



## Introduction to Digital Twins Concepts

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CERTH SmartWins Summer School: Day 1

04 July 2023

Thessaloniki

# DT definition

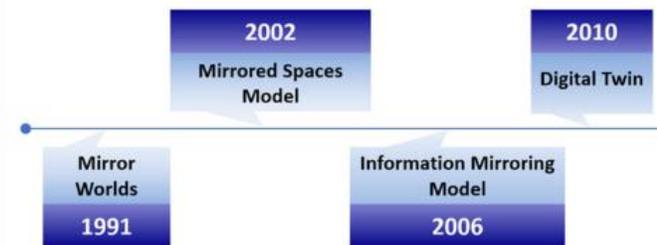
DT concept originated from the aerospace industry → NASA's roadmap(2010) on modeling & simulation → provided 1<sup>st</sup> definition for DTs:

**“An integrated multi-physics, multi-scale, probabilistic simulation of a vehicle or system that uses the best available physical models, sensor updates, fleet history, etc., to mirror the life of its flying twin”**

Since this NASA definition, new definitions in the literature have refer to DT as :

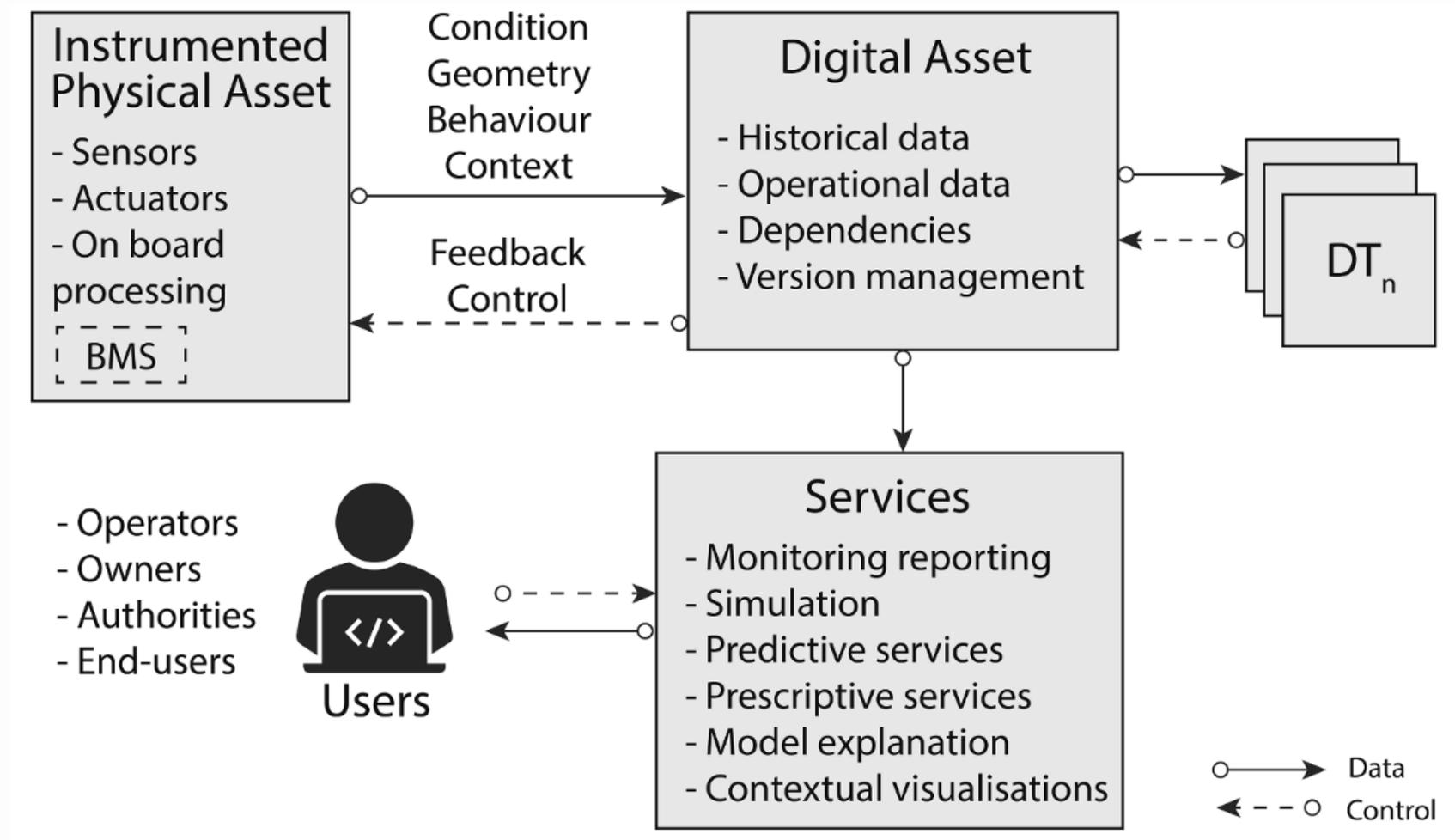
**“Virtual or digital model layout”, “counterpart”, “doppelganger”, “clone”, “footprint”, “software analogue”, “representation”, “information constructs”, or “simulation of its physical counterpart”**

A DT is a dynamic digital representation of an asset/ system and mimics its state or behavior in real-time and vice versa (GE Digital 2017; Bolton et al. 2018, Singh et al. 2021).



Evolution timeline of DT, Singh et al. 2021

# DT conceptual ecosystem



Source: Davila Delgado, J. M., & Oyedele, L. (2021). Digital Twins for the built environment: learning from conceptual and process models in manufacturing.

# DT main purpose

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**Bridge the gap between the real-world & Digital Systems**

**By shifting to a real-time event-driven development approach → quickly and effective integration of all systems**

**More accurate and interactive model of the real world.**

# DT characteristics

## High-fidelity

An accurate replica of its physical counterpart in terms of appearance, content, and functionality → relying on realistic digital models for accuracy.

## Dynamic

Changes alongside the physical system by maintaining a continuous exchange of data between the physical & virtual worlds → accurately mirroring behavior of its physical twin.

## Self-evolving

Evolves in sync w/ its physical counterpart throughout its lifecycle, creating a closed feedback loop → adapting in real-time based on data collected from the physical twin.

## Identifiable

Each physical asset has its unique DT, allowing for global identification and tracking throughout the entire lifecycle, as the data and models associated with the DT evolve along w/ the physical asset.

## Multiscale & Multiphysical

Incorporates properties of physical twin at different scales, from macroscopic geometric details to microscopic characteristics, including physical properties such as structural dynamics, thermodynamics, and material properties.

## Multidisciplinary

Integrates various disciplines, including computer science, engineering, automation, and physics, playing a central role in Industry 4.0.

## Hierarchical

Composed of interconnected sub-models representing different components of a system → hierarchical structure that reflects composition of the physical product.

# Advantages of DTs

**Speed  
prototyping**

Allows for faster design iterations and prototyping by simulating different scenarios and facilitating customization based on user needs and usage data → Enables real-time comparison between predicted and actual performance, leading to improved design decisions.

**Cost-effective**

Reduces prototyping costs by utilizing virtual resources, eliminating needs for physical materials and repeated testing → reduces operating costs, extends equipment life, and optimizes maintenance strategies

**Improved  
maintenance**

Anticipates defects and damage, enabling proactive maintenance scheduling. It provides optimized solutions and continuously improves system processes through the feedback loop between the DT and its physical twin.

**Waste reduction**

Minimizes material wastage by simulating and testing product prototypes virtually → optimized designs & reduced development costs.

**Accessibility**

Allows remote control/monitoring of physical devices, making them widely accessible → crucial in situations w/ limited local access

**Documentation  
/communication**

Centralizes scattered data, simplifying access and maintenance. It enables better understanding, documentation, and communication of the behavior and mechanisms of the physical twin.

**Self-evolving**

Evolves in sync with its physical counterpart throughout its lifecycle, reflecting any changes in either entity → matures along w/ the physical twin.

# DT vs. Simulations

Major difference between a DT & simulation → data interconnection that allows exchanging information between the physical object & virtual object

Simulation

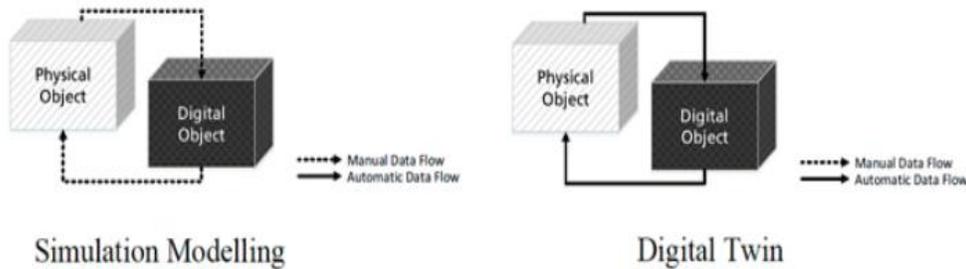
A simulation predicts future states of a physical system based on a set of initial assumptions

Limited capability in evaluating system performance & considering the physical part of the system

DT

A virtual environment for several simulations, with real-time data and a two-way flow of information between physical & virtual object

Increases the accuracy of predictive analytical models. Offers greater understanding for the management and monitoring of products, policies & procedures  
**However → increased costs**



Simulation Modelling vs. Digital Twin (Krasikov, I., & Kulemin, A. N. (2020).

# Classification of DTs

## Hierarchal perspective

**1. Unit level:**  
Based on geometric, functional, behavioural, & operational model of unit-level physical twin.

**2. System level:**  
Combines unit-level DTs in a production system, facilitating data flow and resource allocation. (e.g. products like aircraft or buildings are system-level DTs.)

**3. System of Systems (SoS) level:**  
Multiple system-level DTs are interconnected to form SoS-level DT → collaboration between systems or departments.

## Level of Integreation

**1. Digital Model:**  
Data between the physical & digital object are exchanged manually → any changes in physical object state not reflected in the digital one, & vice versa

**2. Digital Shadow:**  
Data from physical object flow to the digital automatically, but still manually the other way around → any change in the physical object seen in digital copy, but not vice versa

**3. Digital Twin:**  
Automatic bidirectional flow of data between the physical & digital object → changes in either object, physical or digital, directly lead to changes in the other.

## Level of Maturity

**1. Partial DT:**  
Contains a small number of data points, e.g., temperature, humidity, etc., which is useful in determining the connectivity & functionality of DT.

**2. Clone DT:**  
Contains all significant and relevant data from the product/system → prototypes & categorizing development phases

**3. Augmented DT:**  
Utilizes data from the asset along with its historical data and at the same time derives and correlates the useful data using algorithms and analytics.

# IoT's critical role (1/2)

## IoT definition

**“Interconnection of sensing devices providing the ability to share information across platforms through a unified framework, developing a common operating picture for enabling innovative applications”**

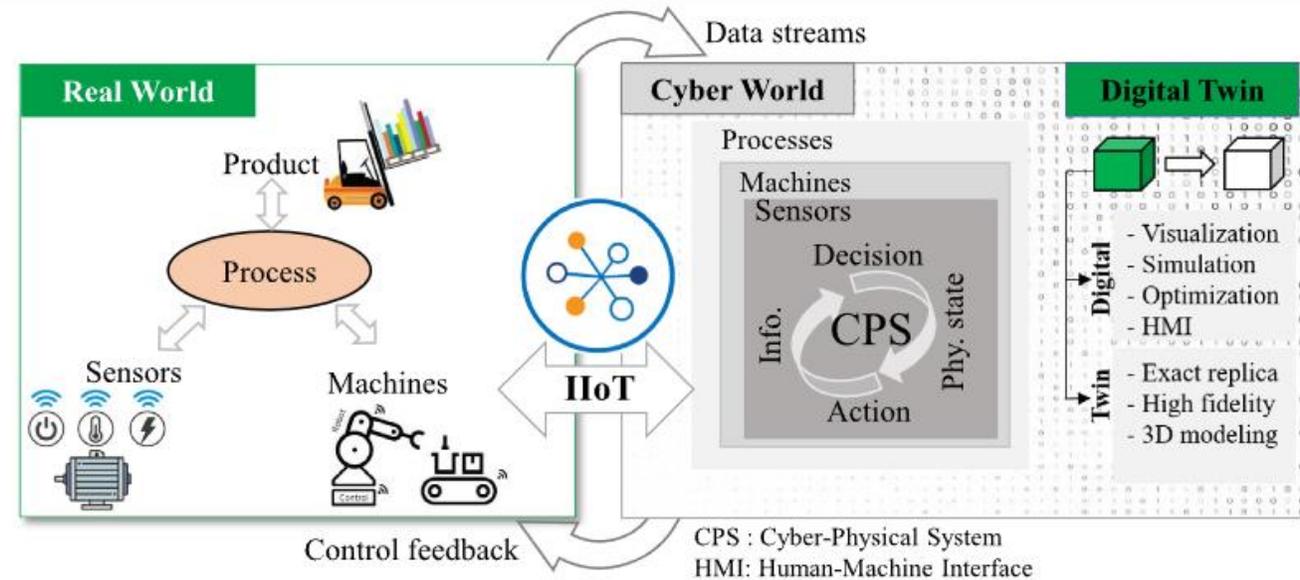
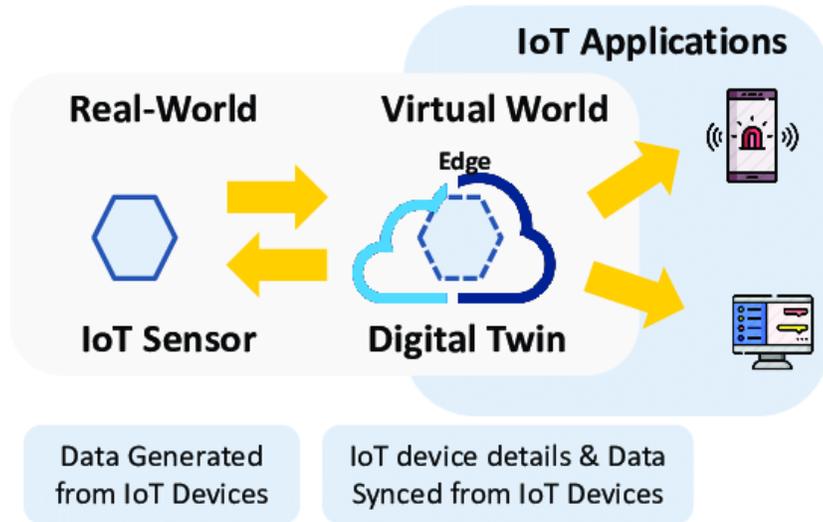
## Data collection

**Data collected from IoT devices such as smart meters, wireless sensor networks, as well as weather stations → utilized for the development of parameters required for DT implementation**

## DT-IoT interaction

**The explosion of IoT sensors → essential for enabling live data feed  
DT depends on use of effective tools for information communication, IoT-enabled devices, and sensor-based data-capturing devices, especially for asset monitoring  
As IoT devices are refined → DT scenarios can include smaller & less complex objects**

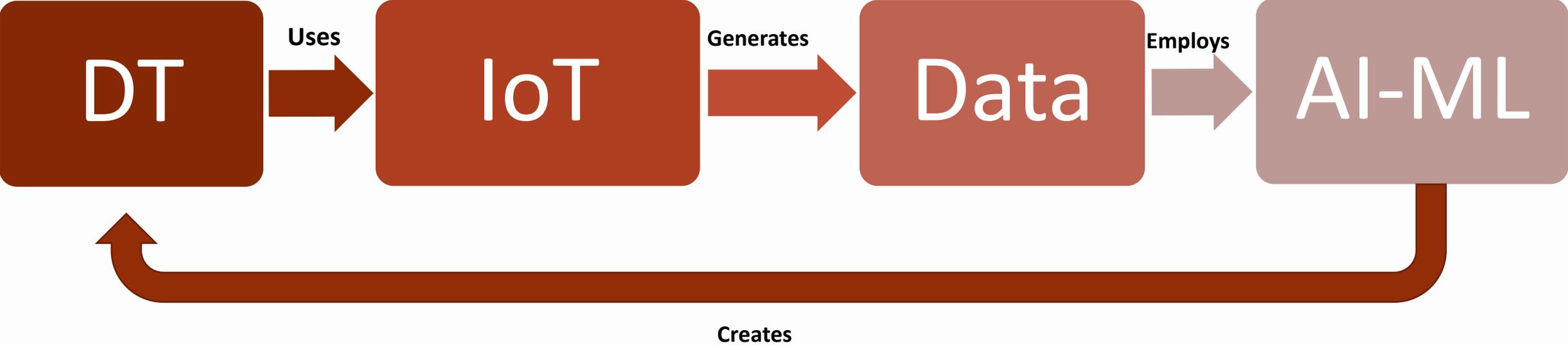
# IoT's critical role (2/2)



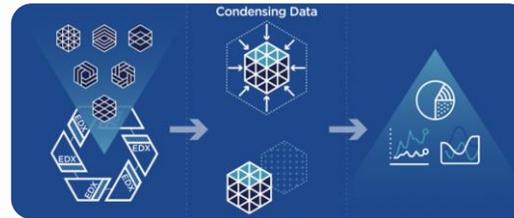
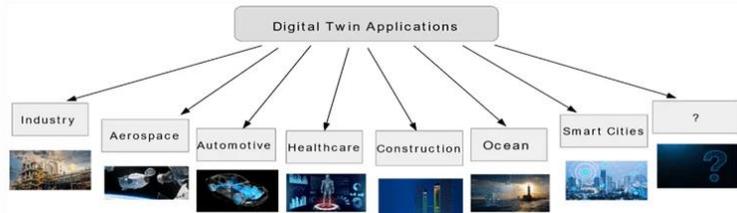
Source: Shah Zeb, Aamir Mahmood, Syed Ali Hassan, MD. Jalil Piran, Mikael Gidlund, Mohsen Guizani, Industrial digital twins at the nexus of NextG wireless networks and computational intelligence: A survey, Journal of Network and Computer Applications, Volume 200, 2022,

# Relationship between IoT, big data, AI-ML, and DT

Several technologies come together to make DT a reality such as 3D simulations, IoT, AI, Big Data, ML, & cloud computing.



# DT concepts in the built environment



## DT applications

DT found its application in industries beyond manufacturing and now supports many different applications

- Manufacturing
- Security
- Healthcare
- Automotive

## Data at the core of DT

Based on available data, different intelligent functions (e.g., AI, ML module, simulation) could be realized for advanced decision-support

### Examples:

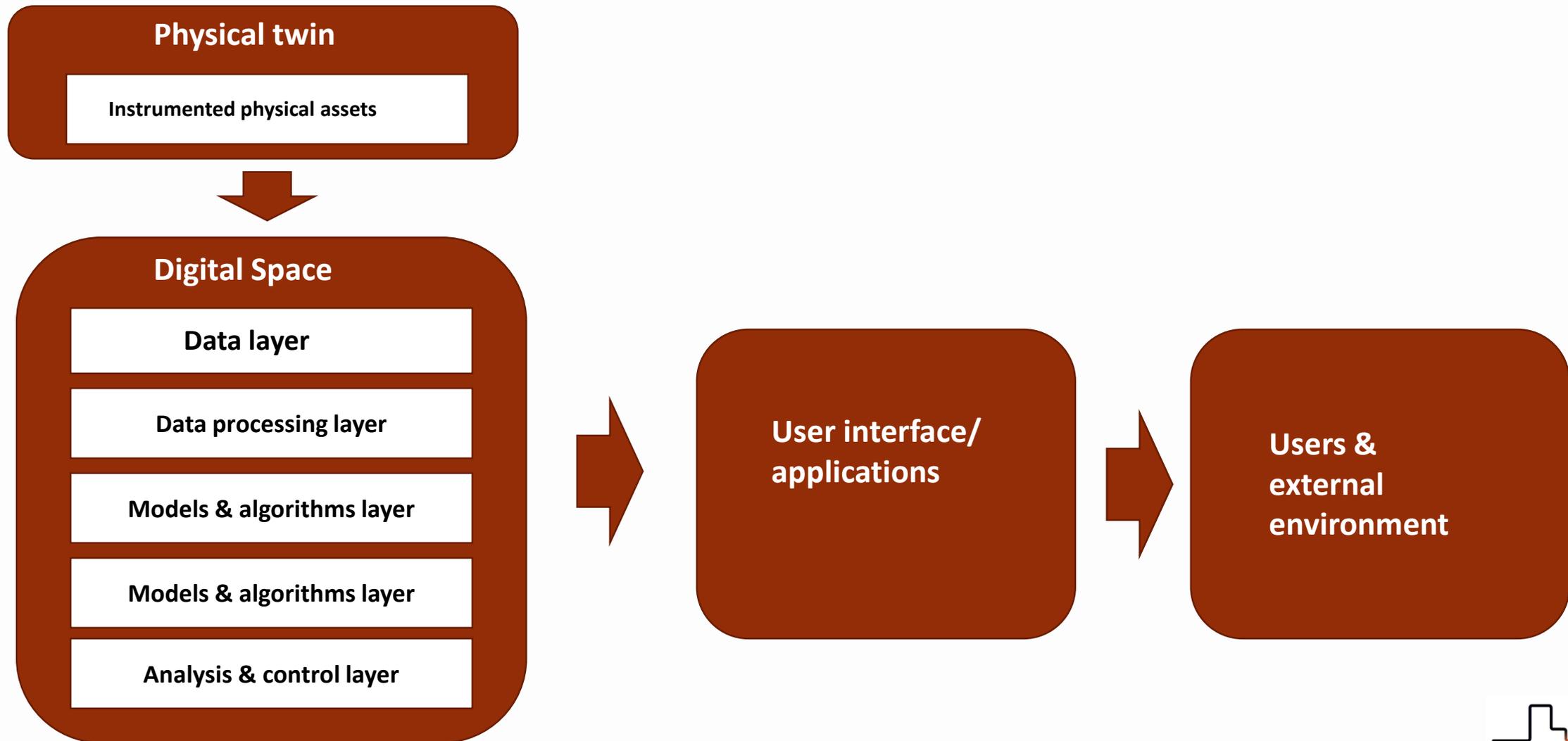
- Transportation prediction,
- Energy usage optimization
- Asset anomaly detection

## DT related with the built environment

More precisely : Architecture, Engineering, Construction, and Facility management (AEC/FM) sectors

DTs are particularly examined in the context of smarter cities/buildings

# Conceptual system architecture for a DT



# DT architecture principles

To construct an effective DT architecture in the building level, the following strategies should be examined

Architecture to be developed using a unified, hierarchical, & extensible approach  
→ implemented in different scales from assets (e.g., pump) to a building

Besides data collection & acquisition, assets need to be connected and relevant information regarding their life cycle (e.g., maintenance history) should be collected as well.

Interaction & communication channels w/ humans are needed to provide in-time services.

Data visualization → essential to monitor current condition & activities in DTs. However, integrating data from different systems → challenging due to varying data requirements and intended uses across applications.

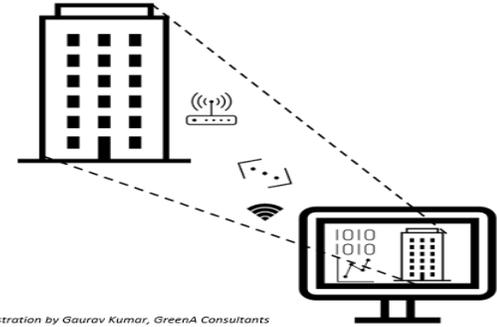
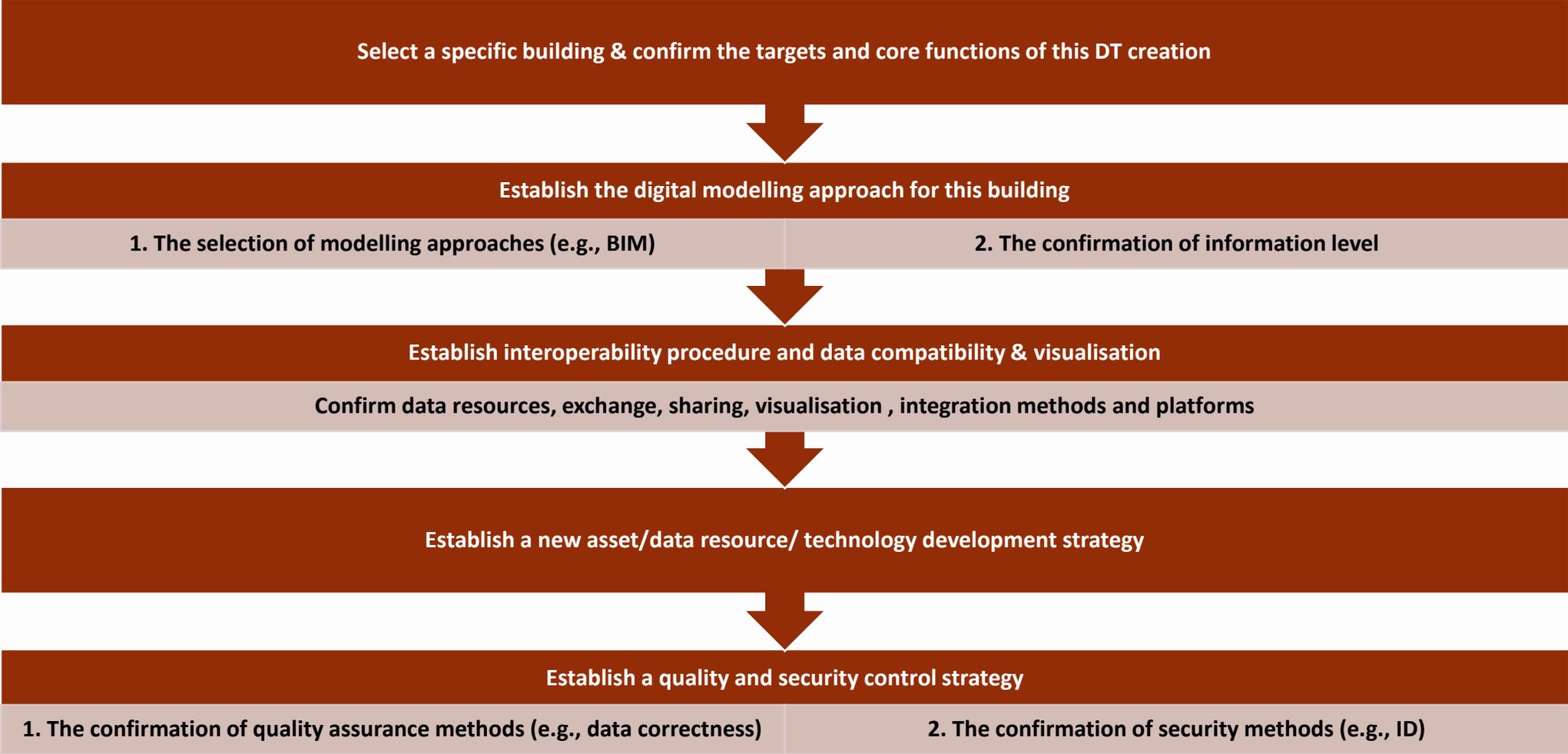


Illustration by Gaurav Kumar, GreenA Consultants



**Adapted from:** Lu, Qiuchen & Parlikad, Ajith Kumar & Woodall, Philip & Xie, Xiang & Liang, Zhenglin & Konstantinou, Eirini & Heaton, James & Schooling, Jennifer. (2019). Developing a dynamic digital twin at building and city levels: A case study of the West Cambridge campus. Journal of Management in Engineering.

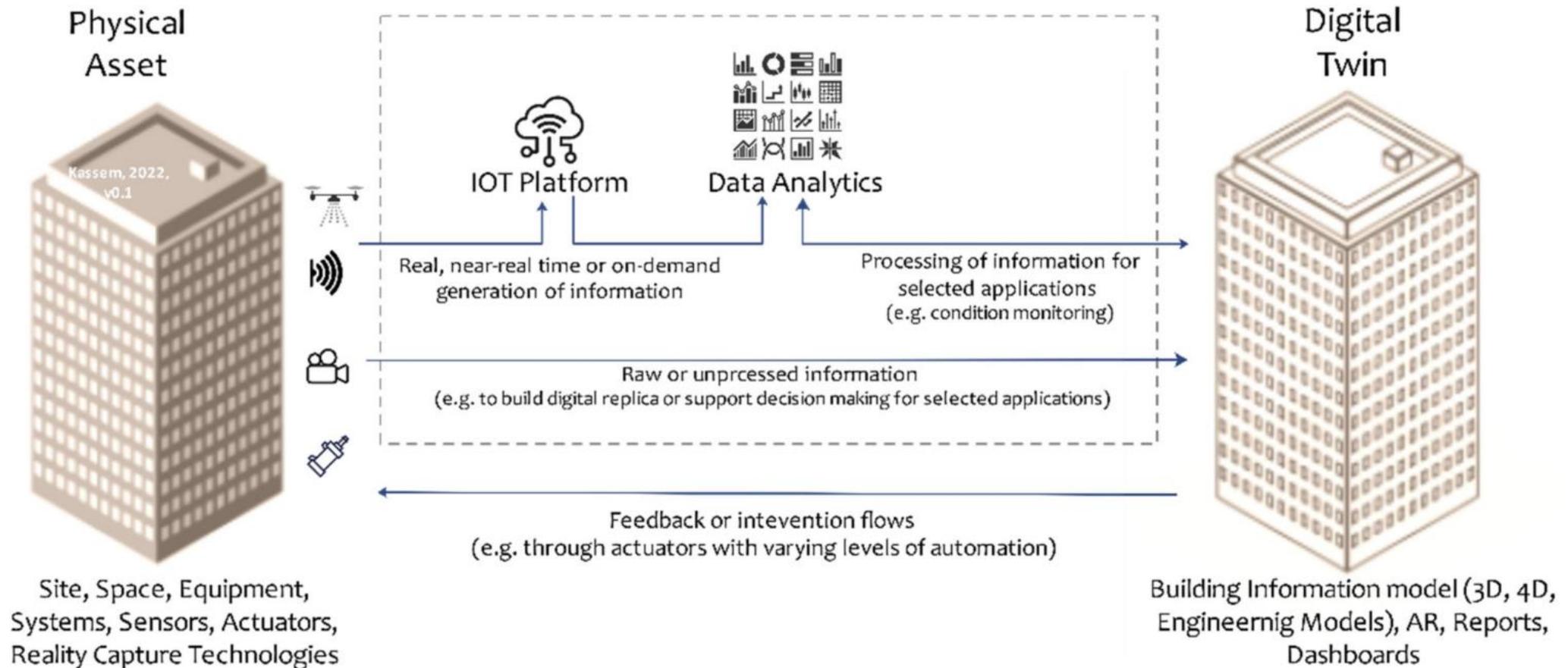
# Flow chart for DT development at building level



Adapted from: Lu, Qiuchen & Parlikad, Ajith Kumar & Woodall, Philip & Xie, Xiang & Liang, Zhenglin & Konstantinou, Eirini & Heaton, James & Schooling, Jennifer. (2019). Developing a dynamic digital twin at building and city levels: A case study of the West Cambridge campus. Journal of Management in Engineering.



# Main components to create a DT of a building



**Source:** Shahzad, M., Shafiq, M. T., Douglas, D., & Kassem, M. (2022). Digital Twins in Built Environments: An Investigation of the Characteristics, Applications, and Challenges. *Buildings*, 12(2)

# Key challenges of DT implementation

**Interoperability & lack of common standards :**  
Buildings operation data need to be collected/ shared from different project stages,actors (e.g.,designers, manufacturers, constructors)

W/o a standard for data records among data sources → difficult to know whether a data record in one system (e.g., a particular device) is the same entity referred to in another data record in another system

**A key challenge in adopting digital twins in the built environment sector is the lack of “an all-inclusive toolset”**

Cost-effectiveness through successful case studies is essential for the uptake of new technologies

**Data quality can be lost when extracting data from source systems**

Even if the data extraction process is perfect, if the data from the source systems contains errors → propagate to the DT

**A key problem in a DT is timing & frequency of synchronizing different copies of data in order to provide up-to-date data to decision makers**

DTs with a requirement to monitor engineering assets in real time, a continuous stream of data will be needed, which shifts the trade-off toward high synchronization costs

**Overall, a cultural change in the built environment sector is essential for fostering digitalization, including the adoption of DT**

Project Partners

