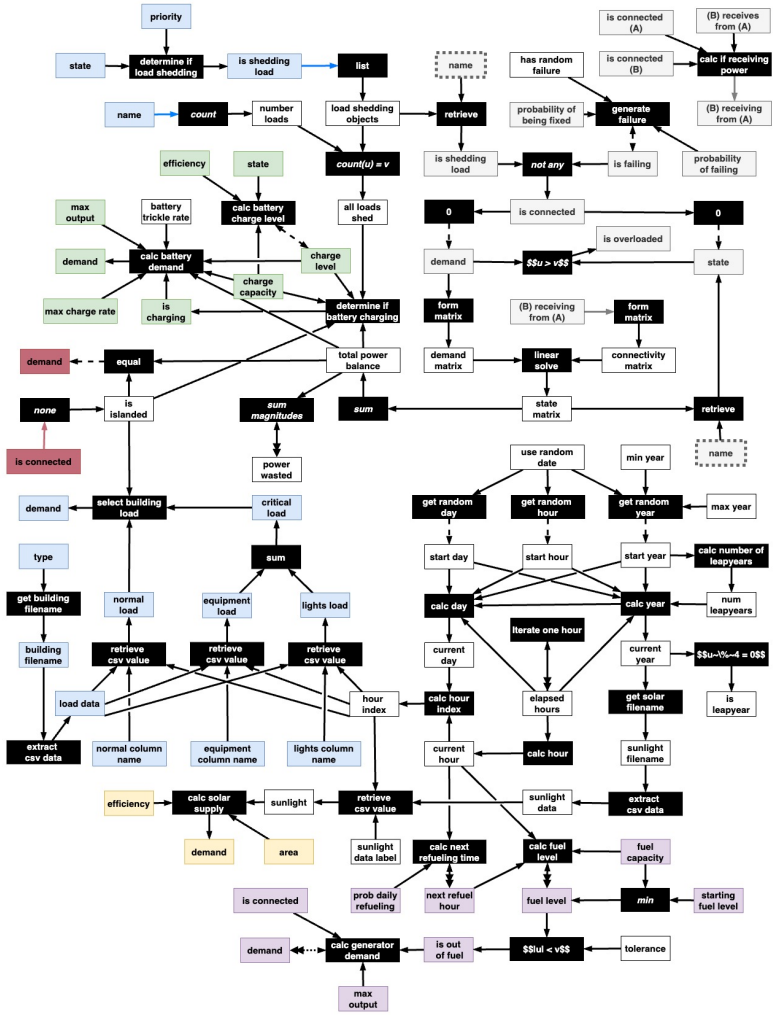
Example with a naval microgrid



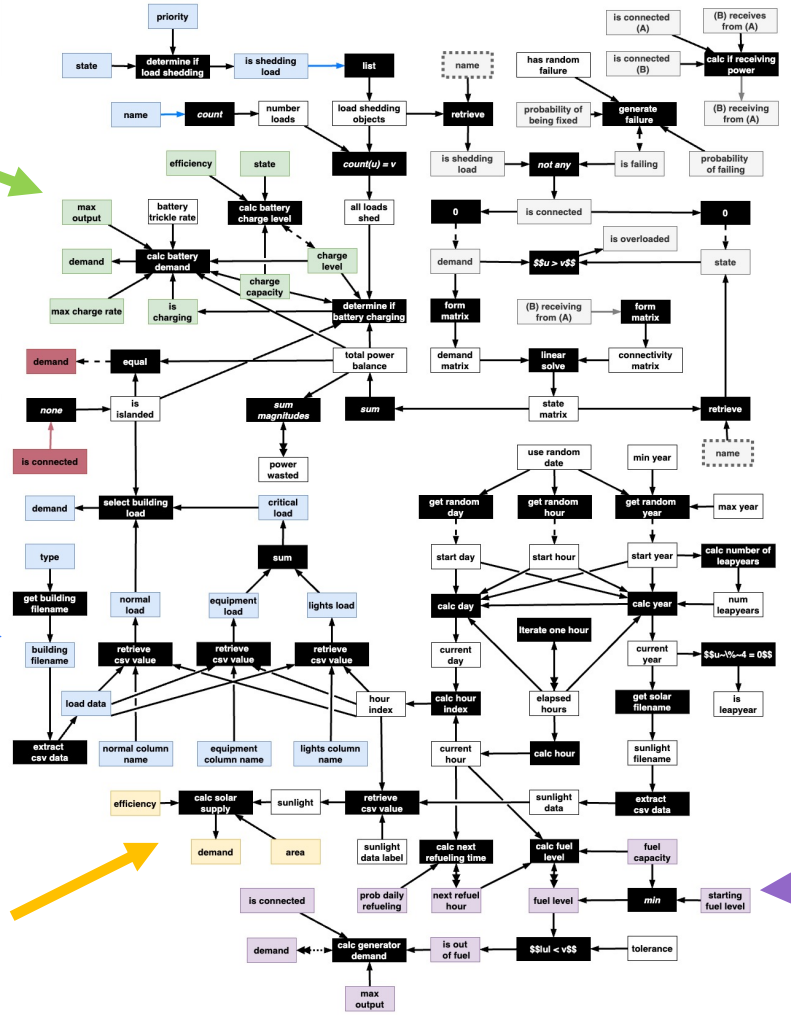
Example available on GitHub by following QR code in corner

Hypergraph contains 200 nodes and 502 edges





Regions of the hypergraph represent different *things* in the system



Batteries

Utility Grid

Loads/Buildings

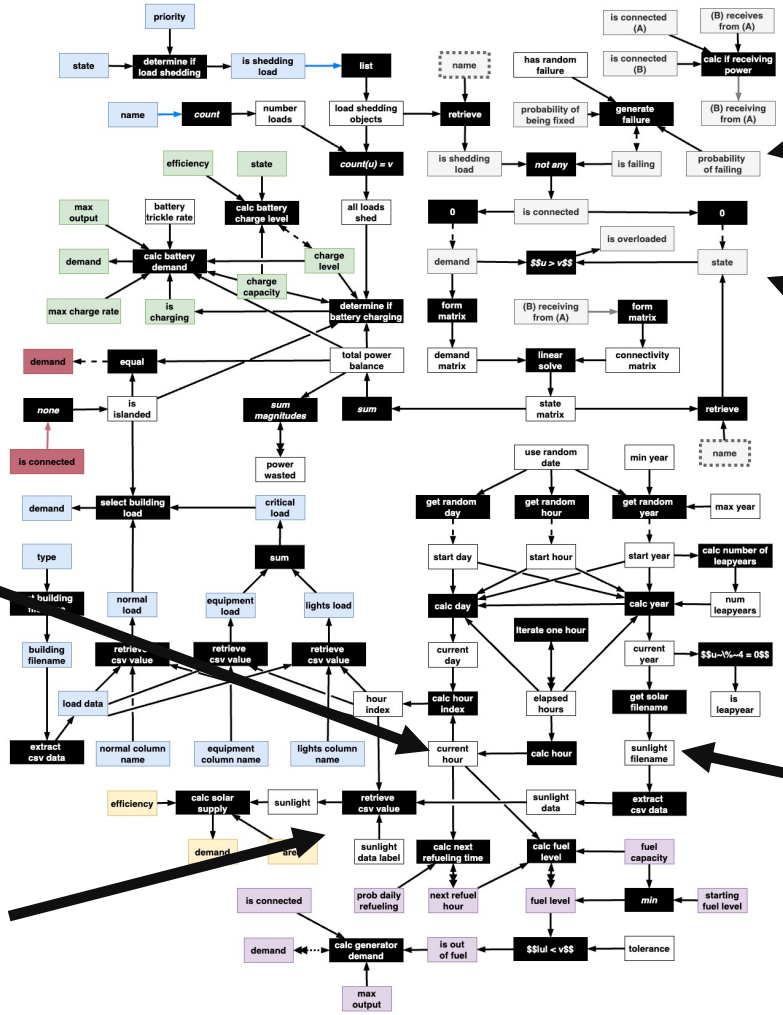
Photovoltaic Cells

Connectivity

Time/Calendar

Diesel Generators

Includes all kinds of simulation and modeling



Stochastic models

Linear equation solvers

Dynamic simulation

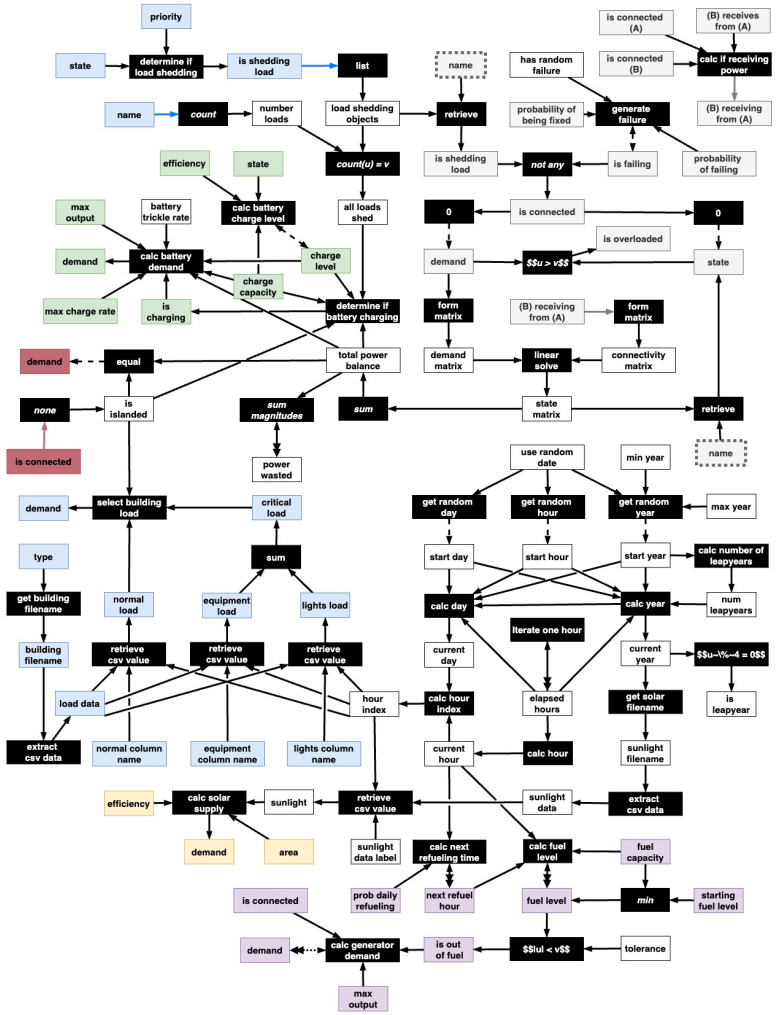
File input/output

Database queries

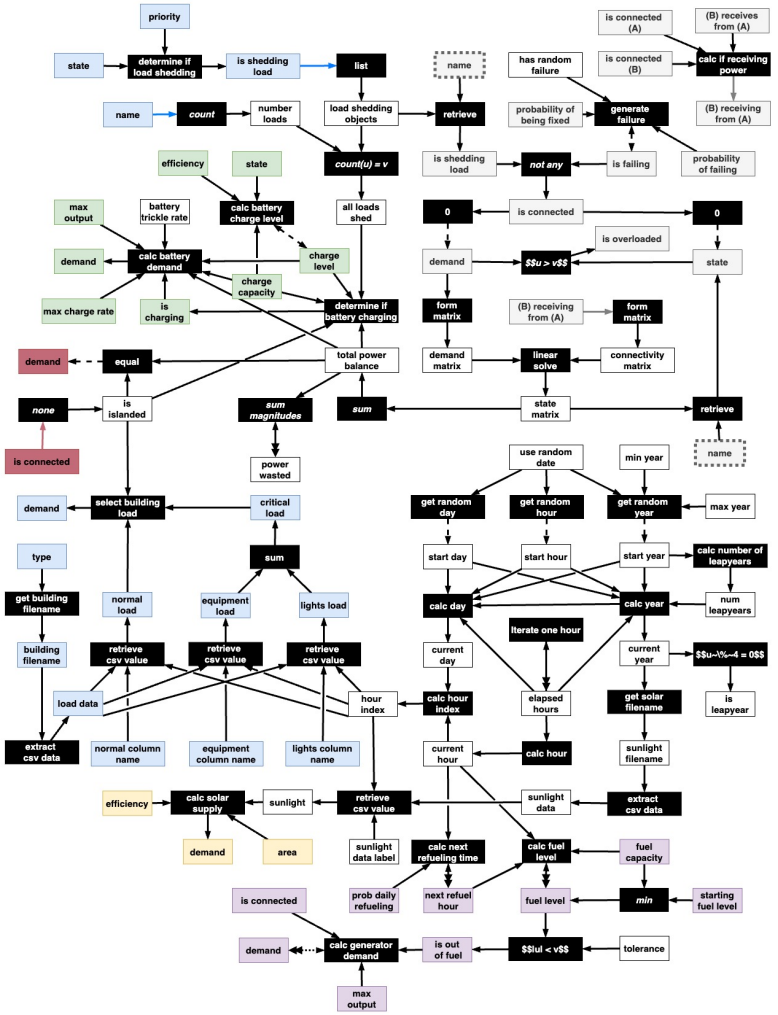
Complexity during searching (creating simulations)

Complexity of the system if fully captured by the constraint hypergraph

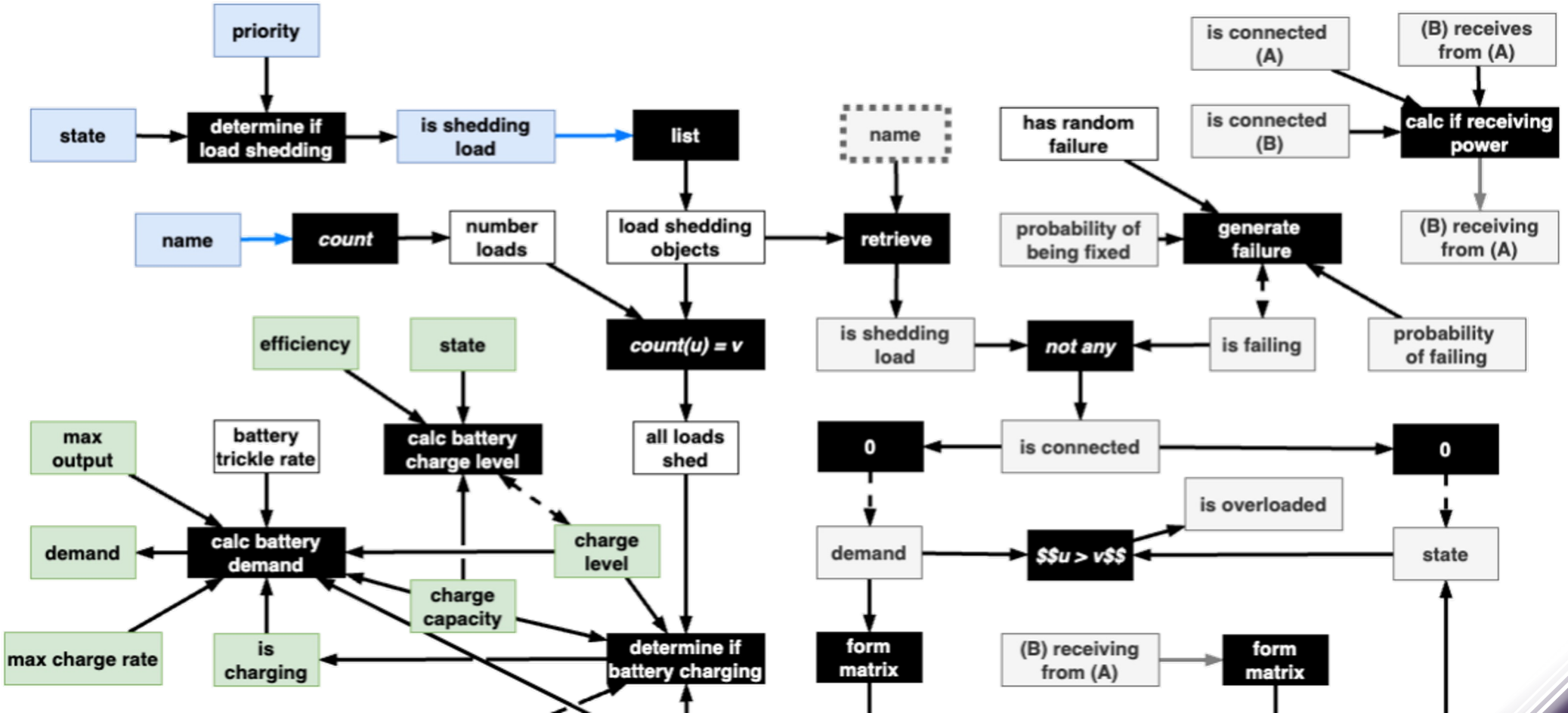
Complexity is mostly an issue when searching, as most pathfinding requires performing the simulation



Machine readable, not human readable



Hard to read for a human means hard to modify



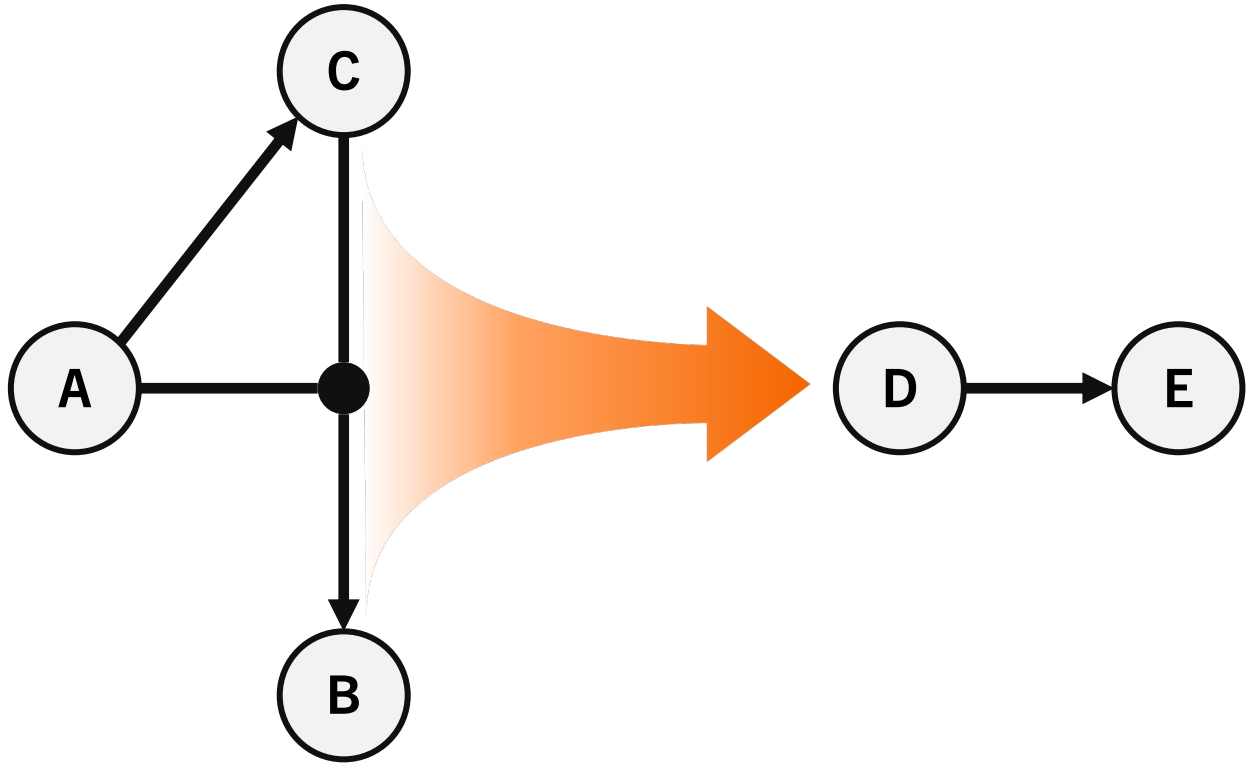
Where does this fit with existing work?

Doesn’t replace modeling

Doesn’t replace software

Doesn’t replace AI

Doesn’t replace science

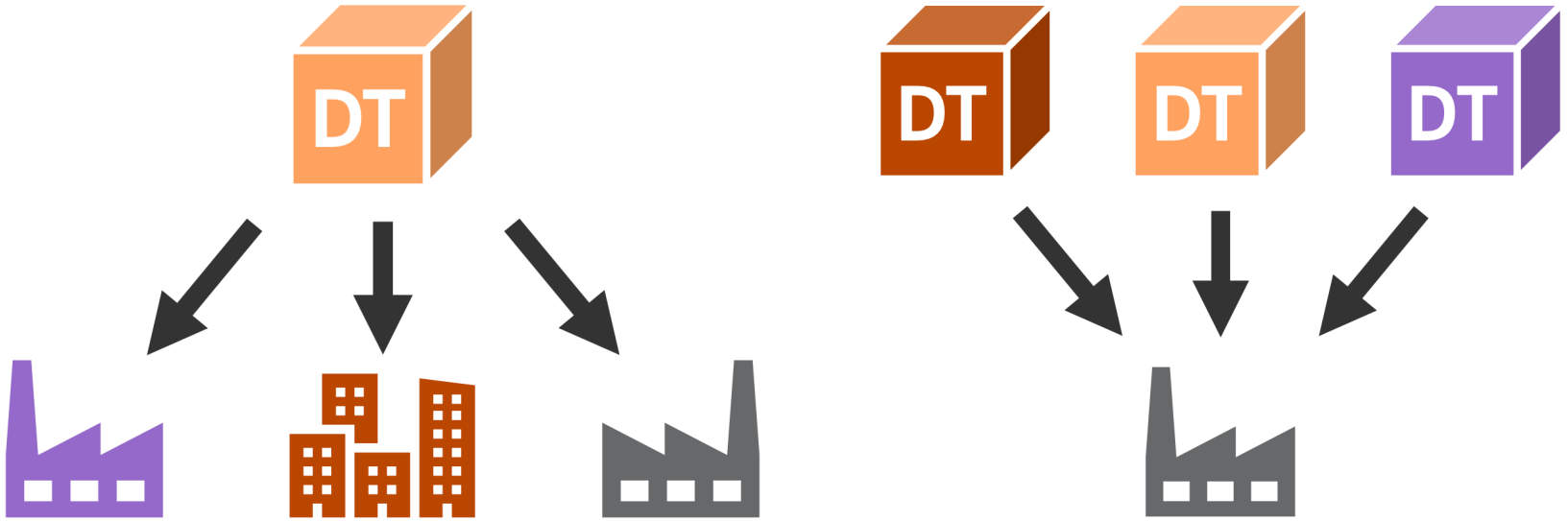


Constraint hypergraphs allow DTs to be composed, redeployed, and adapted without loss of meaning

True value is providing observations of system information—both measured and simulated

Connects all information together into a meaningful network

Wraps all possible simulations into a single model



More information available at our GitHub repository

Please reach out to jhmrrs@clemson.edu

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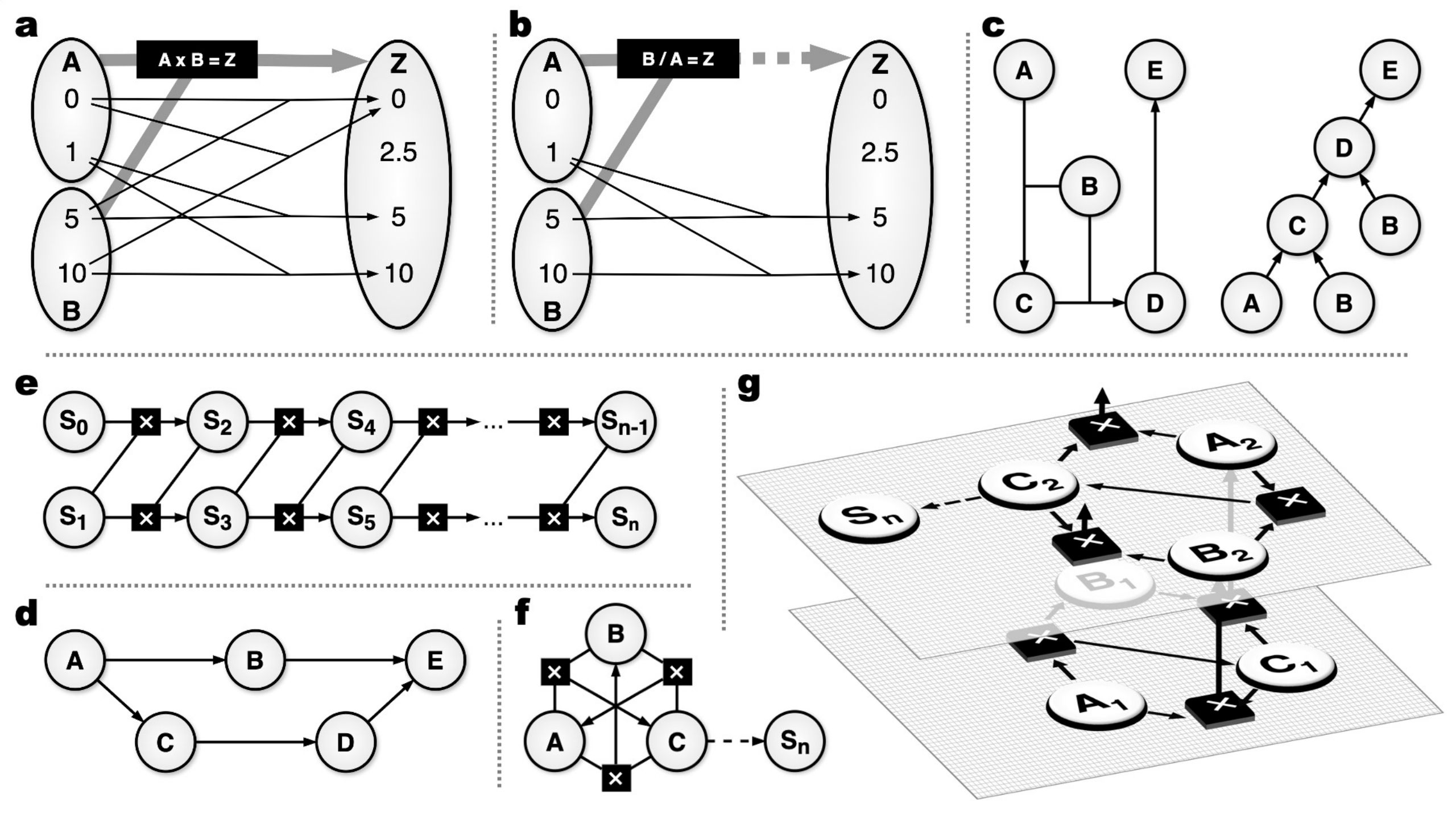
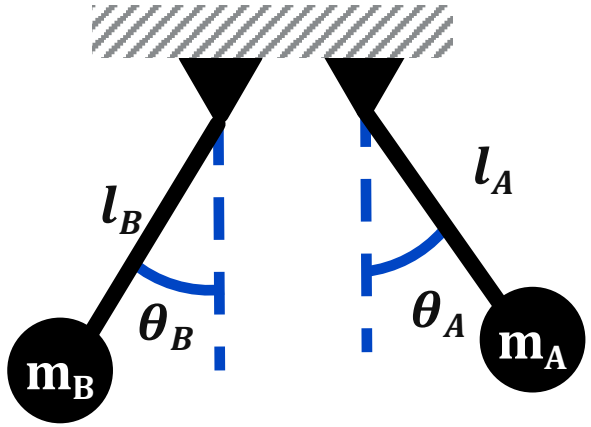

a
A
0
1
5
10
B
A x B = Z
Z
0
2.5
5
10
b
A
0
1
5
10
B
B / A = Z
Z
0
2.5
5
10
c
A
E
B
C
D
E
D
C
B
A
B
e
S0
S2
S4
Sn-1
S1
S3
S5
Sn
d
A
B
E
C
D
f
B
A
C
Sn
g
A2
C2
Sn
B2
B1
C1
A1

Emergent behavior arises out of composed functions

$$\ddot{\theta}_A = -\frac{g}{l_A}\sin\theta_A$$

$$\ddot{\theta}_B = -\frac{g}{l_B}\sin\theta_B$$

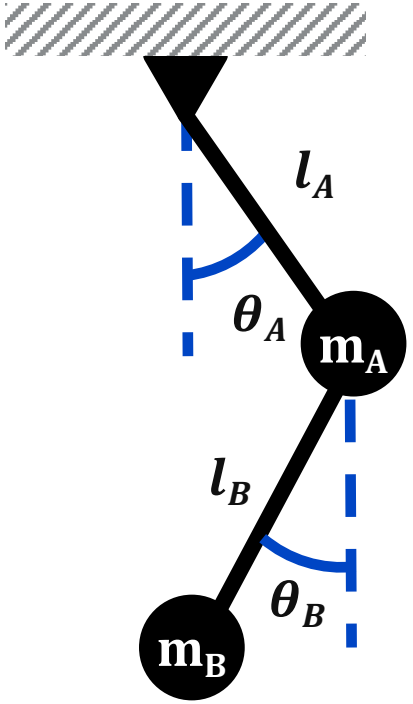


System is fully defined

When the models do not compose, the emergent behavior is not provided

$$\ddot{\theta}_A = -\frac{g}{l_A}\sin\theta_A$$

$$\ddot{\theta}_B = -\frac{g}{l_B}\sin\theta_B$$

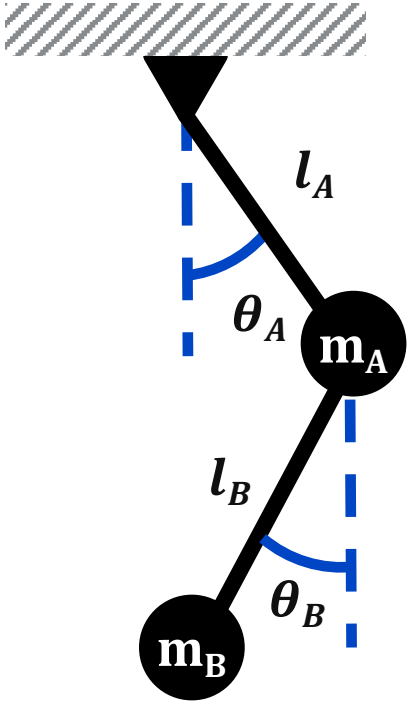


System cannot be represented as desired with the given nodes

Composition must be provided by showing the relationships between nodes

$$\ddot{\theta}_A = -\frac{g}{l_A}\sin\theta_A$$

$$\ddot{\theta}_B = -\frac{1}{l_B}(\ddot{x}_B\cos\theta_B + \sin\theta_B(\ddot{y}_B + g))$$



New model shows how the bobs can be joined

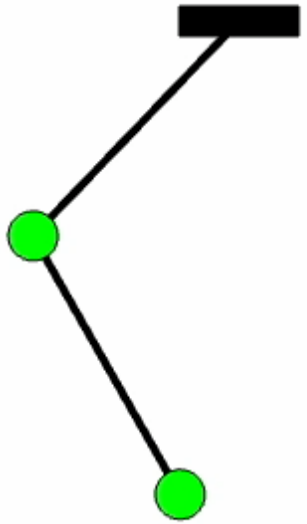
When composition is given, the emergent behavior can be fully realized

$$\ddot{\theta}_B = -\frac{1}{l_B}(\ddot{x}_B \cos\theta_B + \sin\theta_B(\ddot{y}_B + g))$$

where:

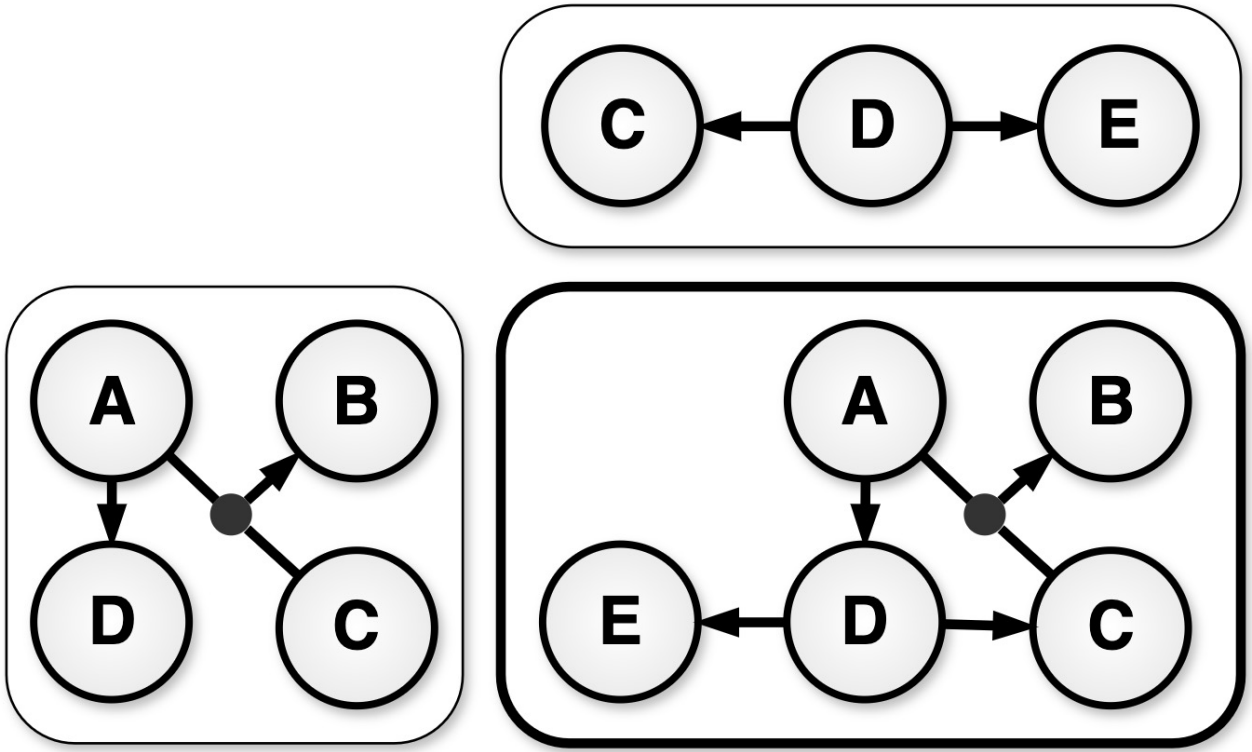
$$\ddot{x}_B = l_A(\alpha_A \cos\theta_A - \omega_A^2 \sin\theta_A)$$

$$\ddot{y}_B = l_A(\alpha_A \sin\theta_A + \omega_A^2 \cos\theta_A)$$



Solving for $\ddot{x}_B$ and $\ddot{y}_B$ using terms from the graph for A gives you the emergent behavior

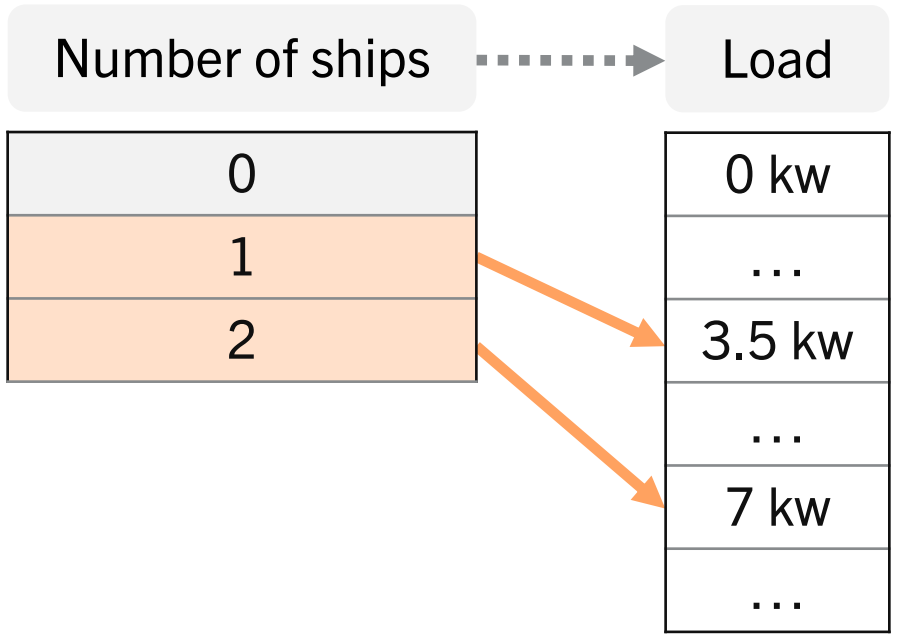
Emergent behaviors only emerge if modeled



Model only shows what you know—unseen behaviors can still exist

Properties of Constraint Hypergraphs: Handle model validity frames

Models with given validity frames can be described by functions that map from a *subset* of a node's values



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