

# Architecting Digital Twins for Model-Centric Engineering: Semantic and Machine Learning Approach

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**By**

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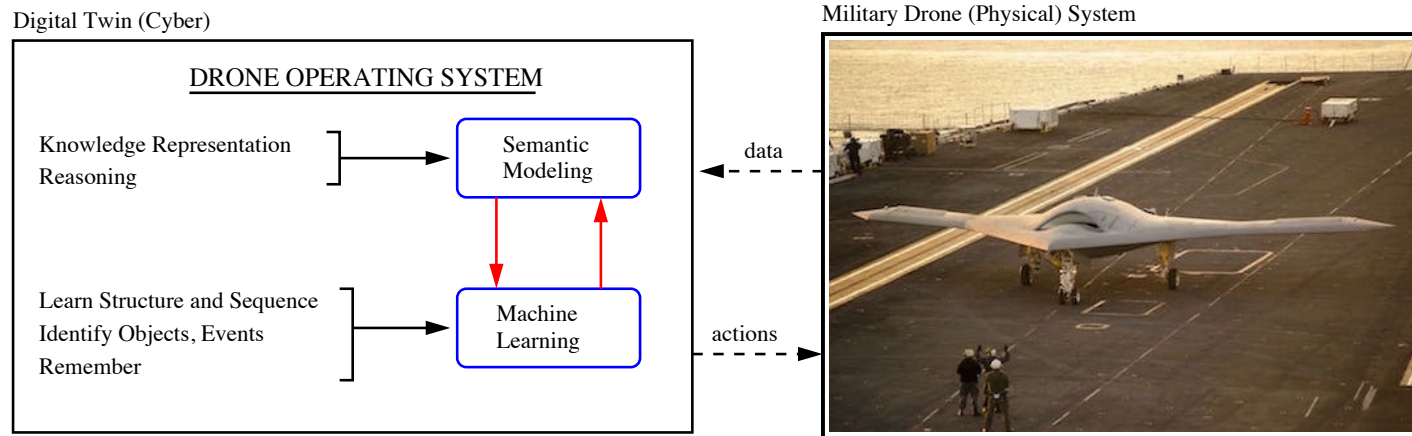
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**Basic Idea:** Explore design of **digital twin architectures** that support **AI and ML** formalisms **working side-by-side** as a team, providing complementary and supportive roles in **collection of data**, **identification of events**, and automated **decision making**.

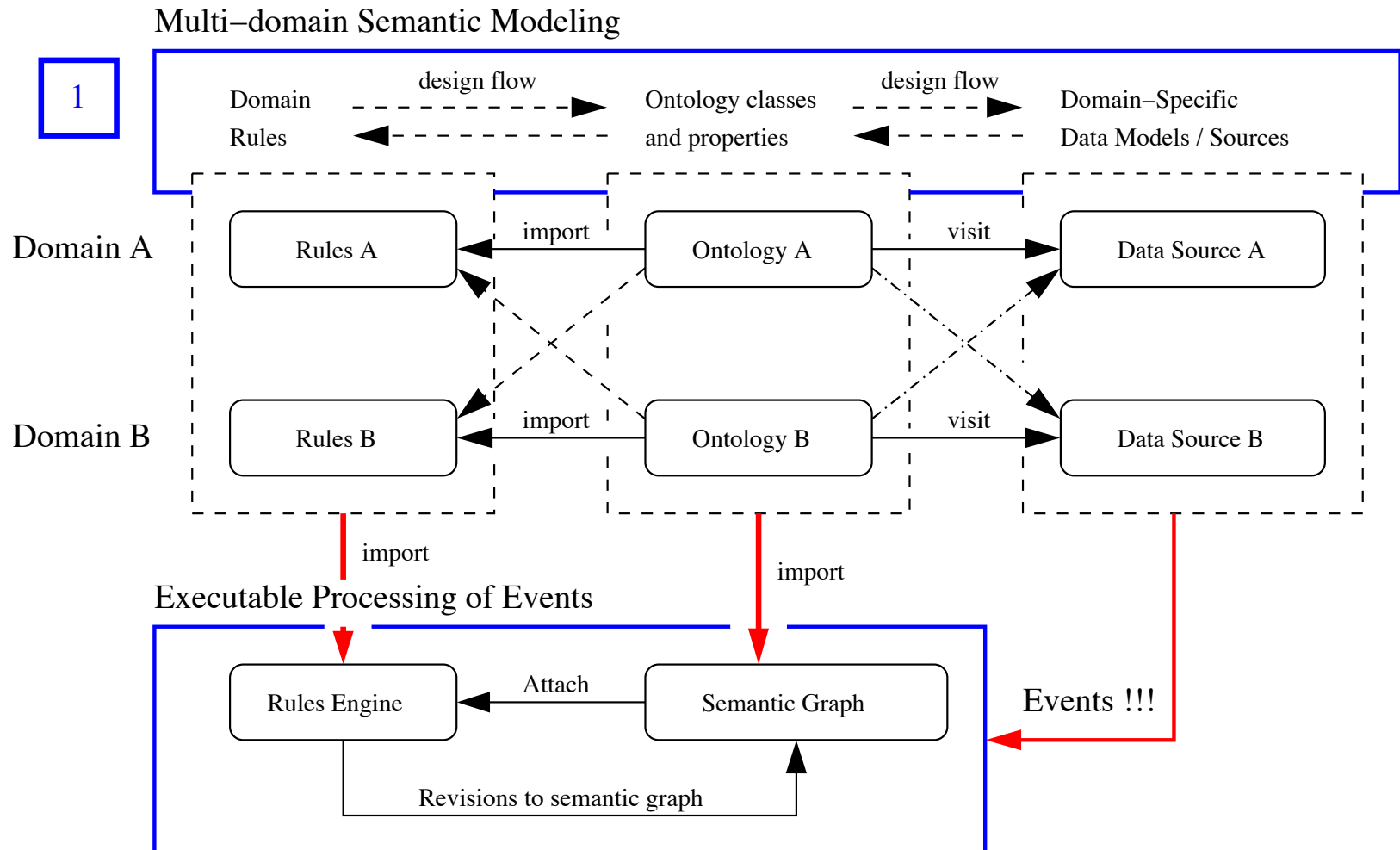


**Research Challenge:** How to design **digital twin elements** and their **interactions** so that collectively they can **support a wide variety** of **systems engineering methods** and **processes**?

**Incubator Goals:** Understand the **range of possibilities** for which **machine learning** of **large-scale graphs** and their attributes support activities in model-centric engineering.

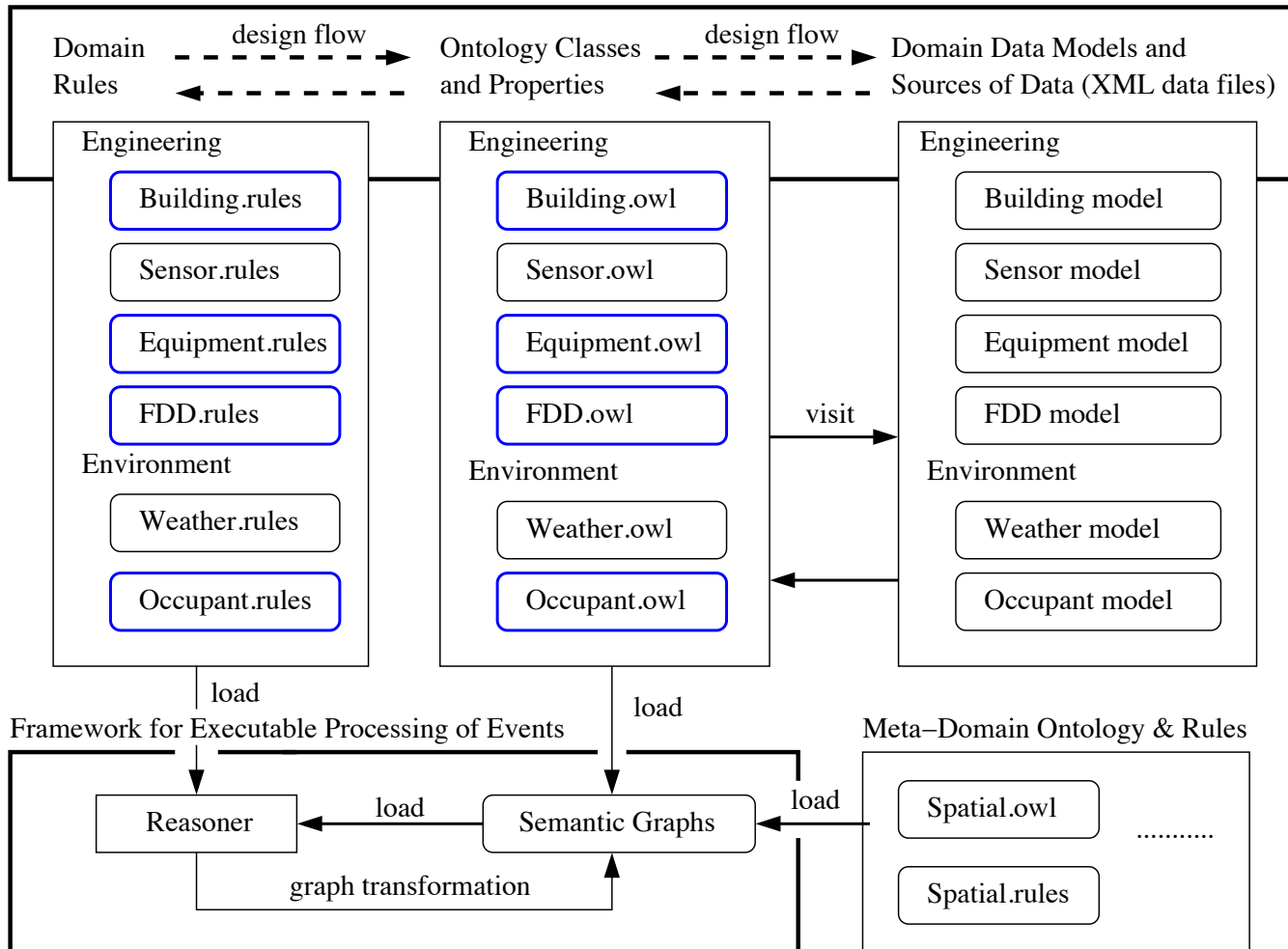
# Architecture for Multi-Domain Semantic Modeling

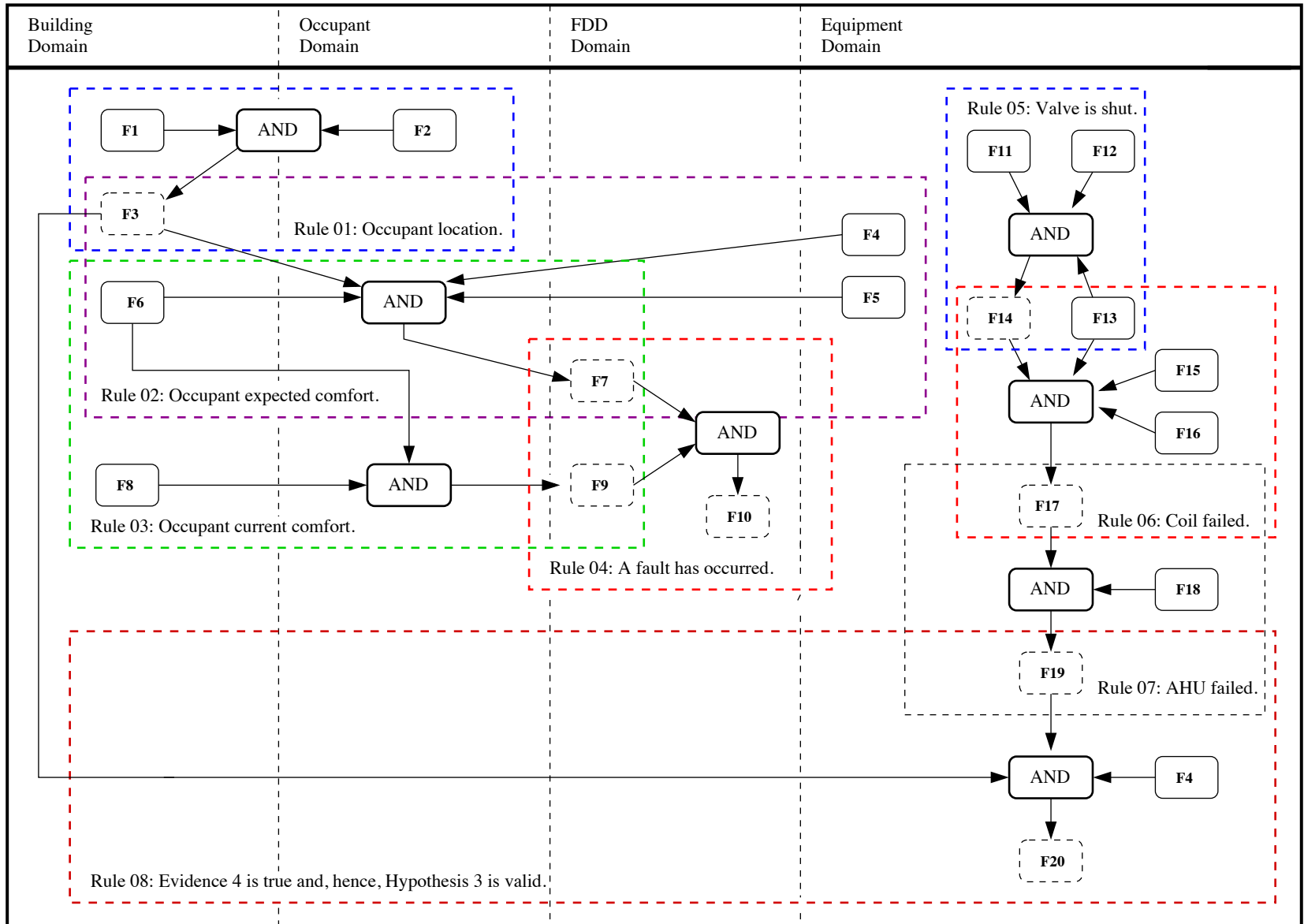
## Step 1: Data-Ontology-Rule Footing (Work at UMD / NIST / SERC in 2017).



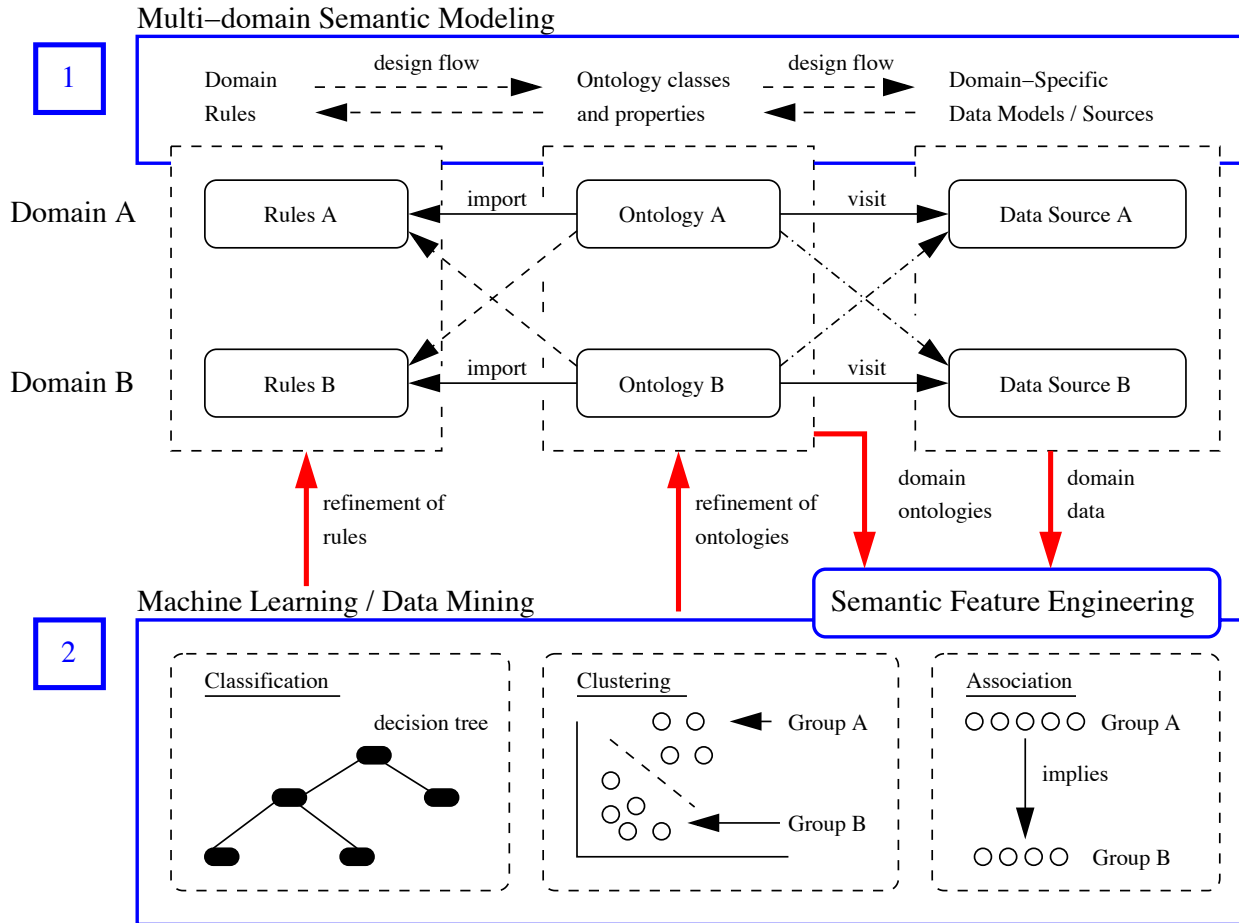
## Example: Detection and Diagnostic Analysis of Faults in HVAC Equipment.

Framework for Concurrent Data-Driven Development of Domain Models, Ontologies and Rules



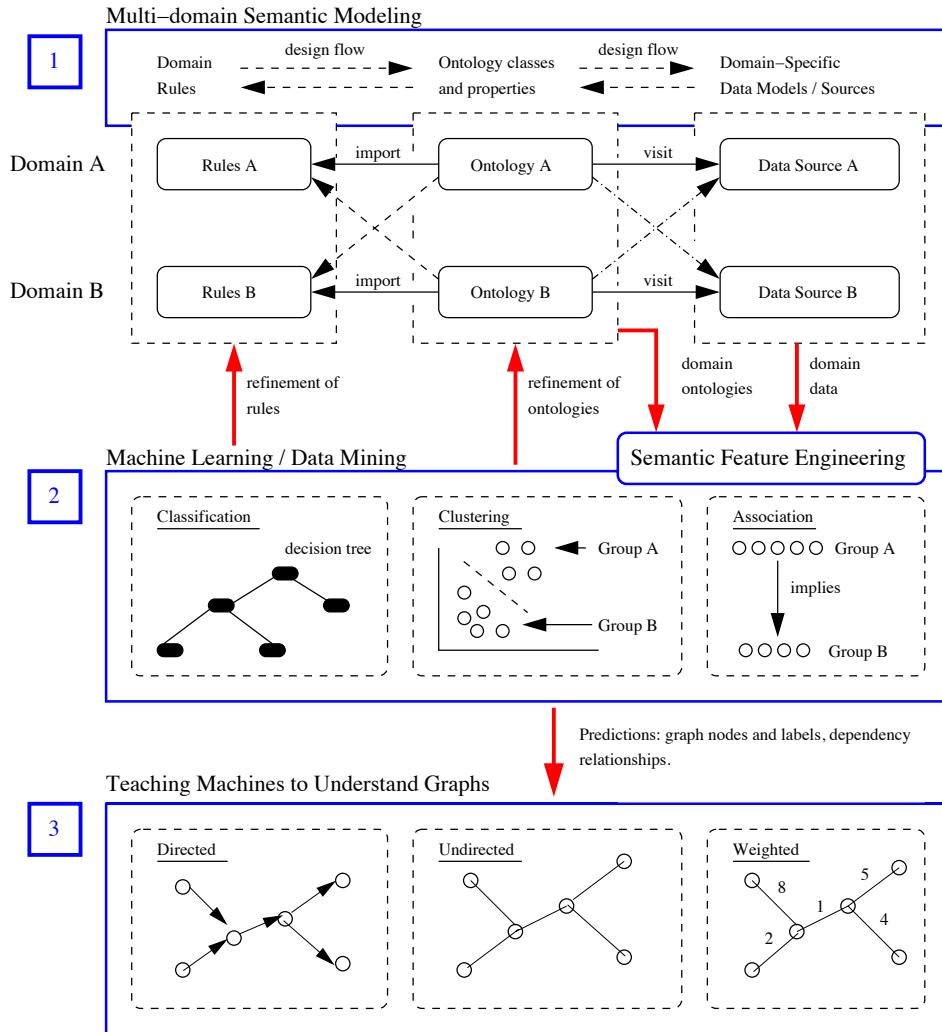


## Step 2: Work at UMD / Building Energy Group at NIST / NCI, 2018-2019



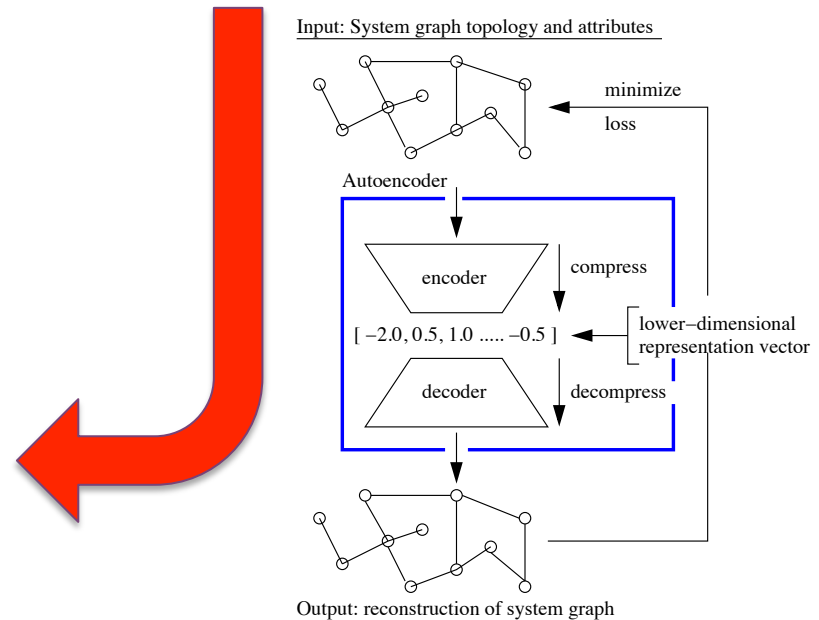
**Research Question:** How can semantic modeling + machine learning / data mining work together as a team?

## Step 3: Focus on Machine Learning of Graphs and Model-Centric Engineering.



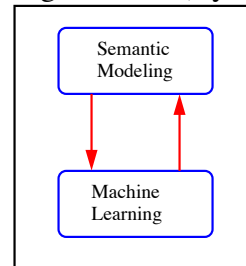
**Observation:** A lot of model-centric engineering boils down to **representation of systems as graphs and sequences of graph transformations punctuated by decision making and work / actions.**

**Hence:** Explore opportunities for **teaching machines to understand graphs.**

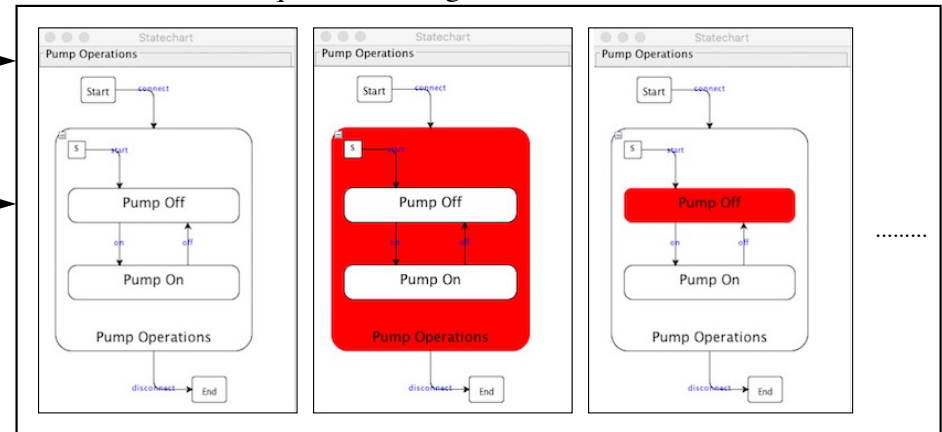


- What **types of graphs** (e.g., undirected, directed, weighted, multi-graph) are **easy** for the **ML to learn**?
- How well do these techniques work with **graph topology** and **attributes** that are **dynamic**?
- What can the ML do that is outside the capability of semantic modeling? And vice-versa?
- How can the **ML improve** the **semantic modeling**? And vice-versa?
- How to **design** the **red arrows** connecting layers 1, 2 and 3?
- How to represent and reason with **uncertainties**?
- How does the difficulty of these challenges increase with **graph size**?
- How to **map AI-ML** capability to **state-of-the-art engineering views**?

Digital Twin (Cyber)



Statechart View: Sequences of Digital Twin Control Actions ...



## Contact Information

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