



# 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://biodt.github.io/dddas4dt/>

- 🔥 **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.biodt.eu](http://www.biodt.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)

\*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



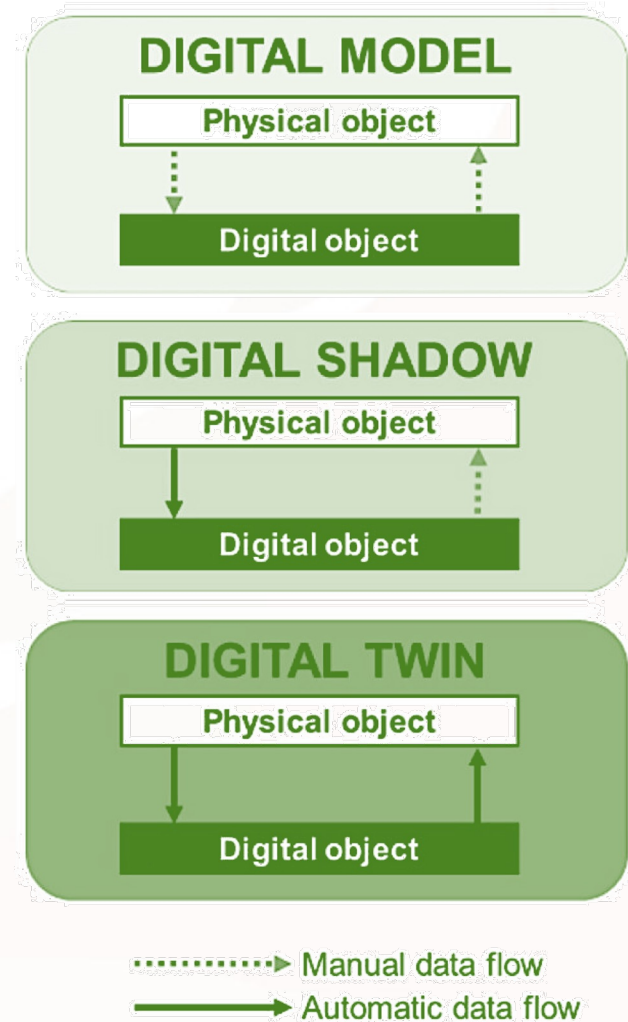
-  Invasive species
-  Endangered species

### Species interactions with each other and with humans

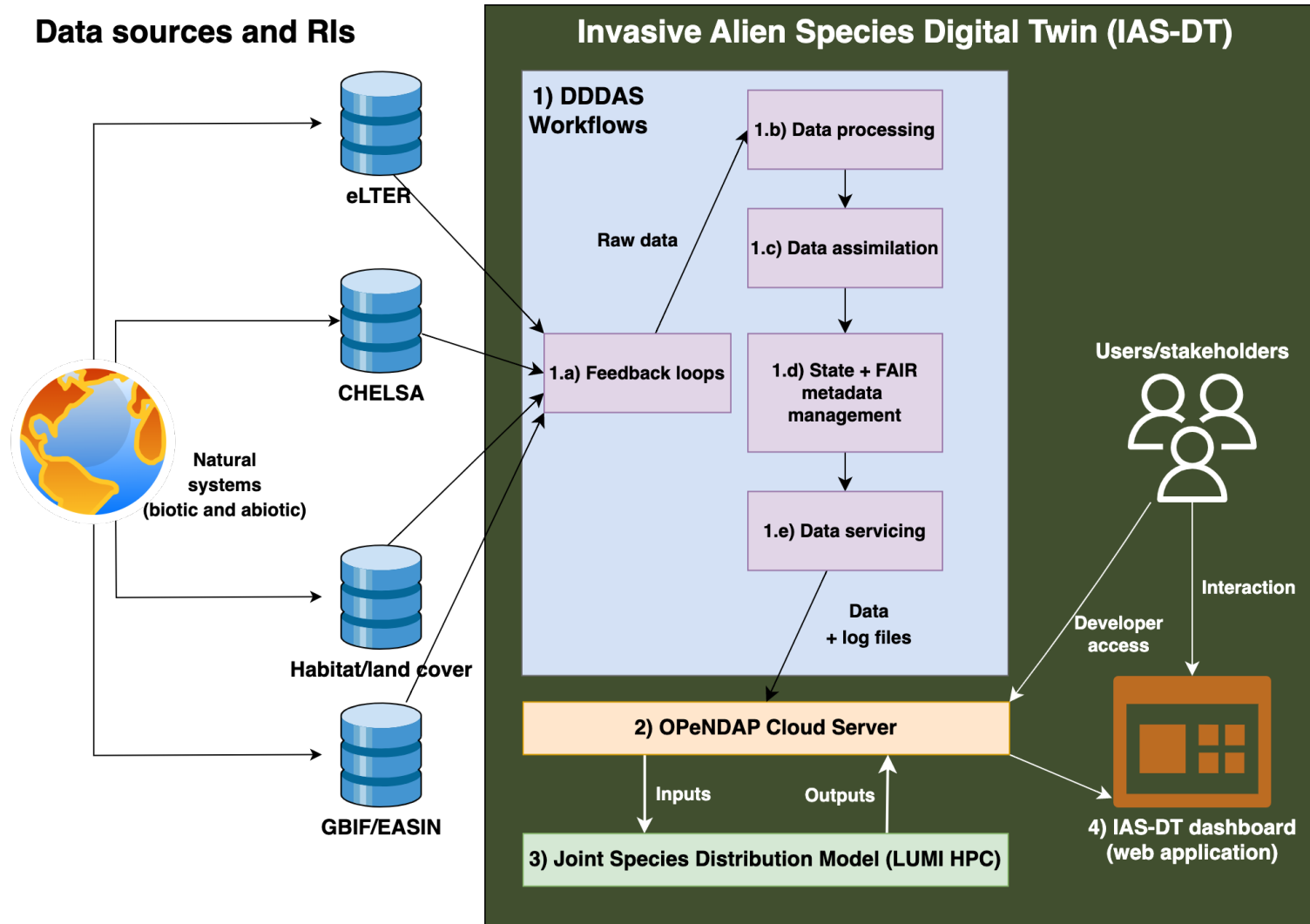


-  Disease outbreaks
-  Pollinators

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.



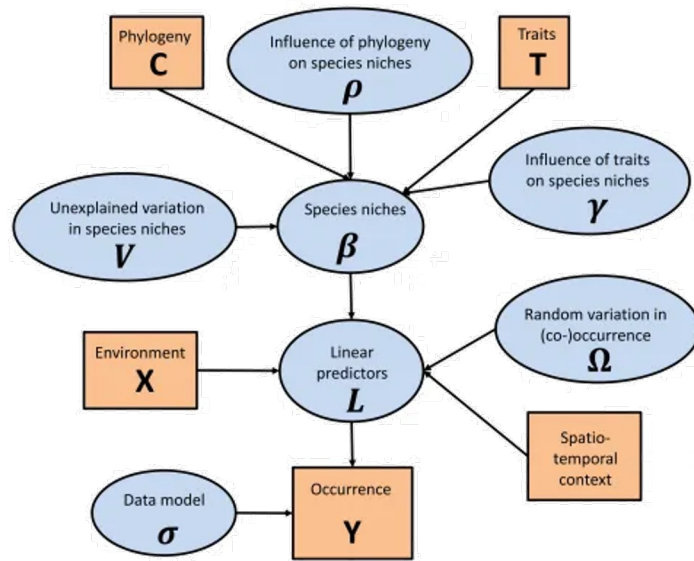
*Source: Open Engineering*



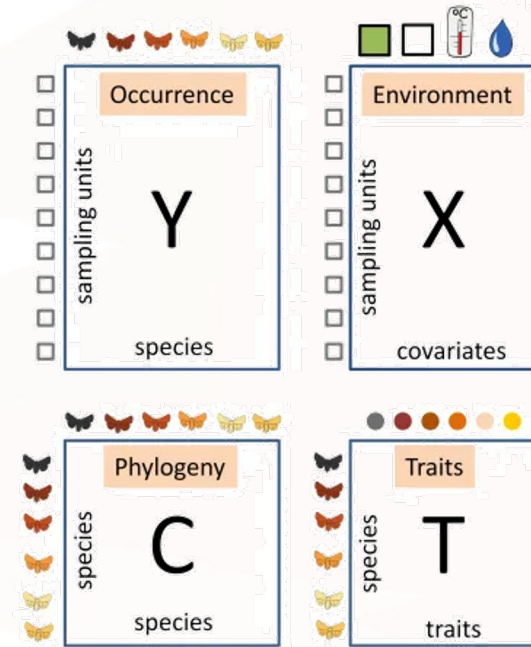
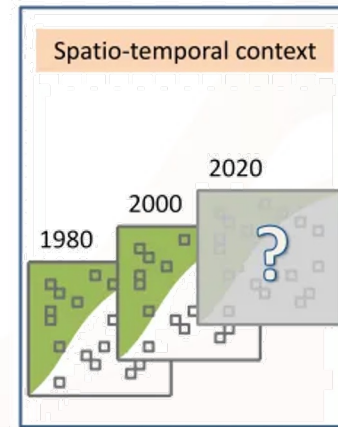
Architecture for Invasive Alien Species Digital Twin (IASDT). Source: Taimur Khan



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

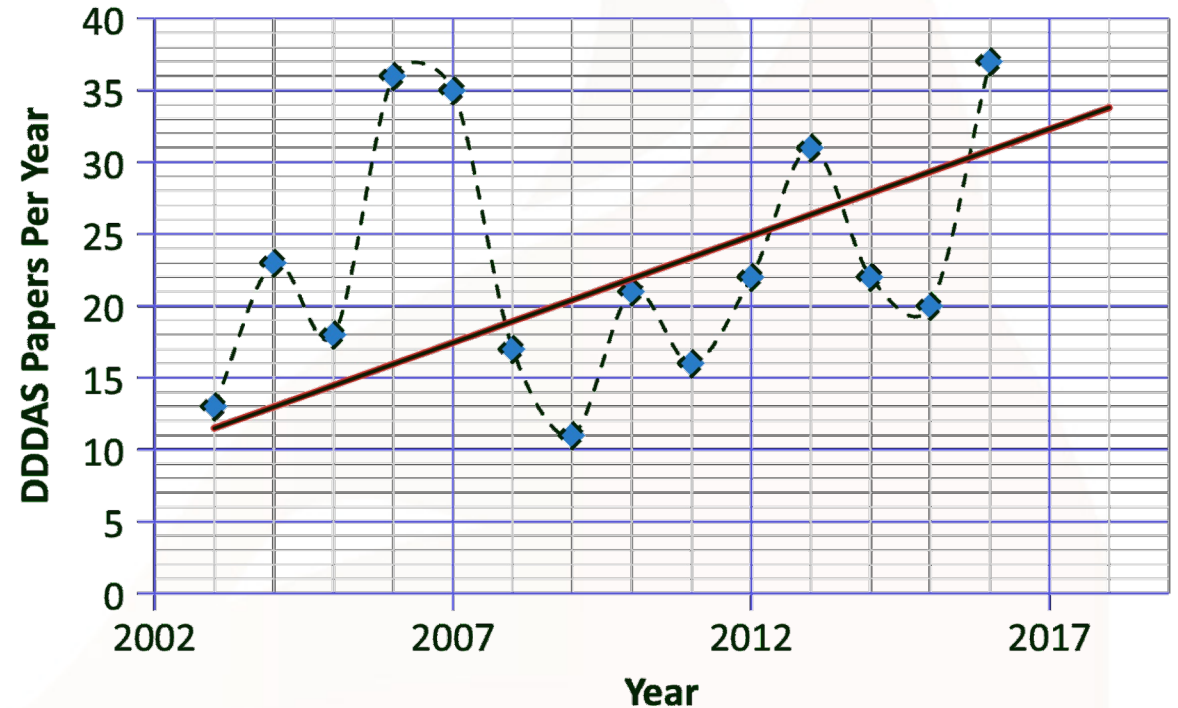


Source: Ovaskainen et al. 2017a

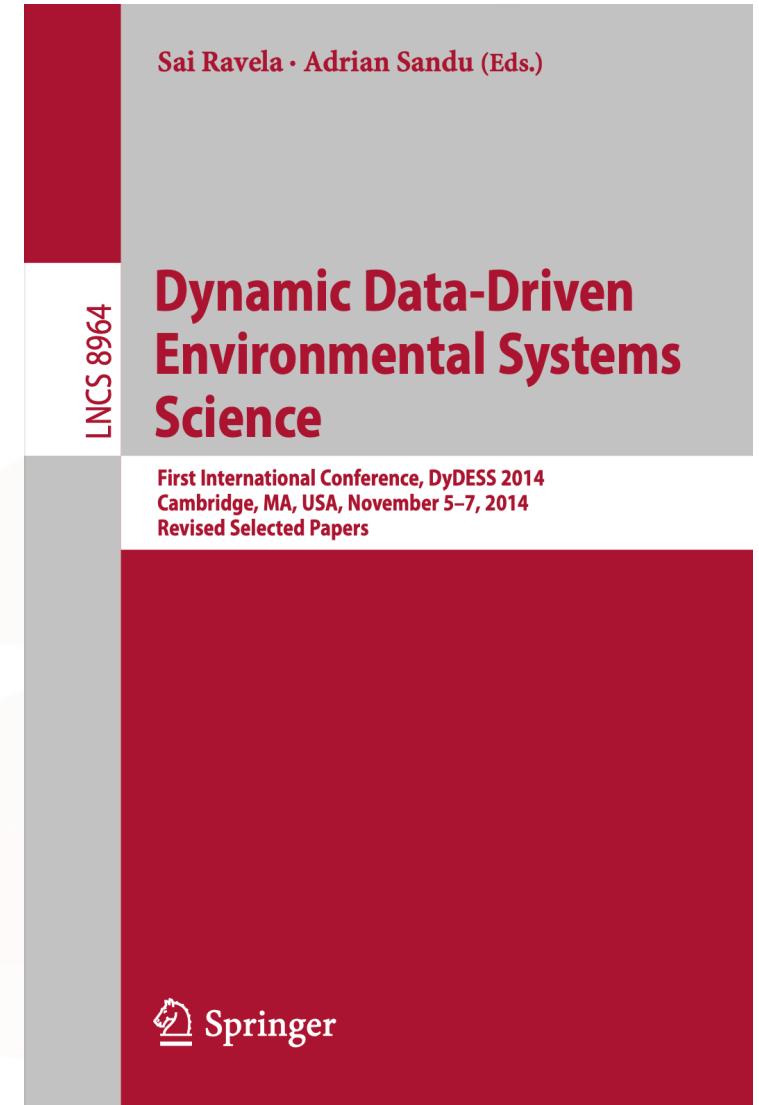
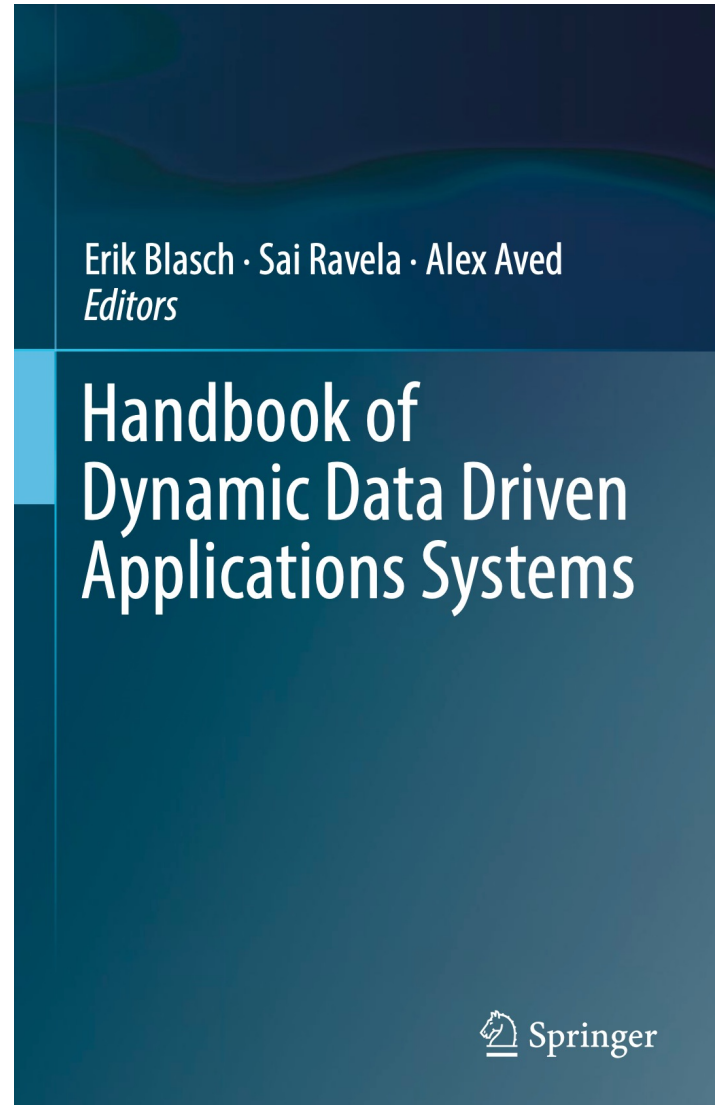
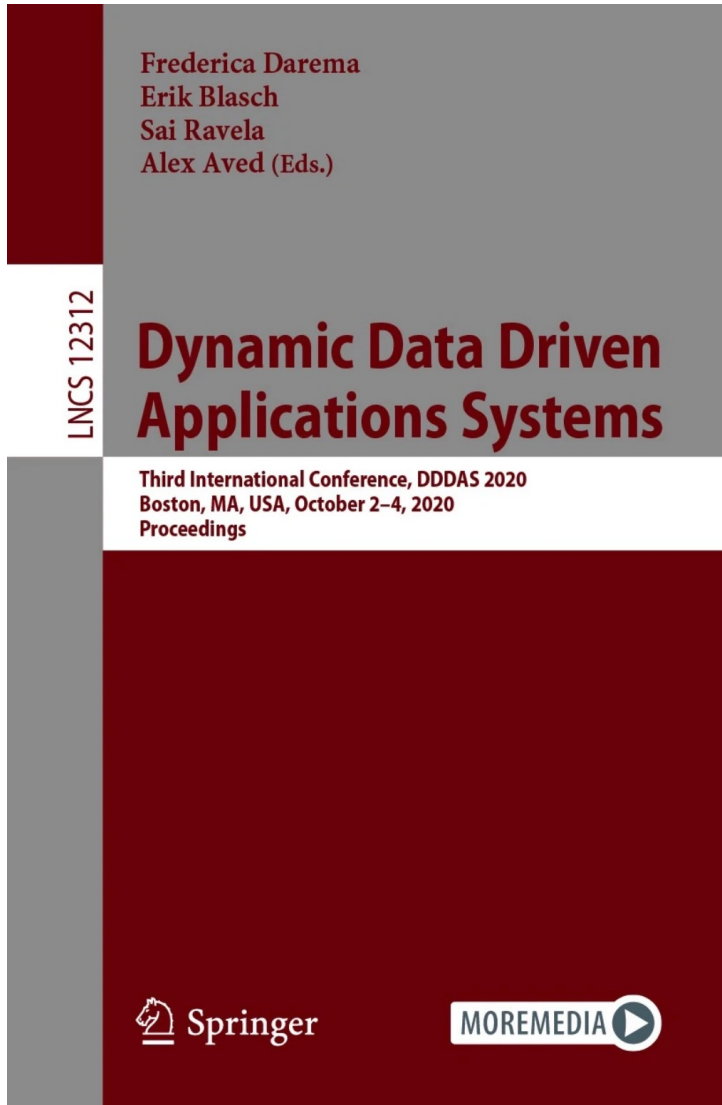


- ❖ No mature DT systems exist, hence a common design is not clear.
- ❖ DT tools are limited to other niches.
- ❖ Literature is sparse for biodiversity DTs systems design.
- ❖ Datasets are updated infrequently, with often lots of heterogeneity.
- ❖ Researchers mostly working with “indirect” data collection.

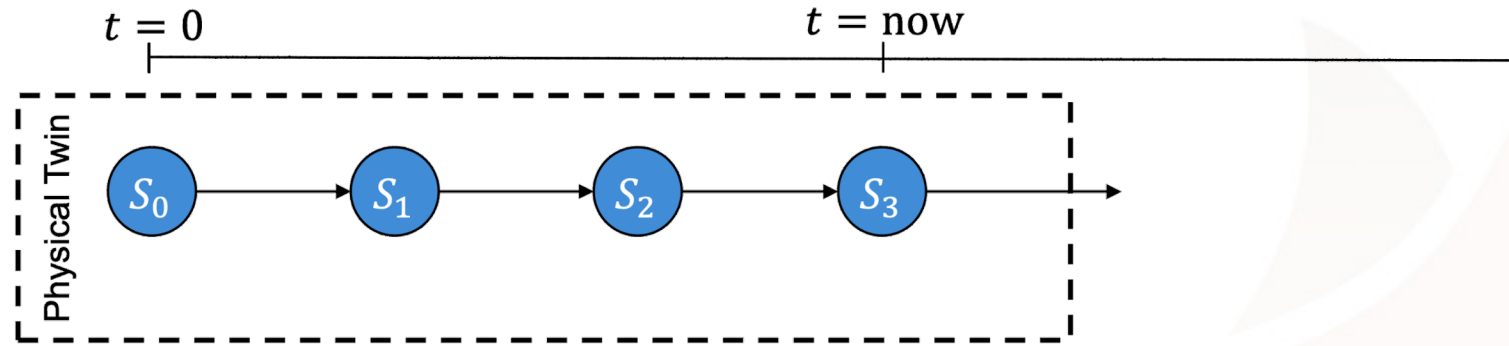
The Data Driven Applications Systems (DD DAS) concept entails *"the ability to dynamically incorporate data into an executing application simulation, and in reverse, the ability of applications to dynamically steer measurement processes"*,  
 creating *"application simulations that can dynamically accept and respond to 'online' field data and measurements and/or control of such measurement."*



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



**Physical State, S:**  
Parameterized state of  
the physical asset

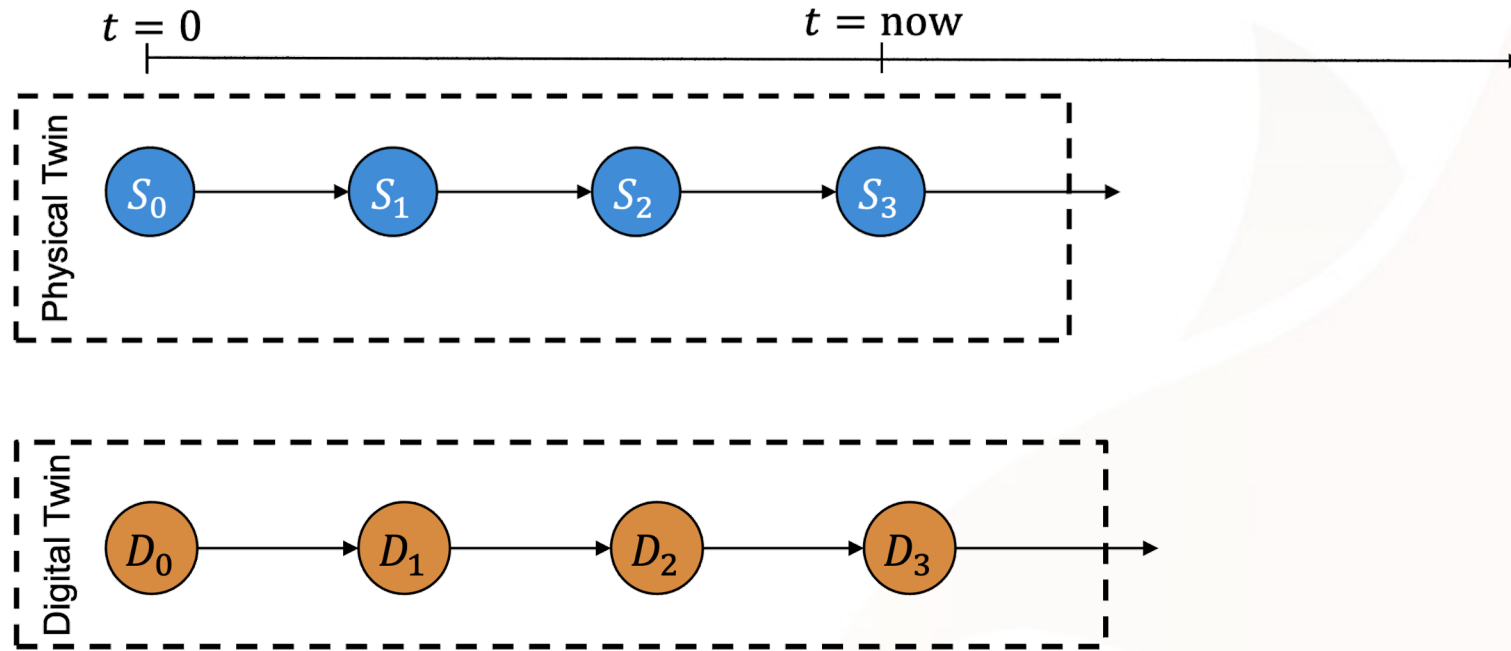


**Physical State, S:**

Parameterized state of the physical asset

**Digital State, D:**

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



**Physical State, S:**

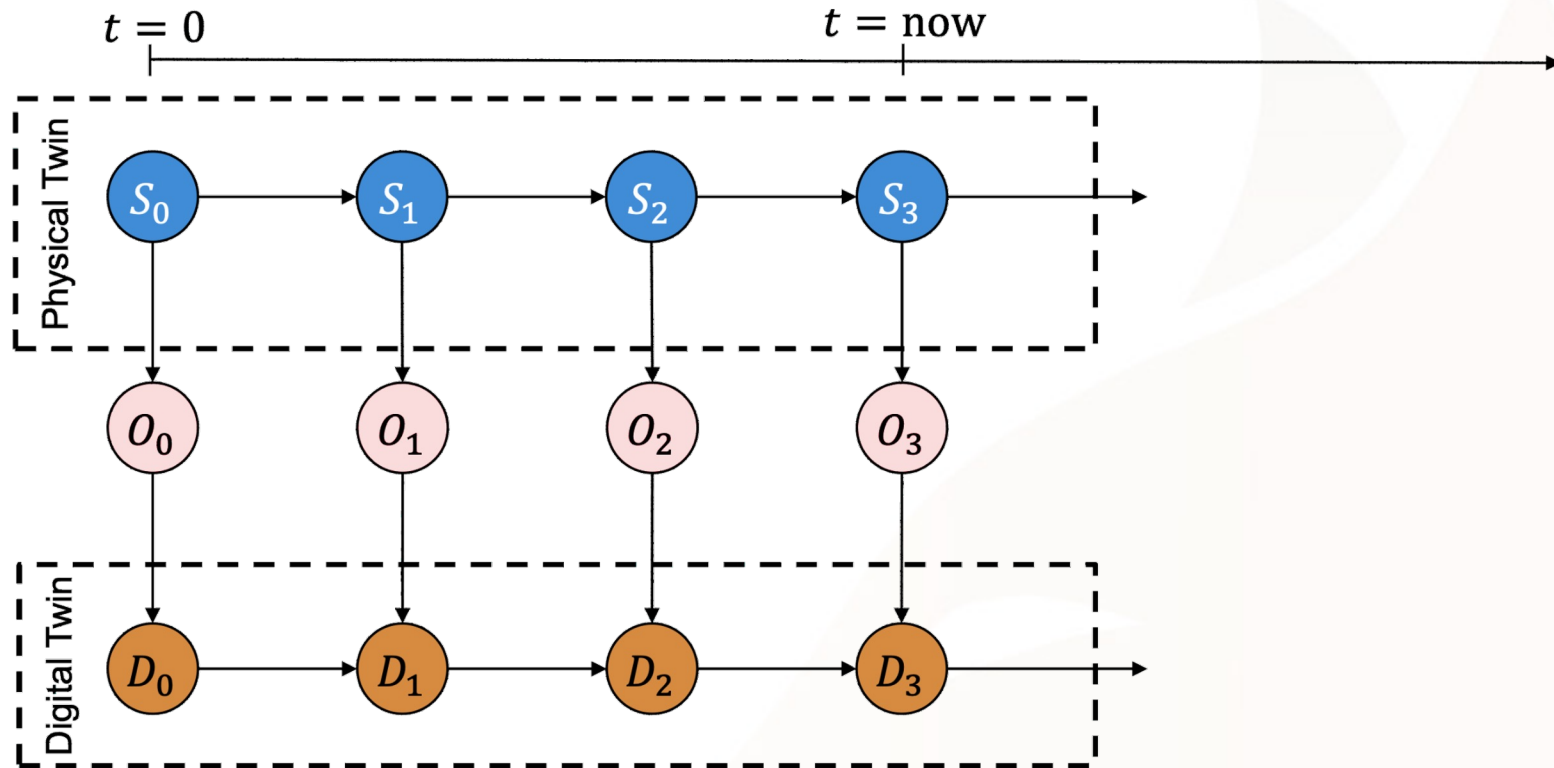
Parameterized state of the physical asset

**Digital State, D:**

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

**Observational data, O:**

Available information describing the state of the physical asset



**Physical State, S:**

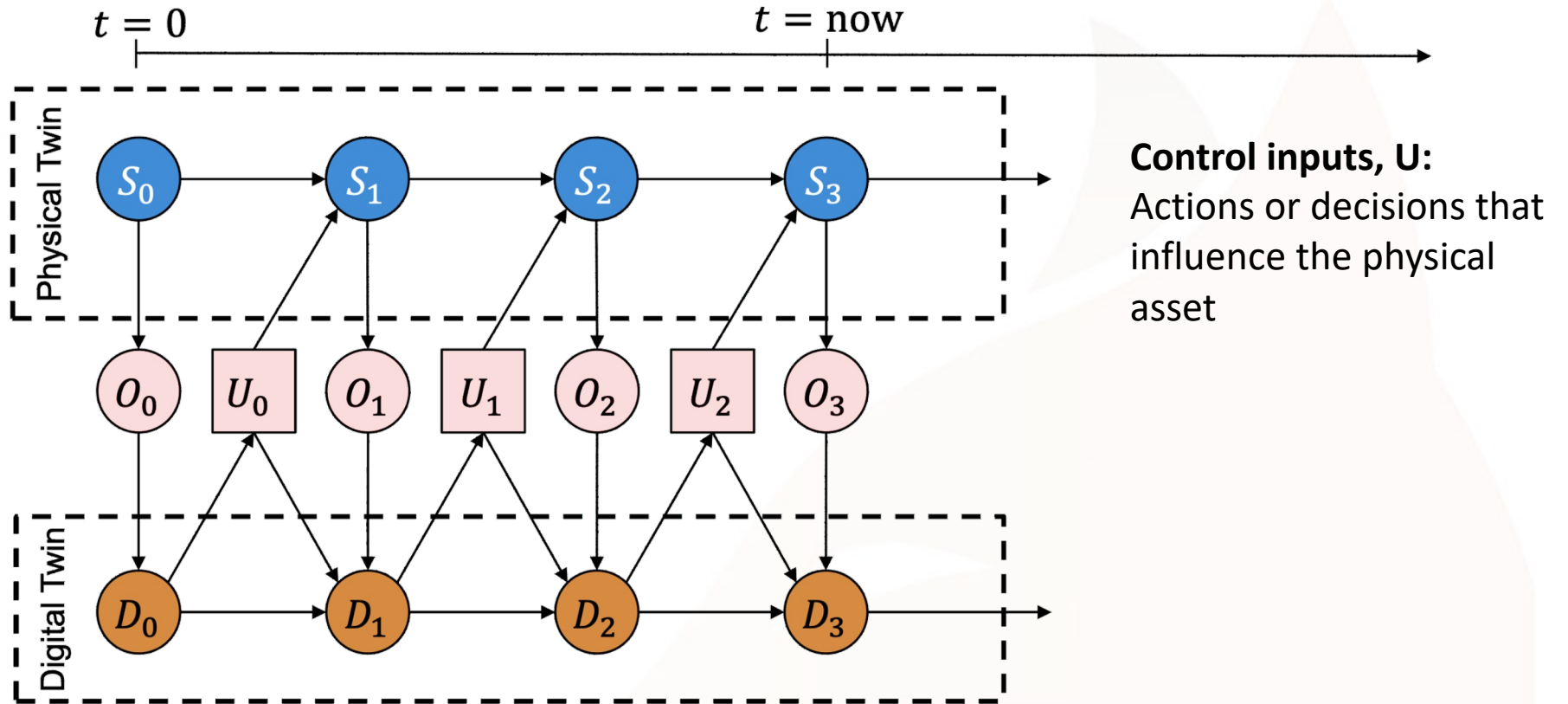
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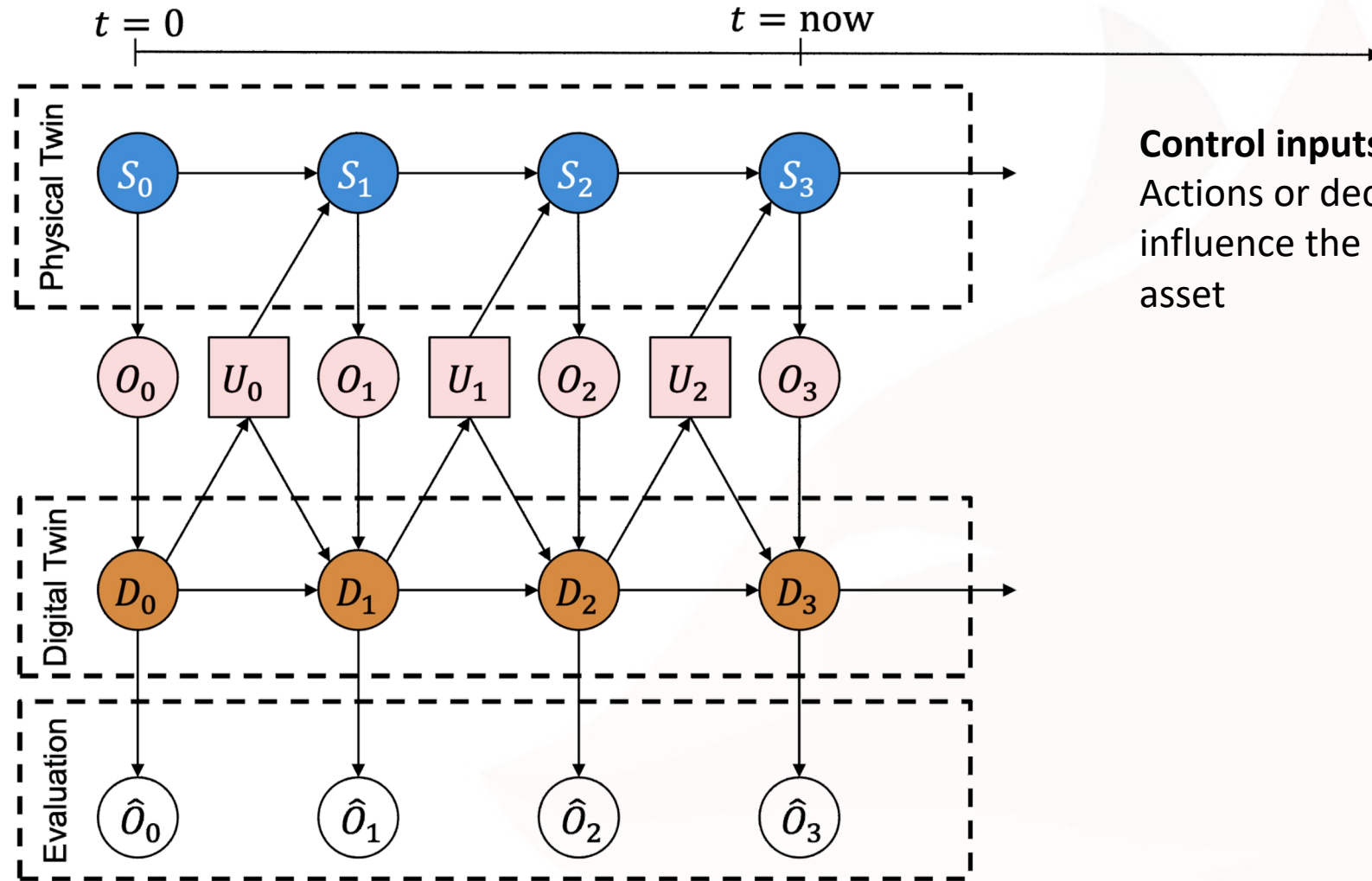
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**Control inputs, U:**  
Actions or decisions that influence the physical asset

## Physical State, $S$ :

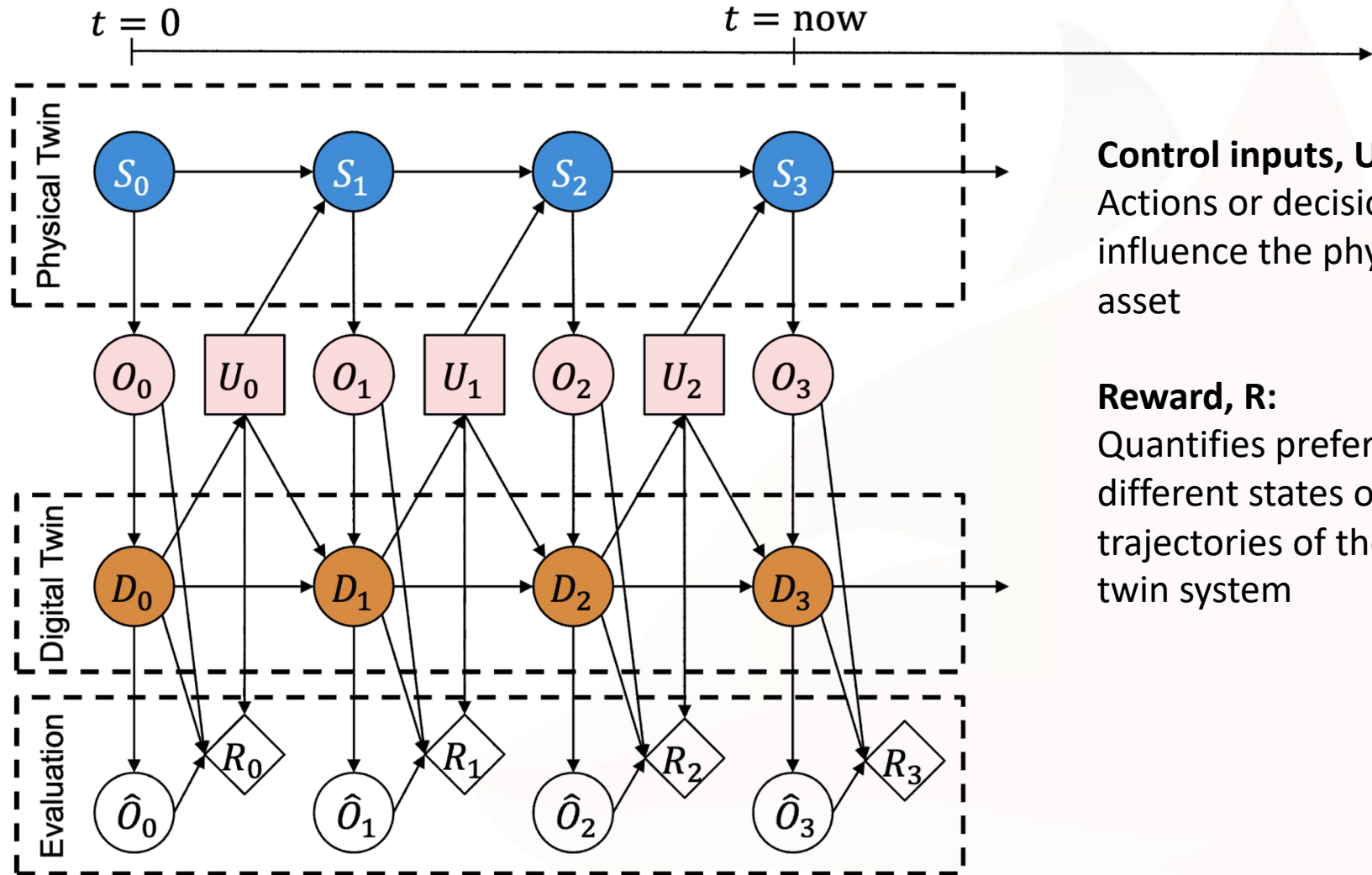
Parameterized state of the physical asset

## Digital State, $D$ :

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

## Observational data, $O$ :

Available information describing the state of the physical asset

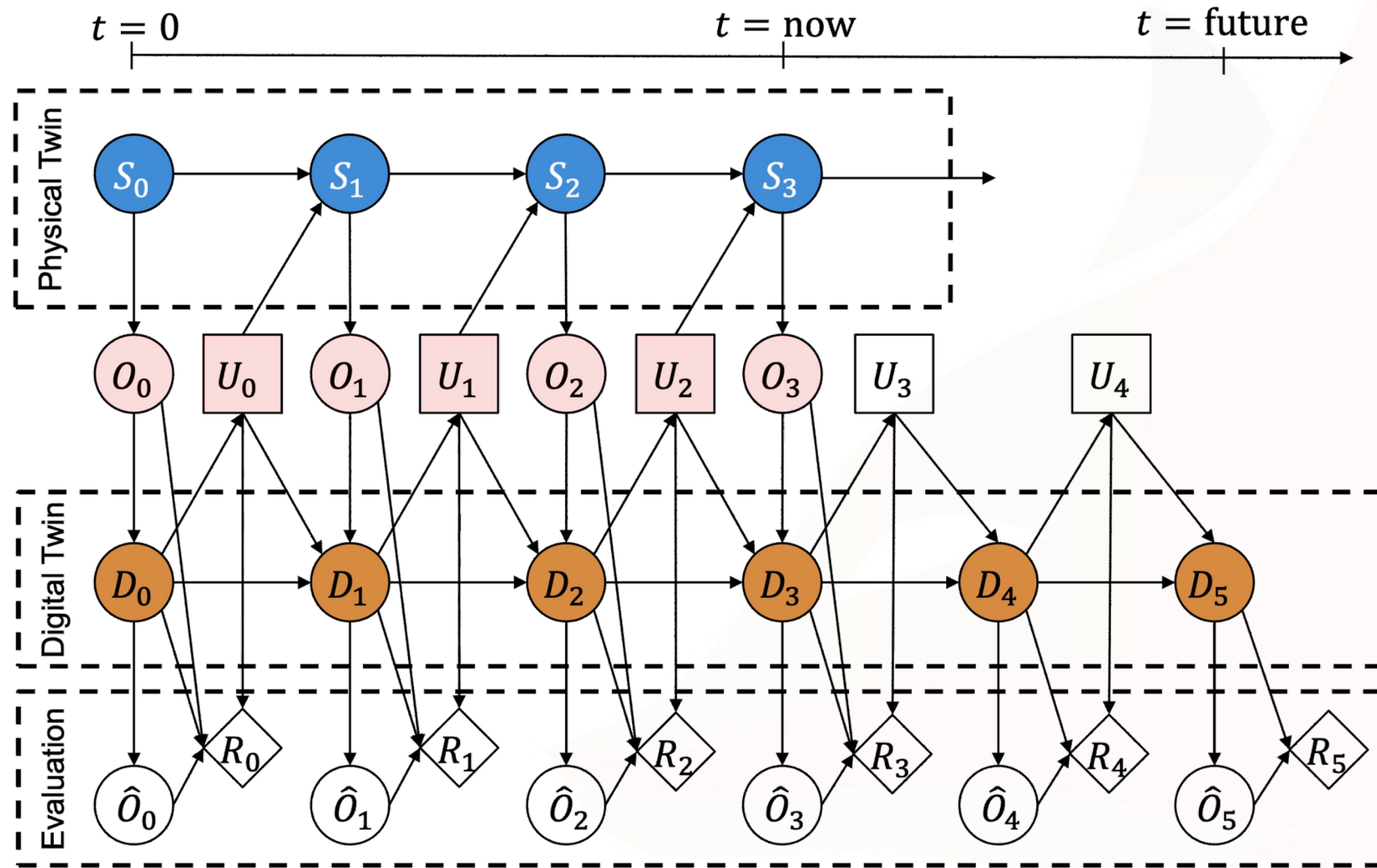


## Control inputs, $U$ :

Actions or decisions that influence the physical asset

## Reward, $R$ :

Quantifies preference of different states or trajectories of the asset-twin system



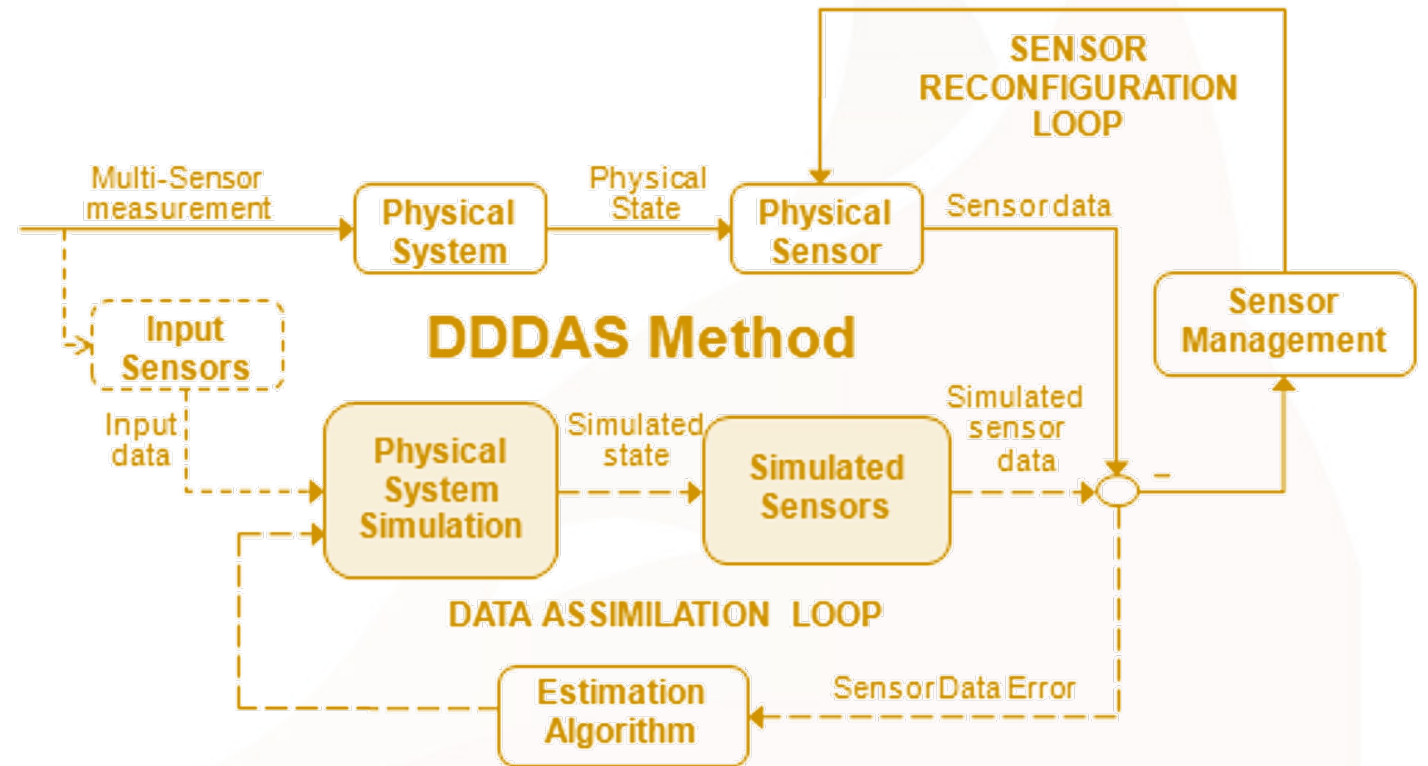
- Based on Dynamic Data-Driven Application Systems (DDDAS) paradigm.

- 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

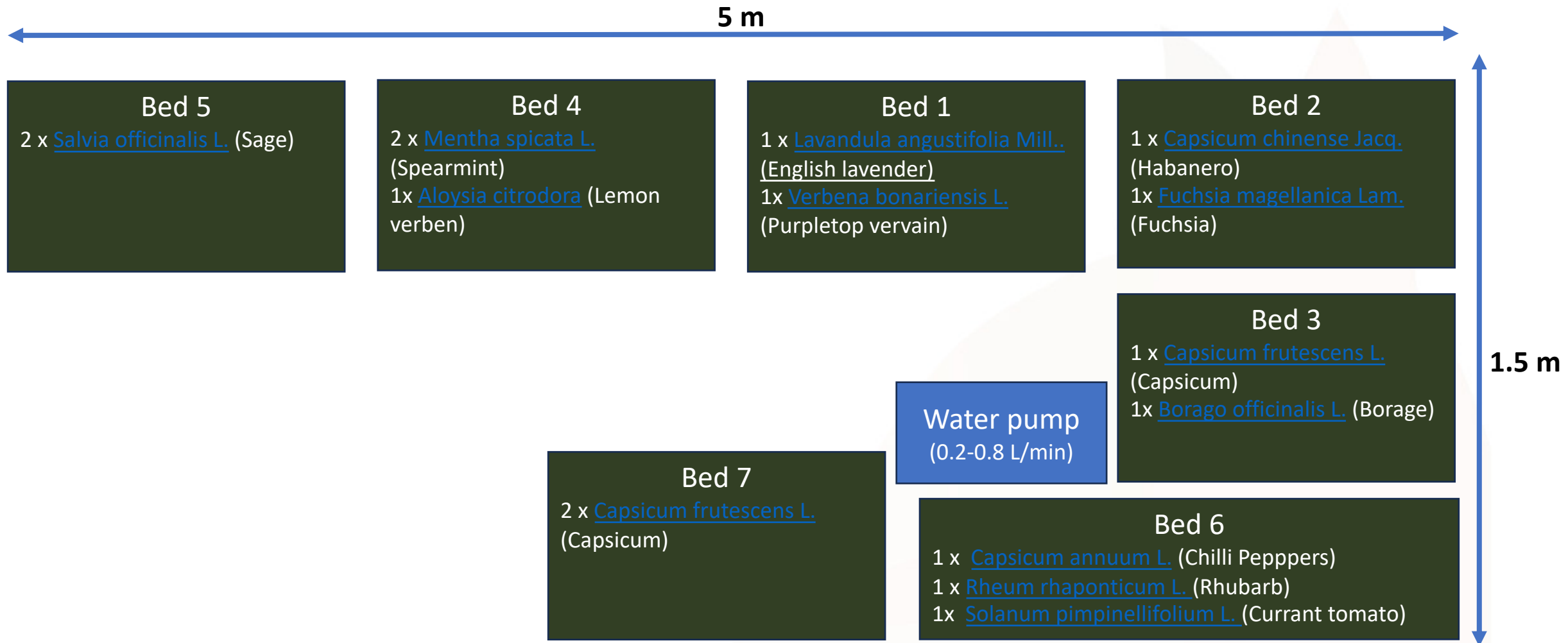
- **vSensors** - “Listen” for changes in data sources.
- **Intakers** - Pull new data.
- **Processors** - Process the data.
- **Assimilators** - assimilate the data into existing datasets.
- **Actuators** - change control inputs.
- **Loggers** - Assign FAIR metadata.

## State Management

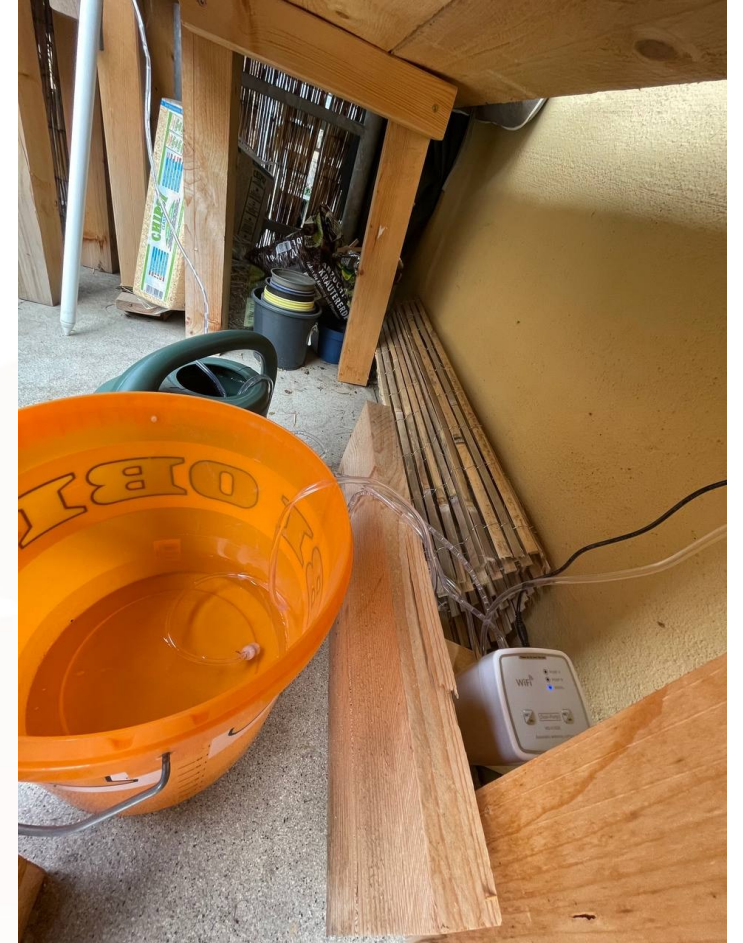
- **State Space** – defined states of the system.
- **Trackers** - Track state of the data.
- **Synchronizers** - State synchronization.
- **Sniffers** - Detect changes in state of the DT.

- 🔥 **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.

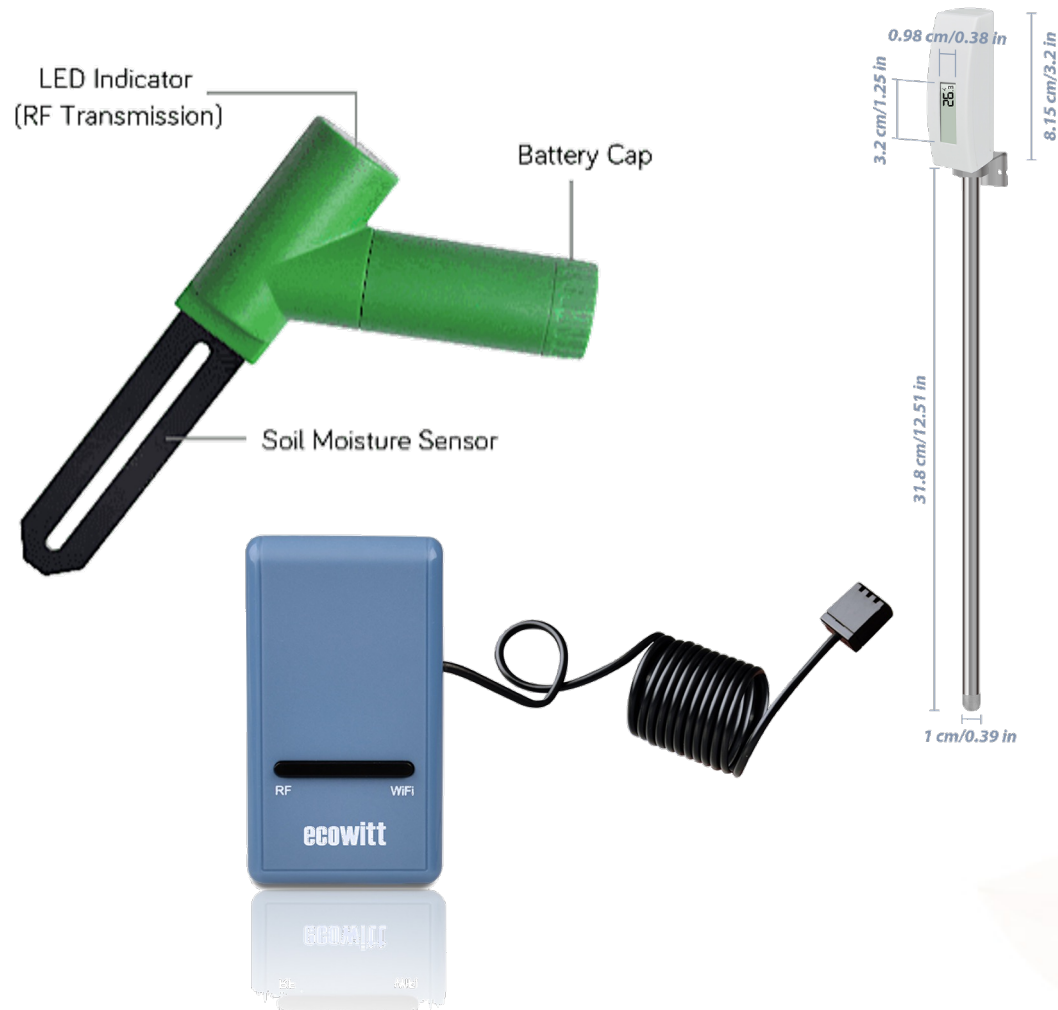
*HINT: Think about what other data sources can be added, what is the state space, what type of model, is needed and what components would be required.*



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







Soil moisture sensor, soil temperature sensor, base station. *Source: ecowitt.com*



WiFi enabled dual motor water pump. *Source: Cikonief.*



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