



# Designing Dynamic Data-Driven Digital Twin Systems in Ecology

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Time	Activity
~ 45 mins	Introduction + Topical Lecture
~ 15 mins	Q&A
~ 10 mins	Pause
~ 45 mins	Exercise: Sample DT Design Schema
~ 10 mins	Close up

- ❖ No coding today, rather a conceptual discussion.
- ❖ Systems Design is subjective, there are always many ways to design things.
- ❖ Workshop website: <https://biadt.github.io/ddd4dt/>

- ❖ **Project name:** Biodiversity Digital Twin for Advanced Modelling, Simulation and Prediction Capabilities (BioDT)
- ❖ **Call title:** Next generation of scientific instrumentation, tools and methods ([HORIZON-INFRA-2021-TECH-01](#))
- ❖ **Duration:** 1 June 2022 – 31 May 2025
- ❖ **Consortium:** 22 partners
  - ❖ 12 countries: Finland (FI), Italy (IT), Czech Republic (CZ), the Netherlands (NL), Estonia (EE), Sweden (SE), United Kingdom (UK), Germany (DE), Austria (AT), Denmark (DK), Norway (NO), Spain (ES)
    - ❖ Incl. one Affiliated Entity and three Associated Partners
- ❖ **Work Package (WP) members:** 140+
- ❖ **Coordinator:** CSC – IT Center for Science
- ❖ **Website:** [www.biadt.eu](http://www.biadt.eu)

A digital twin is a virtual representation of real-world entities and processes, synchronized at a specified **frequency** and **fidelity\***



[Image: digital-strategy.ec.europa.eu](https://digital-strategy.ec.europa.eu/en/policies/digital-twins)

\*Here, fidelity refers to the level of precision captured by the DT in comparison with its physical counterpart.

A **digital twin** (DT) is typically composed of:

- ❖ Data
- ❖ A model that is the representation in terms of behaviour and
- ❖ An application that connects the data and model in a way that makes the outputs of the model relevant, given the specific purpose of the DT

*Since different scopes require different behaviour and fidelity, there cannot be a single twin answering all possible questions*

**Industrial** DTs typically facilitate:

- ❖ Product design
- ❖ Operation of machinery

In **BioDT**, DTs used to:

- ❖ Mimic behaviour observed in nature
- ❖ Meet requirements of BioDT Use Cases
- ❖ Contribute toward EC goal of devising a full DT of the Earth

## Use Cases split into four groups

### Species response to environmental change



❖ Biodiversity dynamics

❖ Ecosystem services

### Genetically detected biodiversity



❖ Crop wild relatives and genetic resources for food security

❖ DNA detected biodiversity, poorly known habitats

### Dynamics and threats from and for species of policy concern



❖ Invasive species

❖ Endangered species

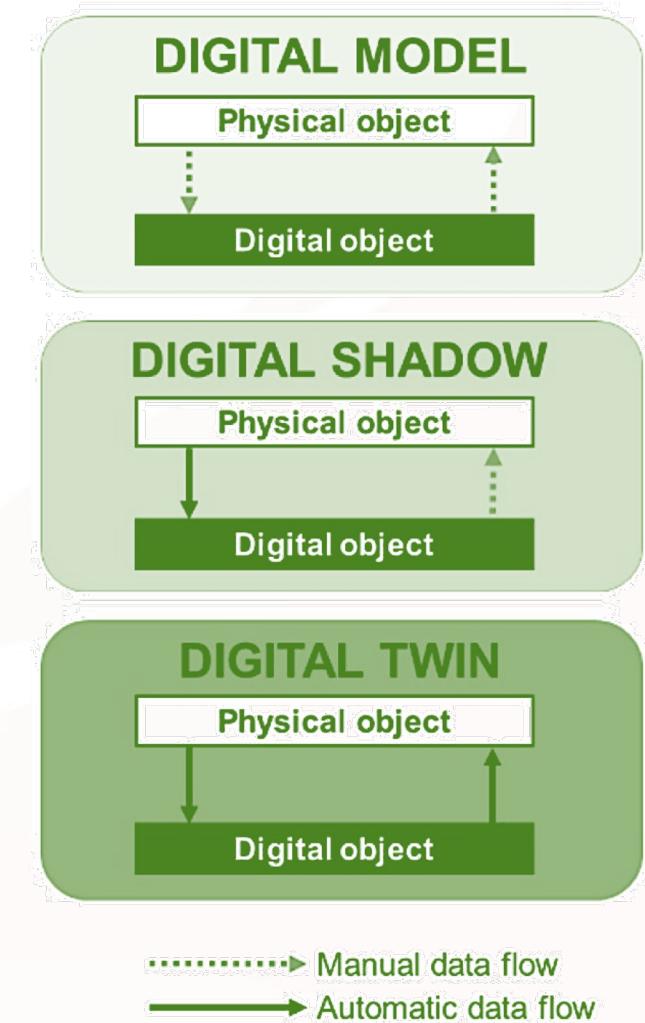
### Species interactions with each other and with humans



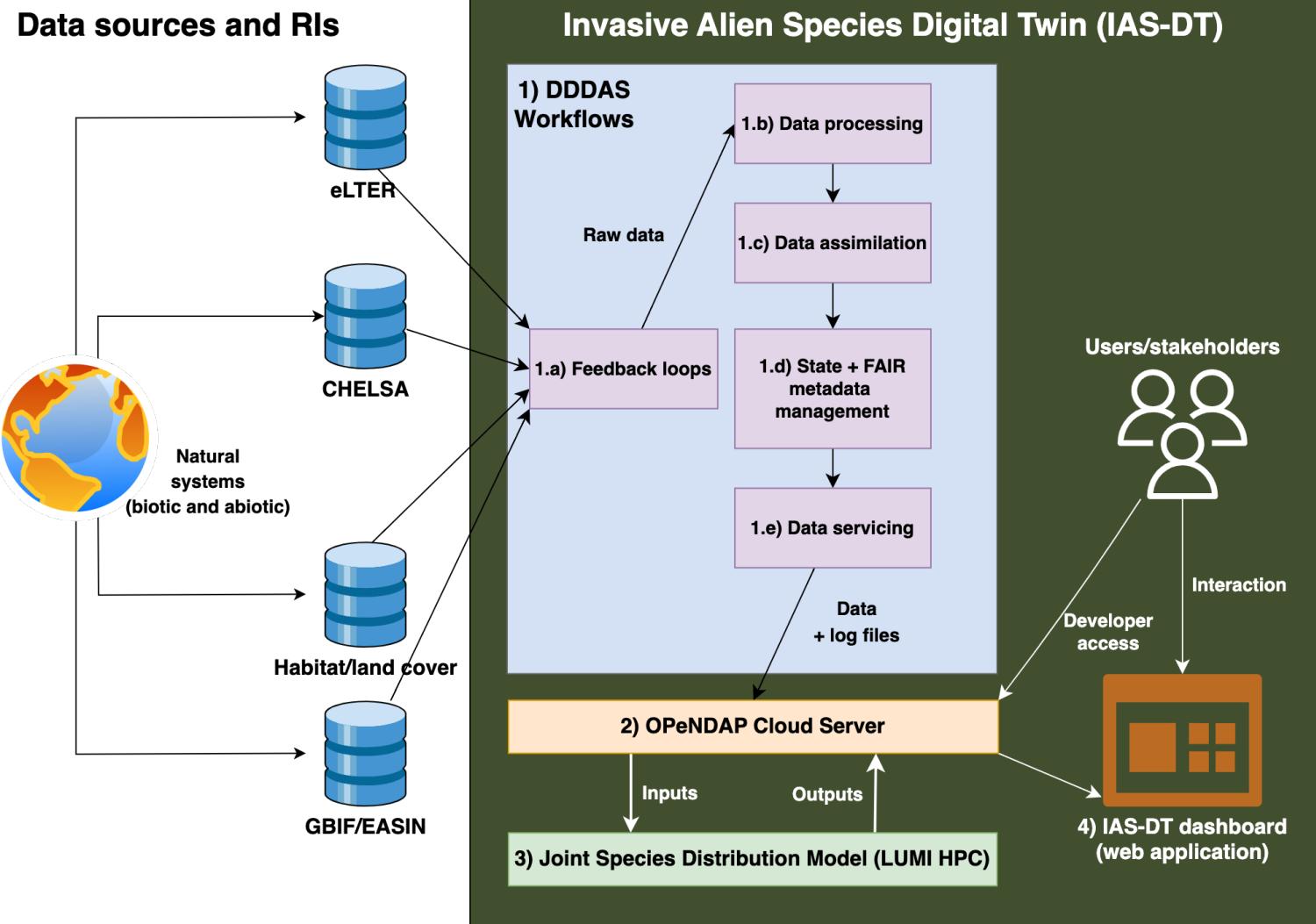
❖ Disease outbreaks

❖ Pollinators

The main difference between DTs, **digital shadows** and **digital models** is the nature and direction of the data flow between the physical and virtual systems.

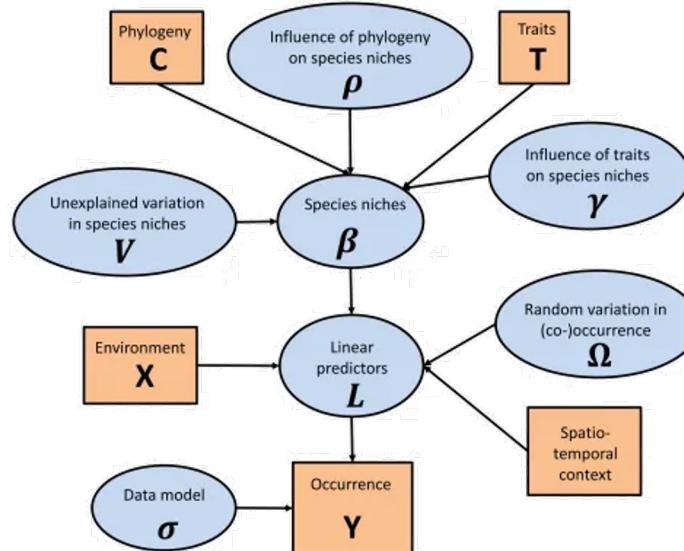


*Source: Open Engineering*

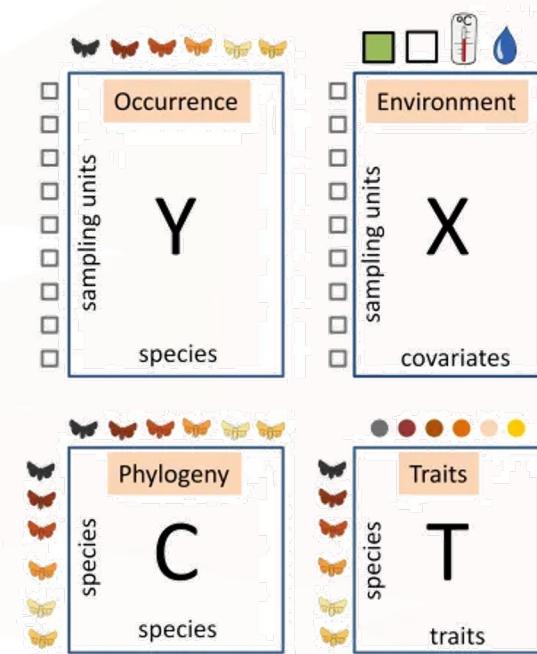
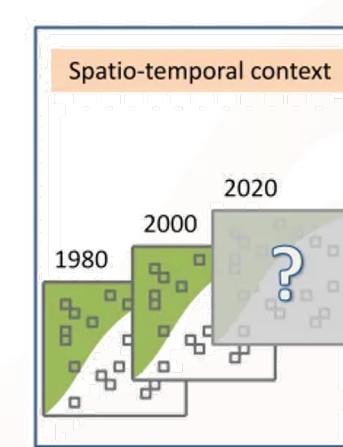


Architecture for Invasive Alien Species Digital Twin (IASDT). Source: Taimur Khan

- ❖ Predictive Digital Twin.
- ❖ State data ranging in ~ 100s of GBs.
- ❖ No direct data collection/sensor access.
- ❖ SDM = Hierarchical Modelling of Species Communities (HMSC).



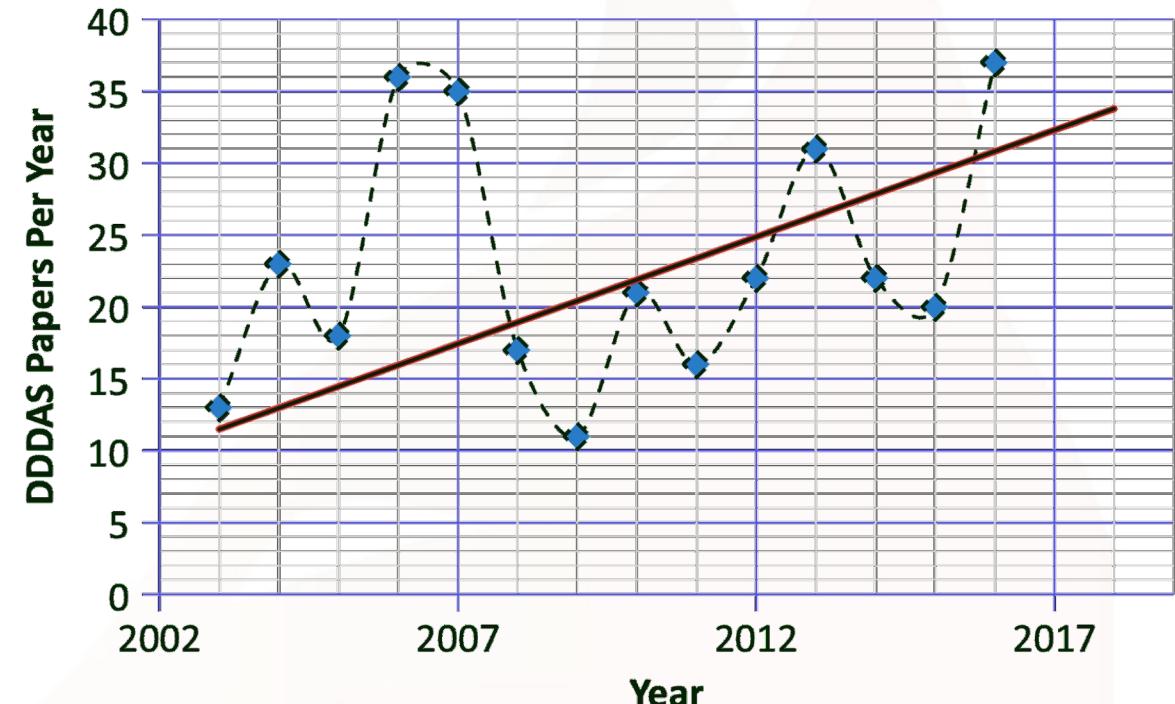
Source: Ovaskainen et al. 2017a



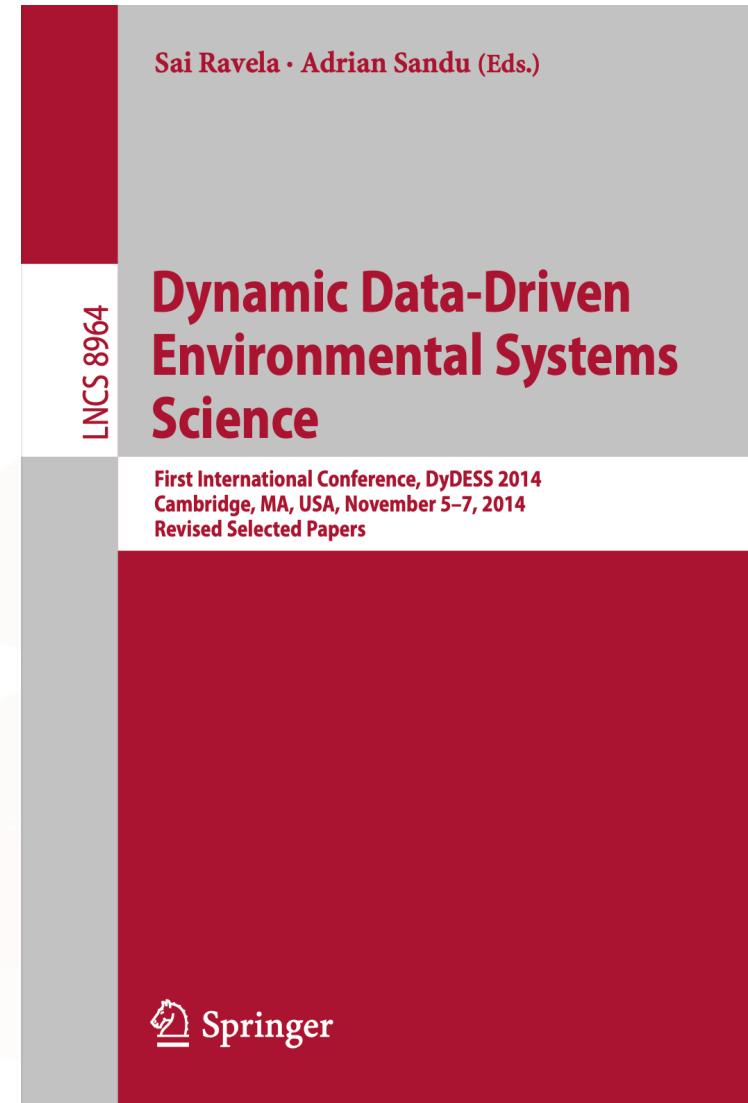
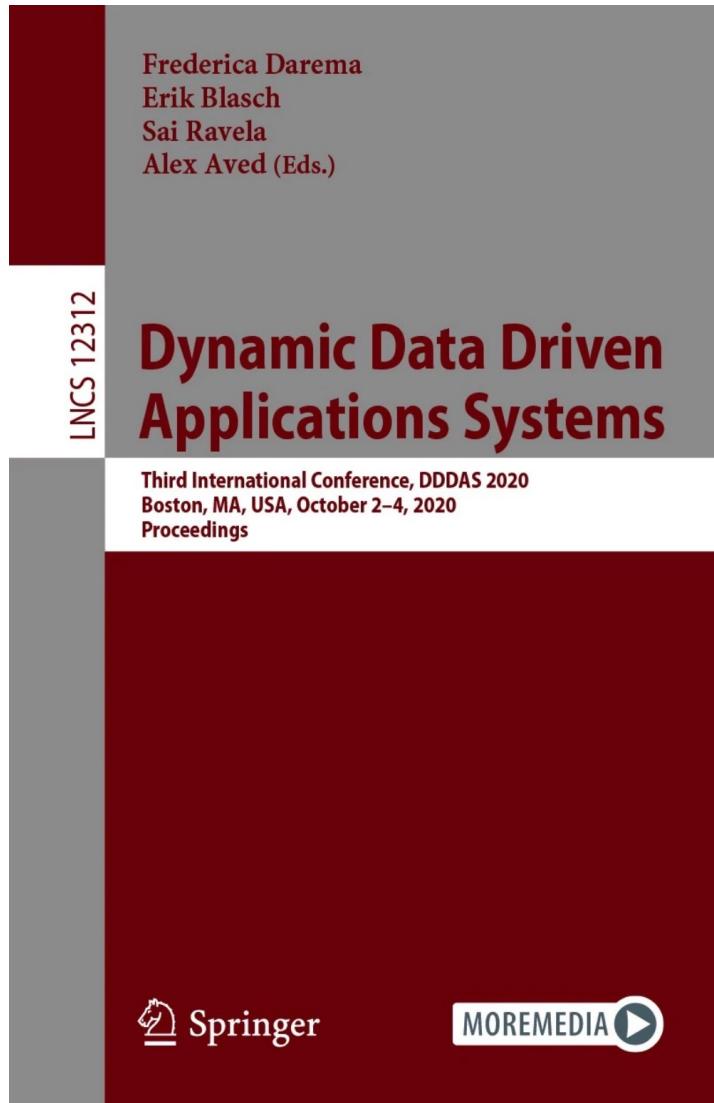
- ❖ No mature DT systems exist, hence a common design is not clear.
- ❖ DT tools are limited to other niches.
- ❖ Literature is sparse for biodiversity DTs systems design.
- ❖ Datasets are updated infrequently, with often lots of heterogeneity.
- ❖ Researchers mostly working with “indirect” data collection.

The Data Driven Applications Systems (DD DAS) concept entails "*the ability to dynamically incorporate data into an executing application simulation, and in reverse, the ability of applications to dynamically steer measurement processes*",

*creating "application simulations that can dynamically accept and respond to 'online' field data and measurements and/or control such measurement."*

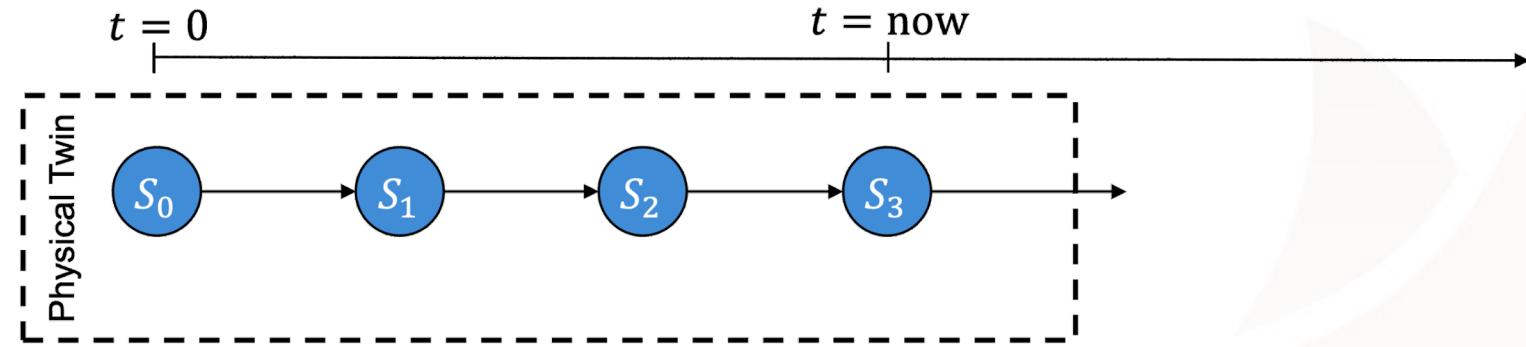


DDDAS papers per year. Source: Handbook of DDDAS (2018).



**Physical State, S:**

Parameterized state of  
the physical asset

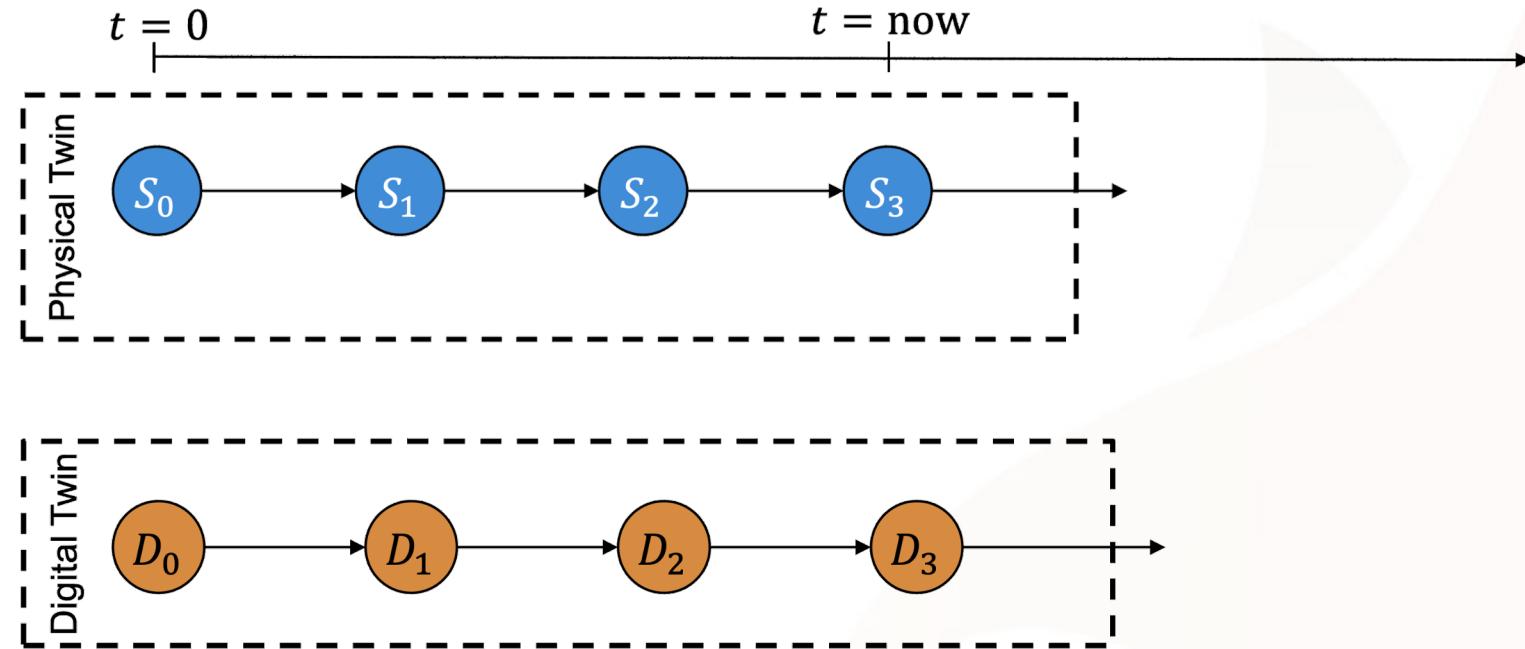


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**Digital State, D:**

Parameters (model inputs) that define the computational models comprising the digital twin



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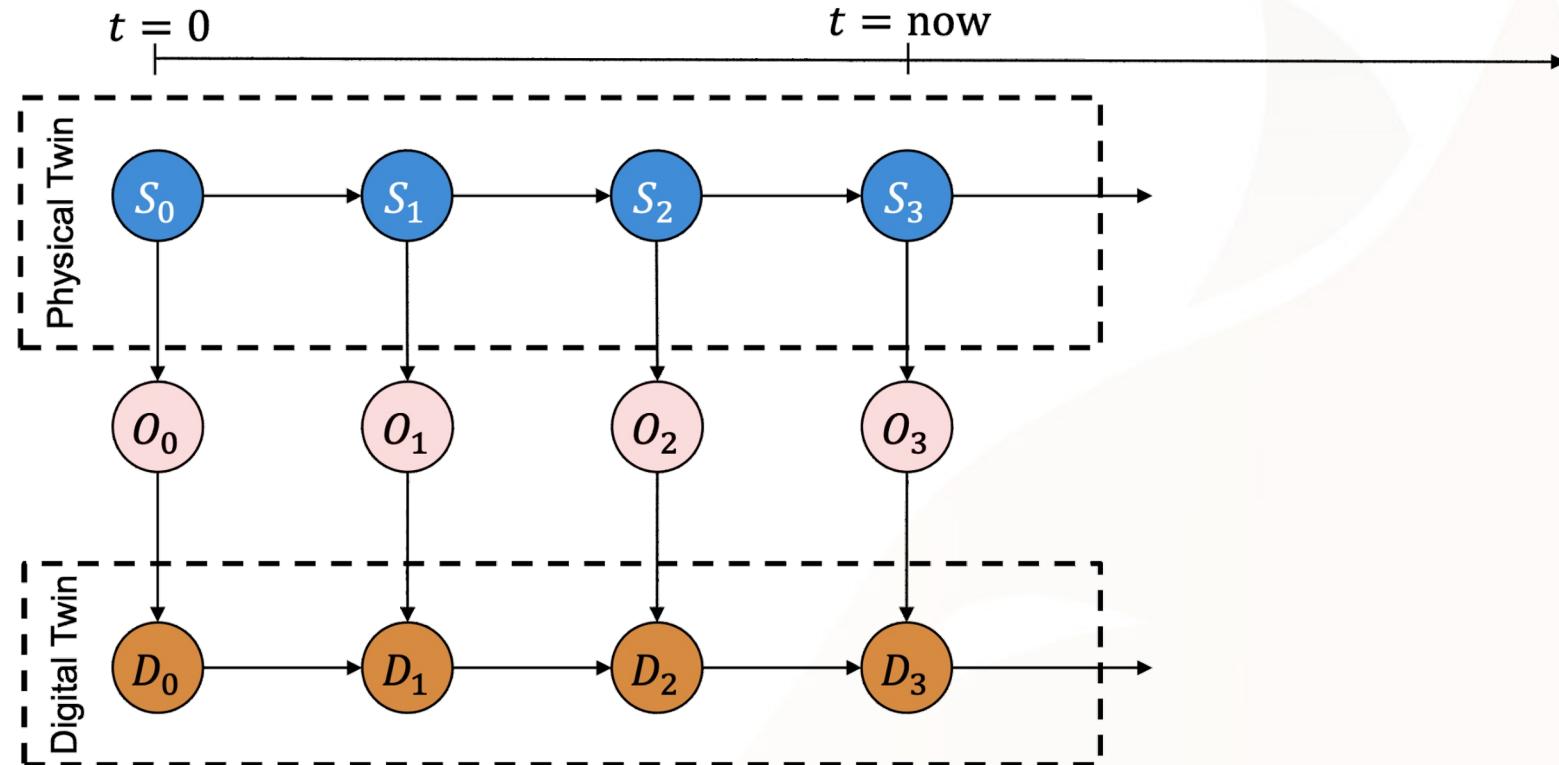
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**Observational data, O:**

Available information describing the state of the physical asset



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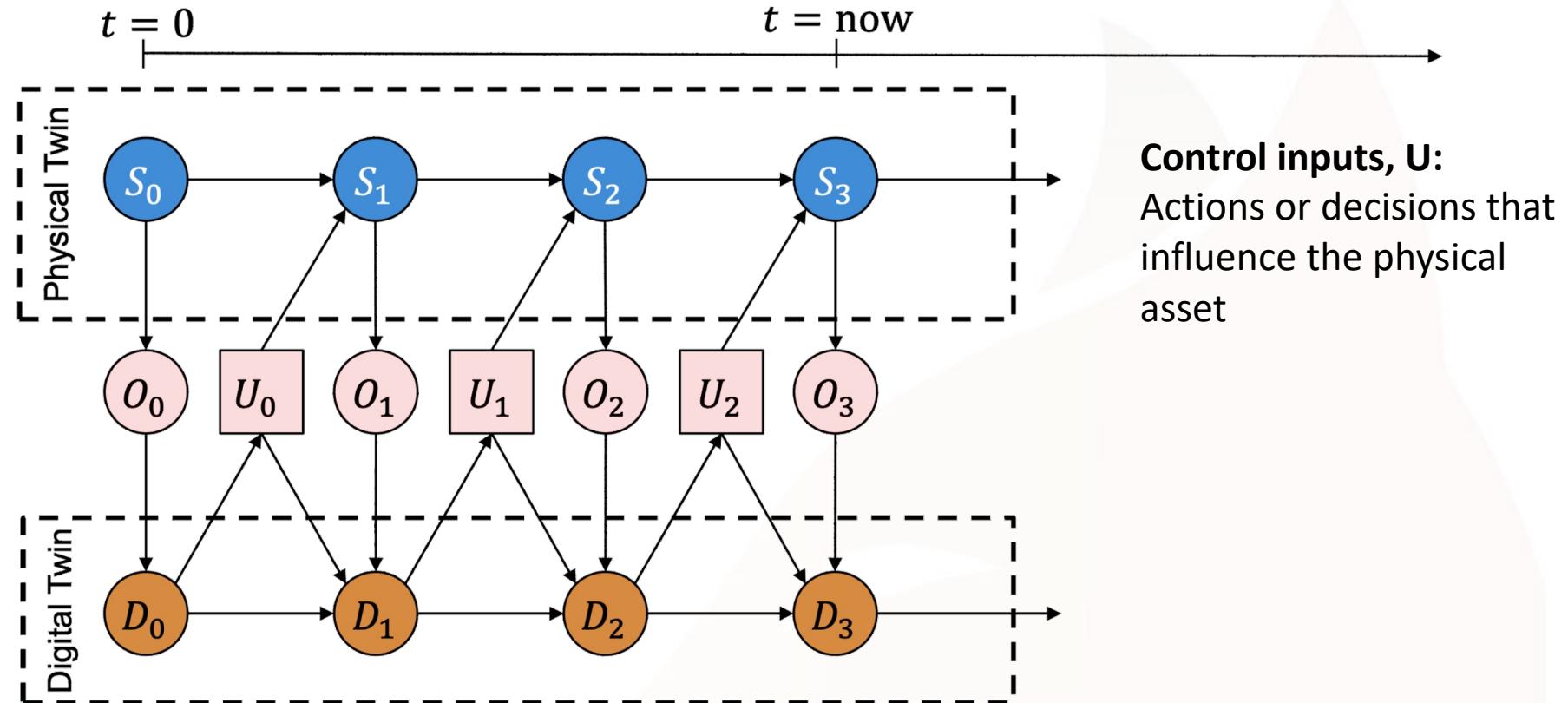
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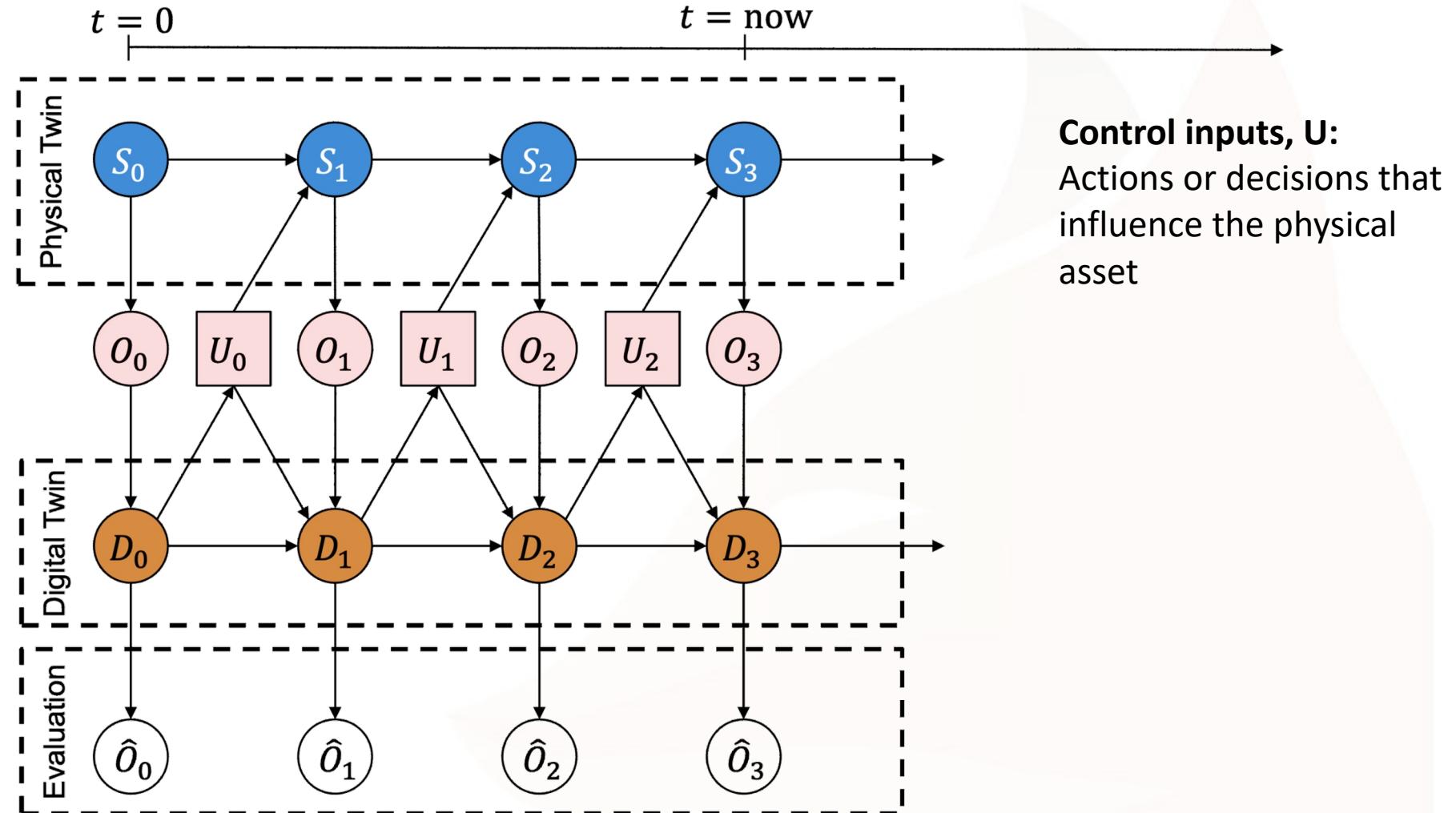
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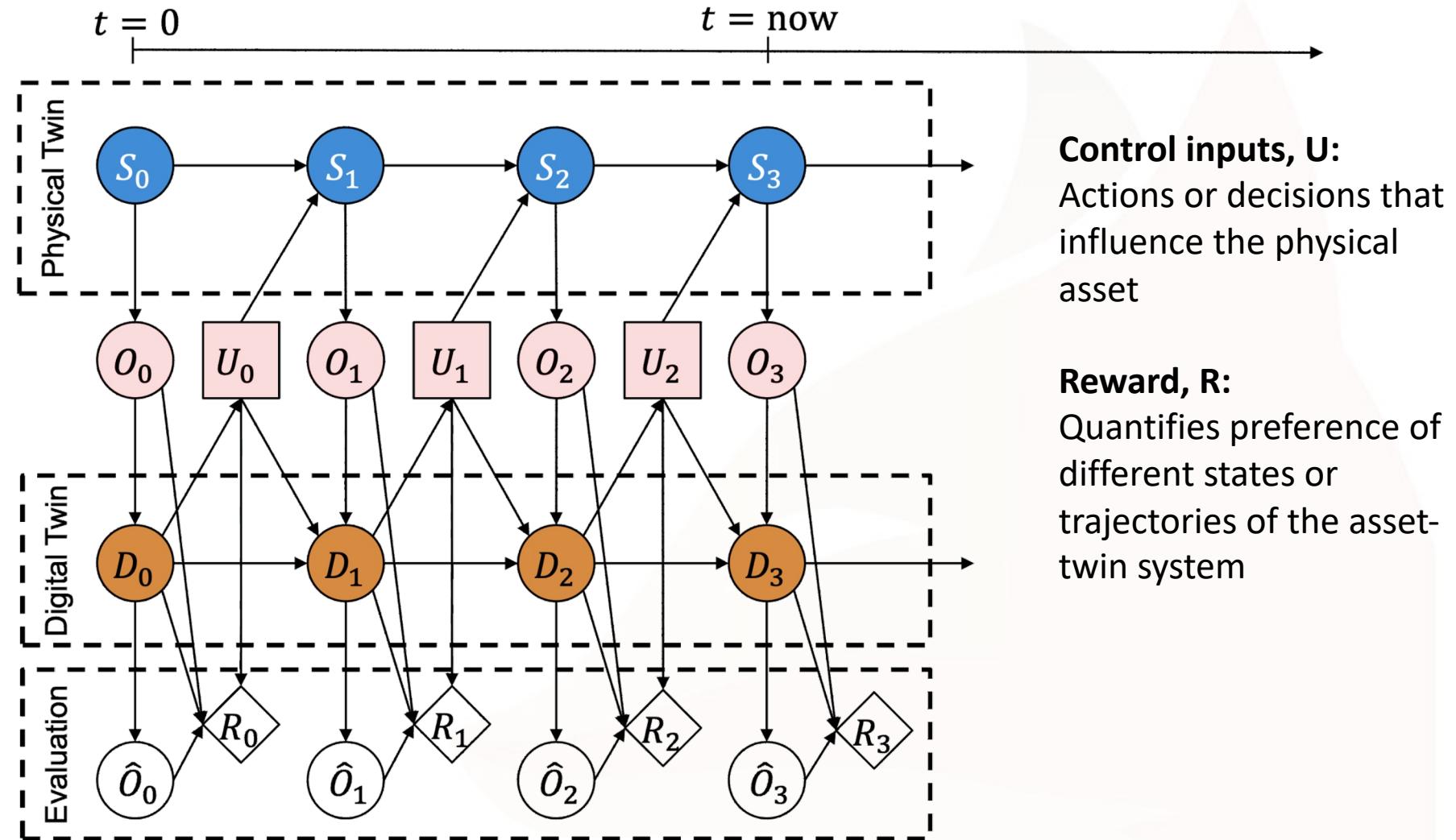
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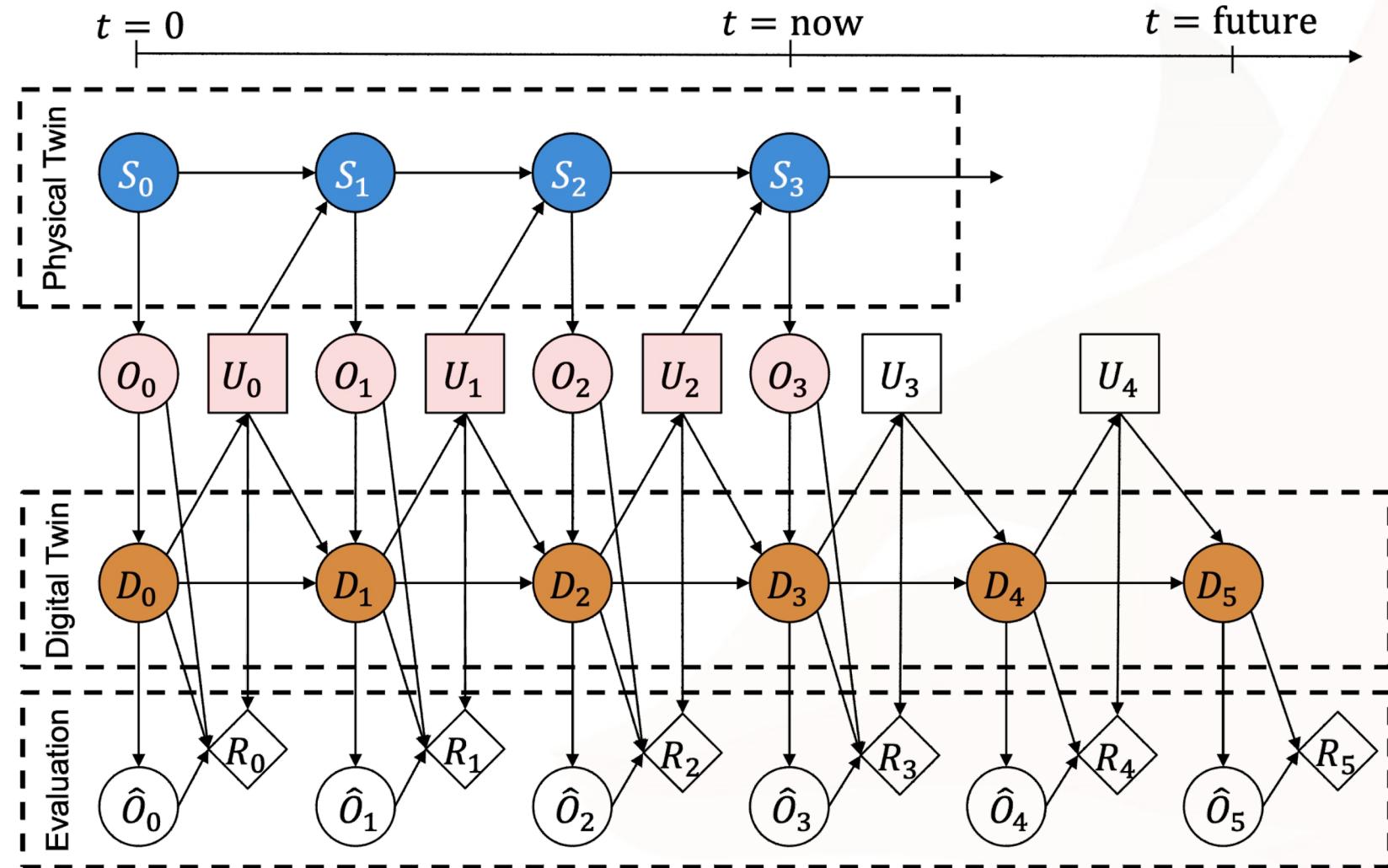
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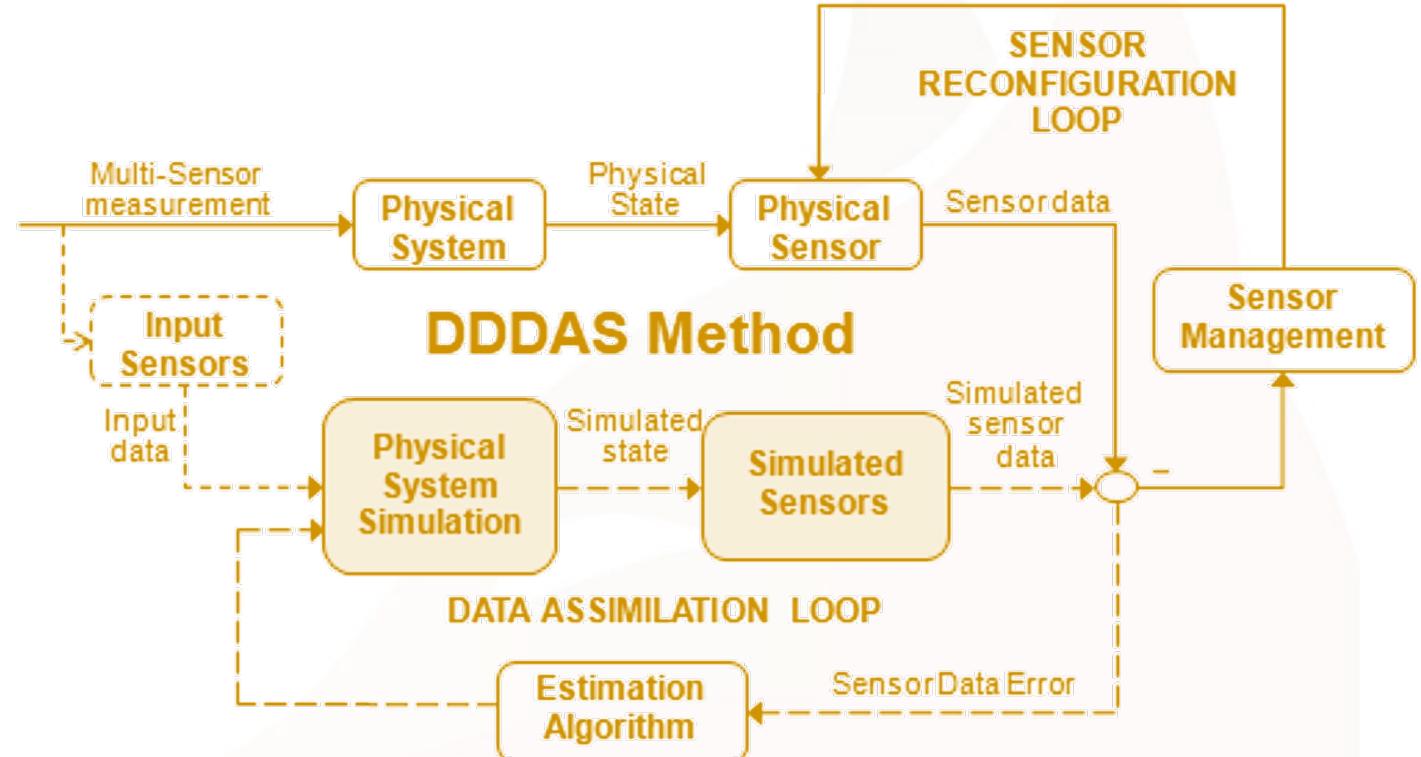
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Available information describing the state of the physical asset





- ❖ Based on Dynamic Data-Driven Application Systems (DDDAS) paradigm.
- ❖ DDDAS Components:
  - ❖ Feedback loop.
  - ❖ State management.
  - ❖ Sensor reconfiguration loop.
- ❖ Other components:
  - ❖ Data servicing.
  - ❖ Model.
  - ❖ User interface.



Sample DDDAS system. Source: [1dddas.org](http://1dddas.org)

## Feedback Loop

- **vSensors** - “Listen” for changes in data sources.
- **Intakers** - Pull new data.
- **Processors** - Process the data.
- **Assimilators** - assimilate the data into existing datasets.
- **Actuators** - change control inputs.
- **Loggers** - Assign FAIR metadata.

## State Management

- **State Space** – defined states of the system.
- **Trackers** - Track state of the data.
- **Synchronizers** - State synchronization.
- **Sniffers** - Detect changes in state of the DT.

❖ **System:** Soil Watering DT

❖ 8 soil beds with sensor network and watering system

❖ **Observational Data (from sensors):**

- ❖ Soil moisture (%)
- ❖ Soil temperature (F)

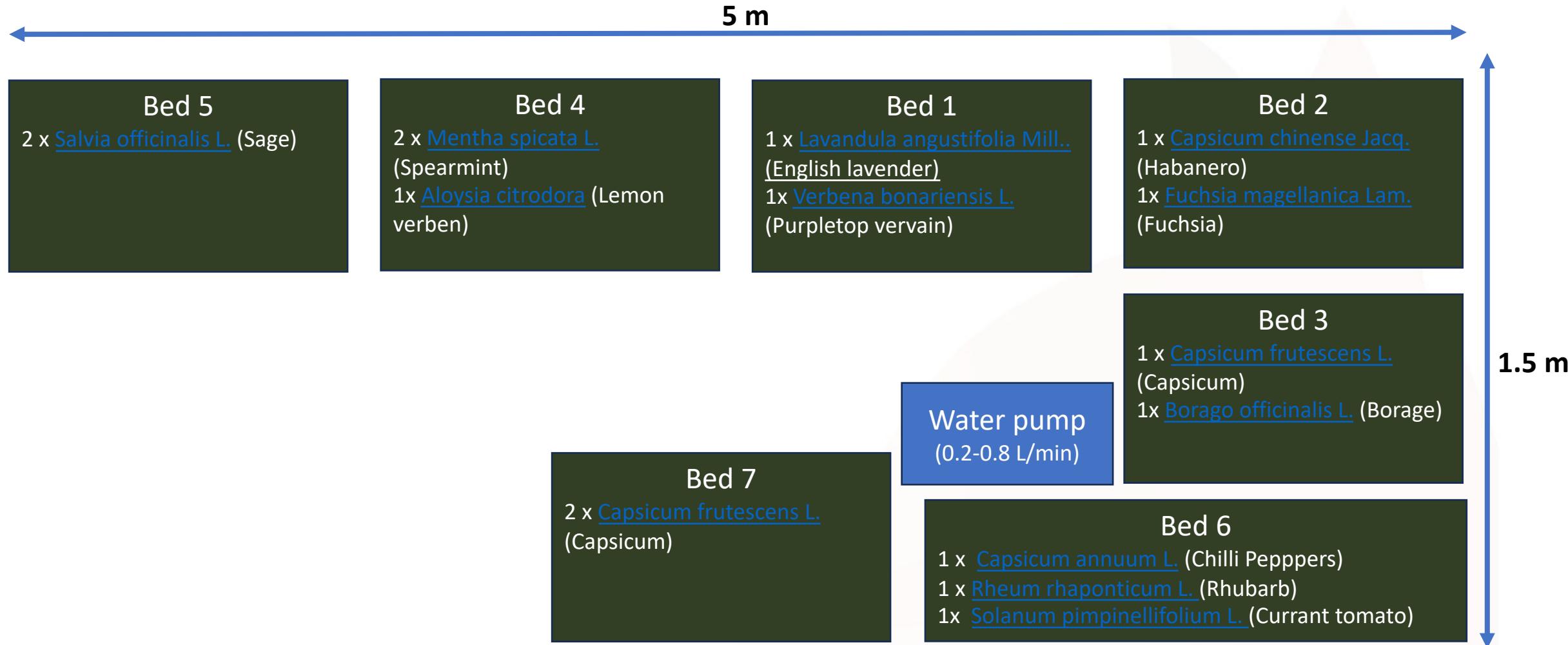
❖ **Control inputs:**

- ❖ WiFi controlled water pumps with on/off states.

❖ **Model:** ? (*e.g. linear regression, rates-of-change*)

❖ Use the DT schema template on draw.io to create a DDDAS-based DT of the given Soil-Plant system that automates soil watering based on soil moisture and soil temperature data.

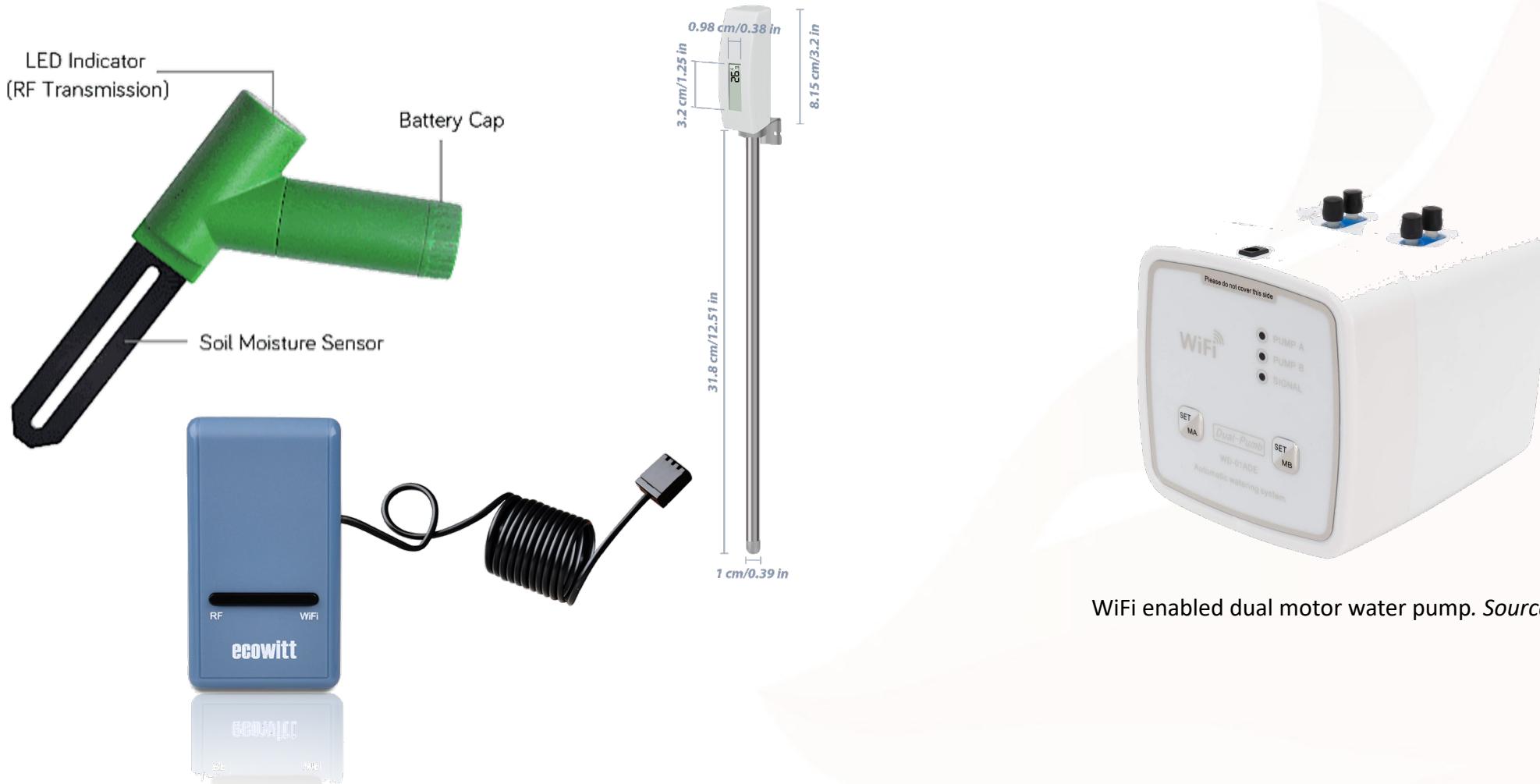
*HINT: Think about what other data sources can be added, what is the state space, what type of model, is needed and what components would be required.*



Relative positions and numbering of soil beds and water pump. Source: Taimur Khan.

## Exercise: Design a DT around given system





Soil moisture sensor, soil temperature sensor, base station. Source: [ecowitt.com](http://ecowitt.com)

WiFi enabled dual motor water pump. Source: [Cikonielj](http://Cikonielj).



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