

# Engineering complex and sustainable systems through digital twins

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# Dr. Istvan David



MSc in Business Inf. Systems (2014)  
MSc in Computer Engineering (2013)  
BSc in Computer Engineering (2011)  
Budapest Tech (BME), Hungary

modeling and simulation  
engineering processes  
PhD in Computer Science (2014–19)  
University of Antwerp, Belgium

 IVADO Postdoctoral Laureate 2021

digital twins for C(B)PS  
reinforcement learning  
Postdoctoral researcher (2021–2023)  
Université de Montréal, Canada

digital twins and sustainability  
energy-aware systems  
Assistant Professor (2023–)  
McMaster University, Canada



R&D Software Engineer (2012–2014)  
Mentor Graphics, Hungary

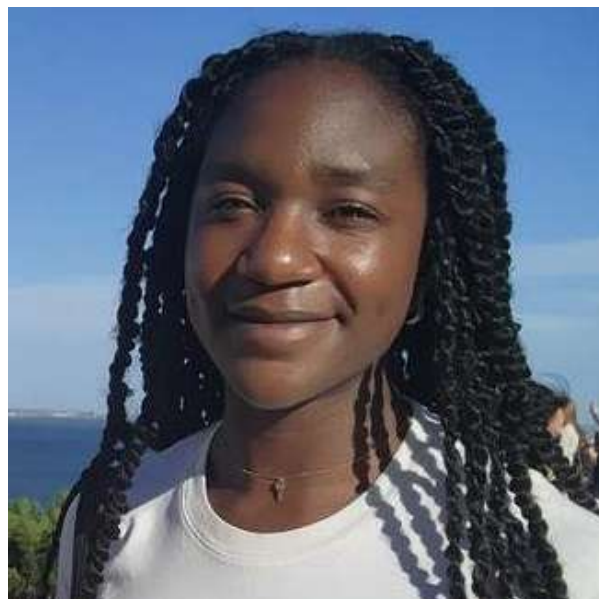


Head of Digital Innovation (2019–2020)  
Trigo Group, Hungary/France



# Sustainable Systems and Methods

<https://istvandavid.com/team>



 *Sustainable systems **by** sustainable methods*

Sustainability

Sustainable systems engineering  
Energy-efficient simulators  
Green computing      Environmental sustainability  
Human-in-the-loop      System evolution

Digitalization

Digital twins      Digital thread  
Machine learning / AI  
Tool chains and process tools

Modeling &  
Simulation

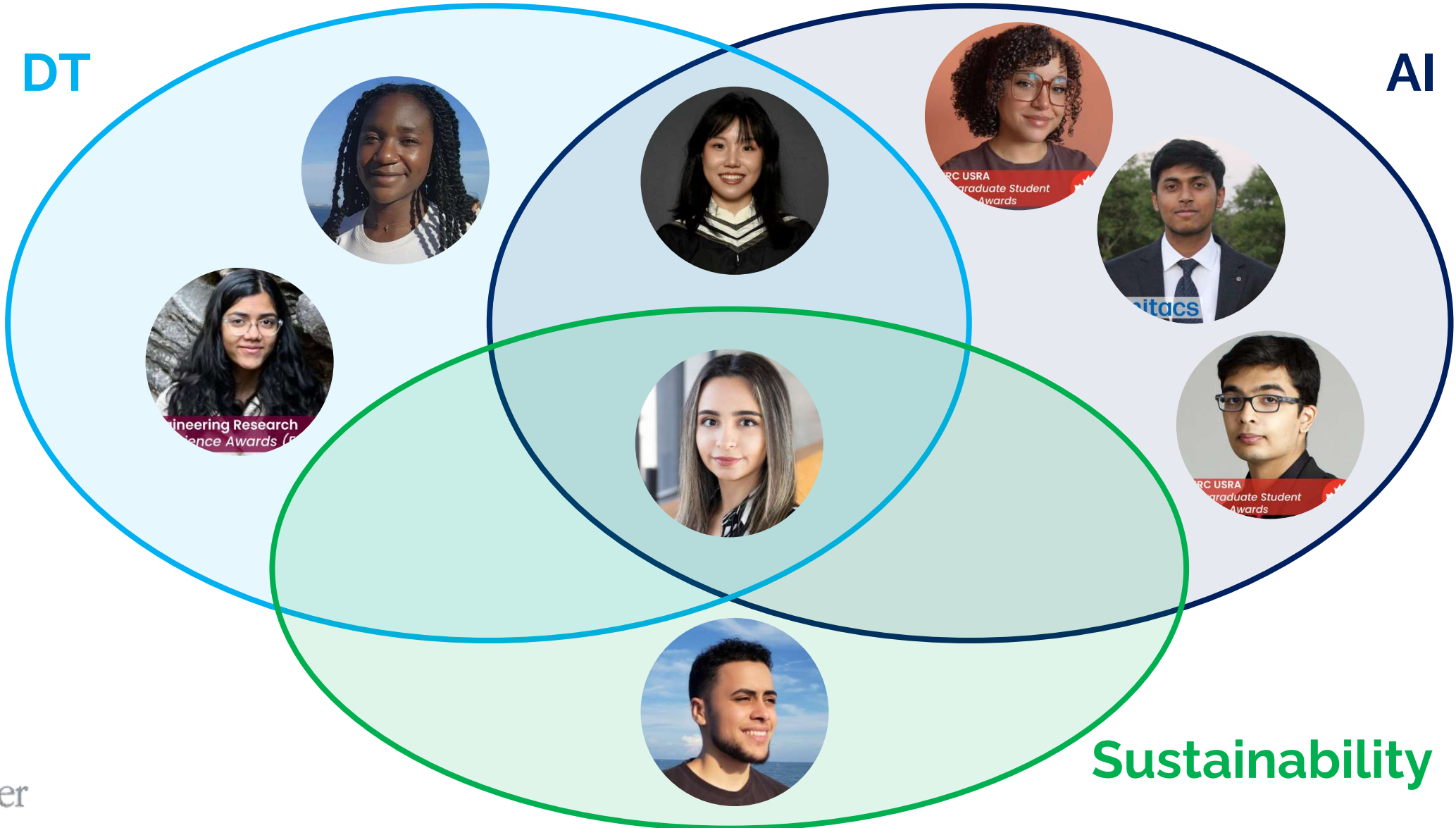
Model-driven engineering (MDE)  
Model-based systems engineering (MBSE)  
Multi-paradigm modeling (MPM)  
Co-simulation      Discrete event simulations



# Sustainable Systems and Methods

DT

AI



Sustainability





**McMaster**  
University

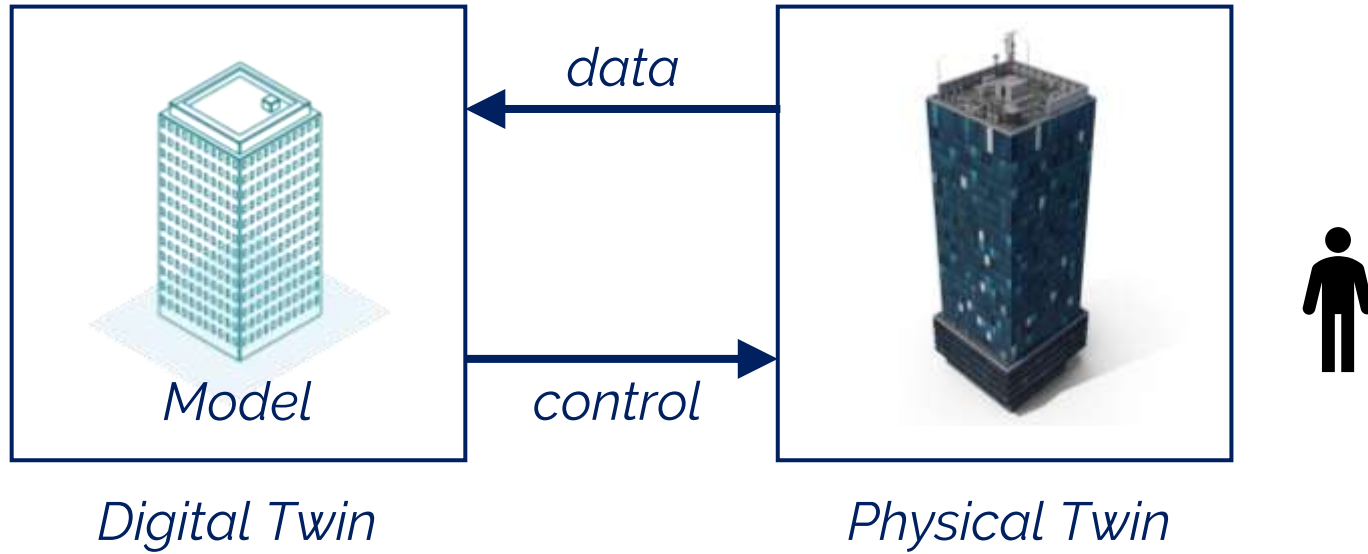




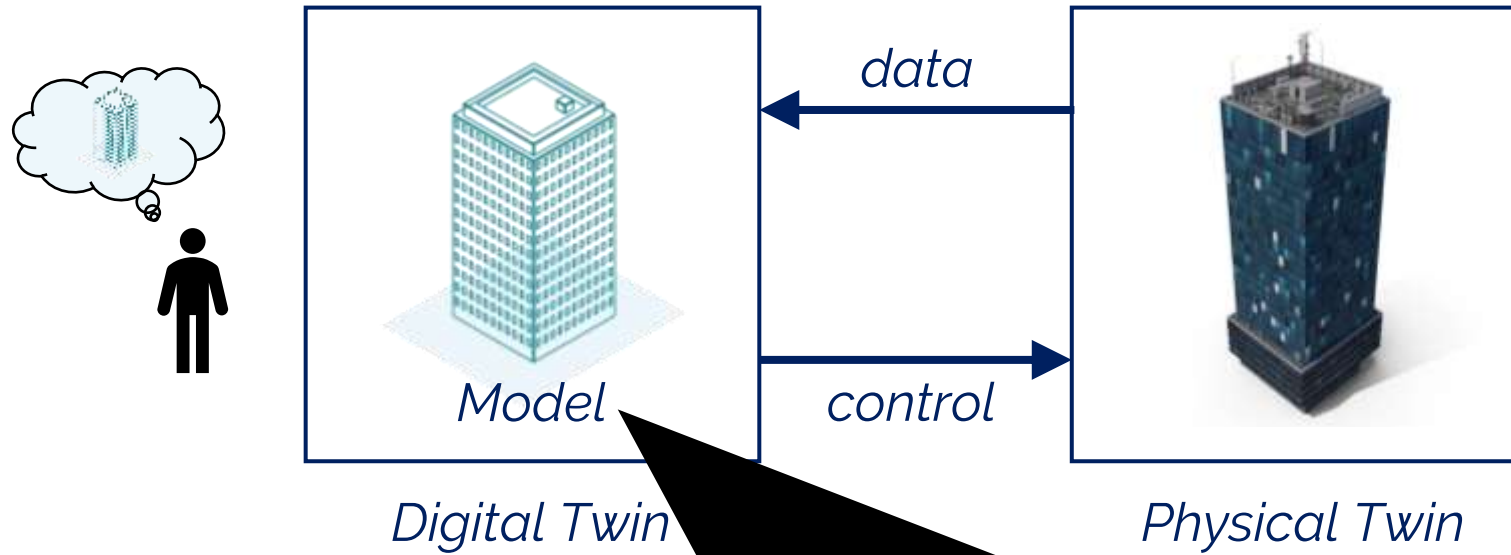
Ye olde digital twin



# A fit-for-purpose definition of “digital twin”



# A fit-for-purpose definition of “digital twin”



This block contains several hand-drawn diagrams and equations on a black background, illustrating various engineering and modeling concepts:

- Top Left:** A mechanical system diagram showing a mass  $m$  on a spring-damper system with stiffness  $k$  and damping  $b$ . A force  $F$  is applied to the mass, and its displacement  $x$  is measured. A circled  $+C$  is also present.
- Top Center:** The heat conduction equation  $\frac{\partial T}{\partial t} = \alpha \nabla^2 T$  is written above a 3D surface plot showing a temperature distribution with a color gradient from blue (low) to red (high).
- Bottom Left:** A control loop diagram with two nodes labeled 1 and 2, connected by arrows in a clockwise cycle.
- Bottom Center:** A block diagram of a control system with a feedback loop.
- Bottom Right:** A block diagram showing an input  $S_e$  entering a block labeled  $1/s$ . The output of this block branches into three paths:  $C = 1/k$ ,  $R \cdot b$ , and  $I \cdot m$ .

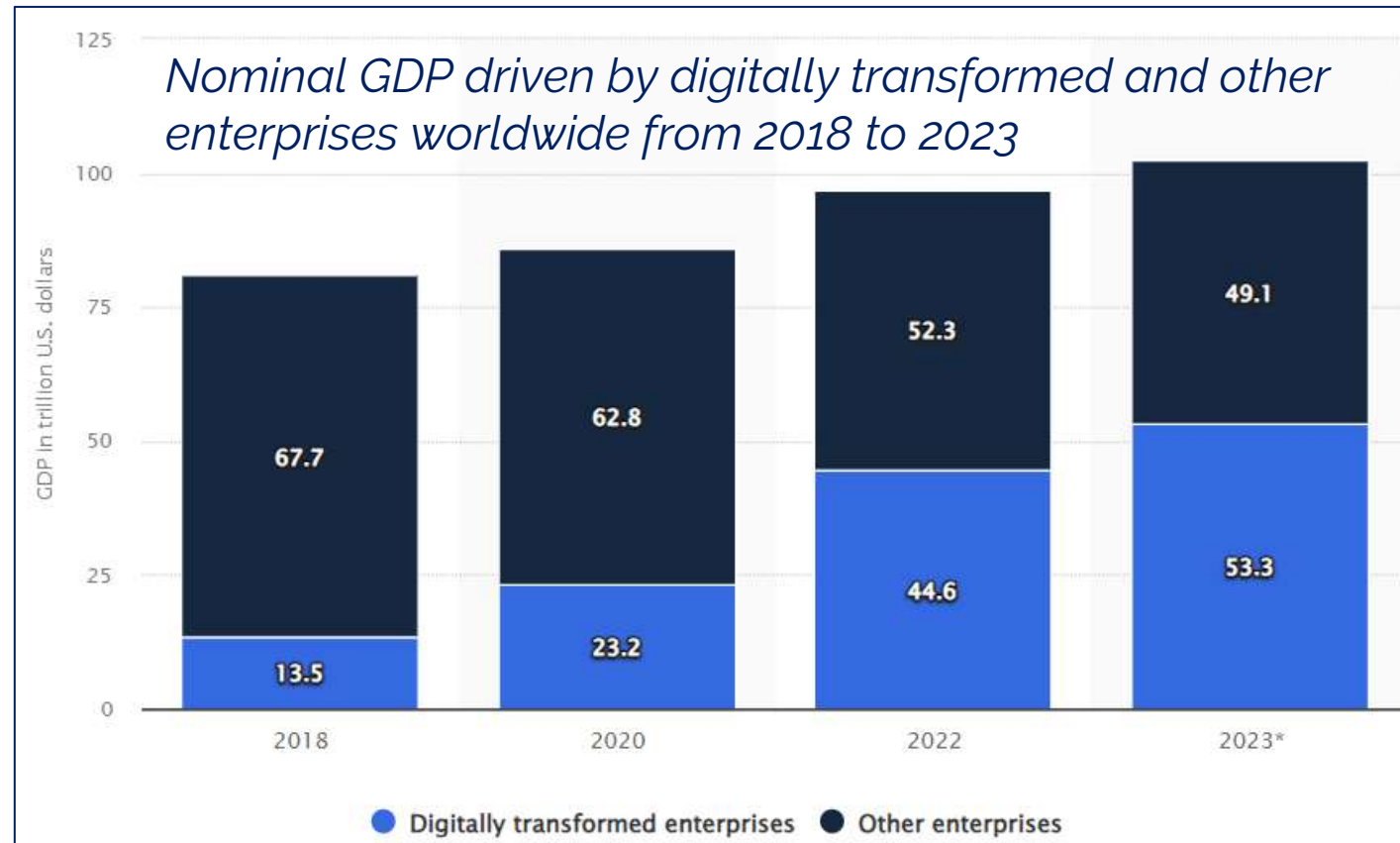
# Digitalization and digital transformation

## Industry 4.0 and 5.0

I5.0 complements the existing I4.0 approach by specifically putting research and innovation at the service of the transition to a **sustainable, human-centric** and **resilient European industry**



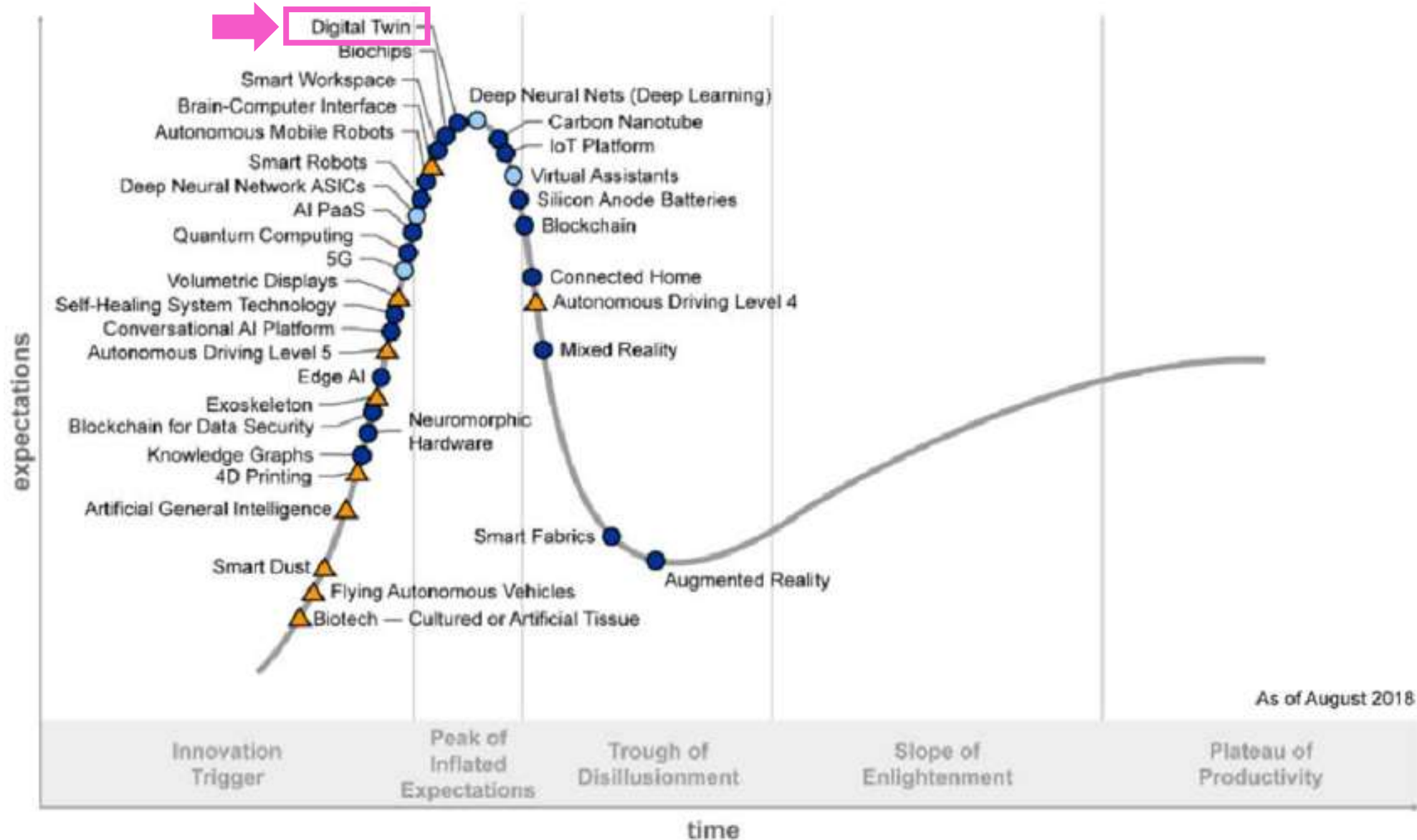
Digital Transformation Pyramid



<https://www.statista.com/statistics/1134766/nominal-gdp-driven-by-digitally-transformed-enterprises/>



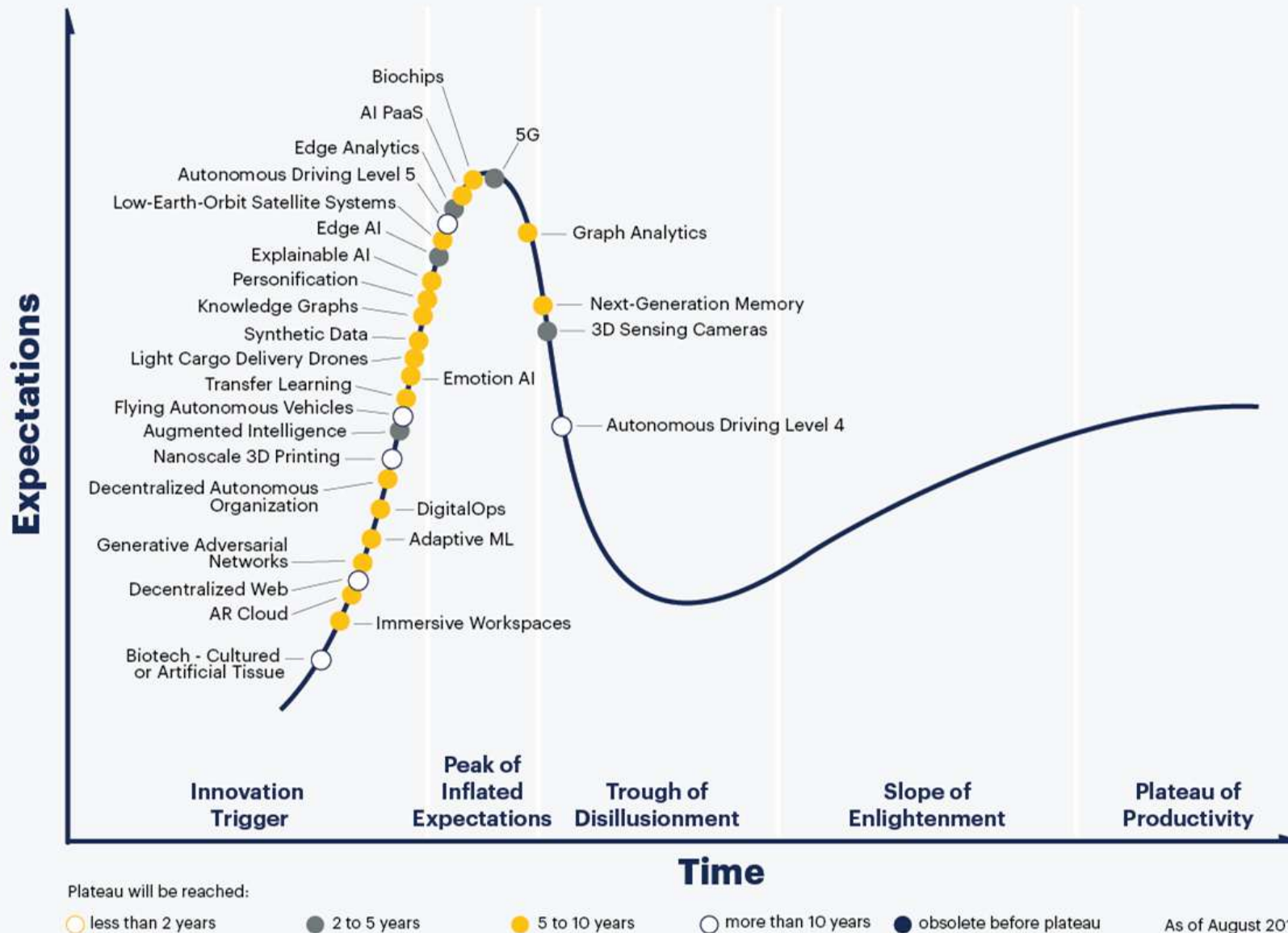
# 2018



Plateau will be reached:

- less than 2 years
- 2 to 5 years
- 5 to 10 years
- ▲ more than 10 years
- ⊗ obsolete before plateau

# Gartner Hype Cycle for Emerging Technologies, 2019



# Hype Cycle for Emerging Technologies, 2020



Plateau will be reached:

○ less than 2 years

● 2 to 5 years

● 5 to 10 years

▲ more than 10 years

⊗ obsolete before plateau

As of July 2020

# Hype Cycle for Emerging Technologies, 2021



Plateau will be reached:

○ less than 2 years

● 2 to 5 years

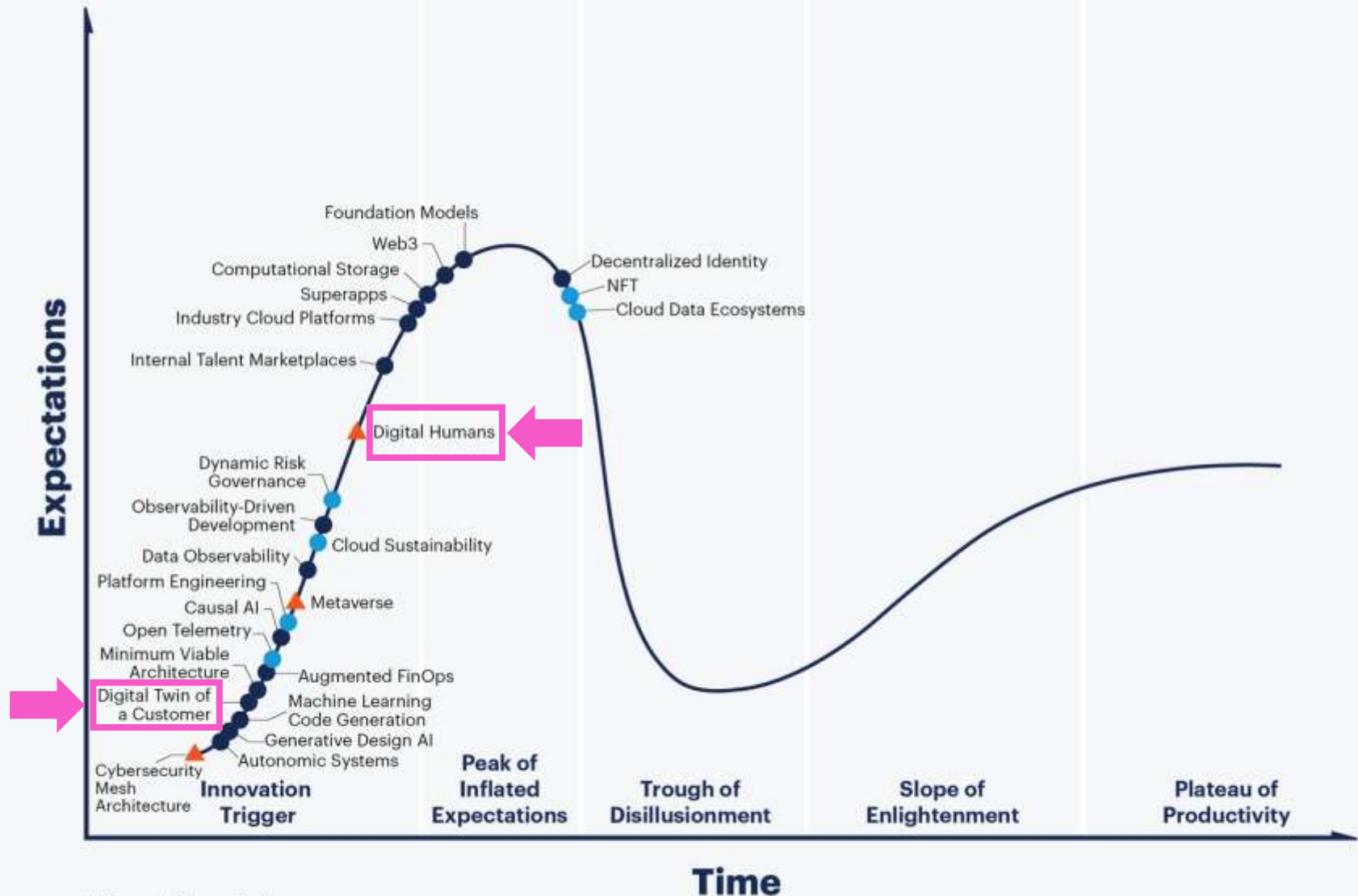
● 5 to 10 years

▲ more than 10 years

⊗ obsolete before plateau

As of August 2021

# Hype Cycle for Emerging Tech, 2022



Plateau will be reached:

- less than 2 years
- 2 to 5 years
- 5 to 10 years
- ▲ More than 10 years
- ⊗ Obsolete before plateau

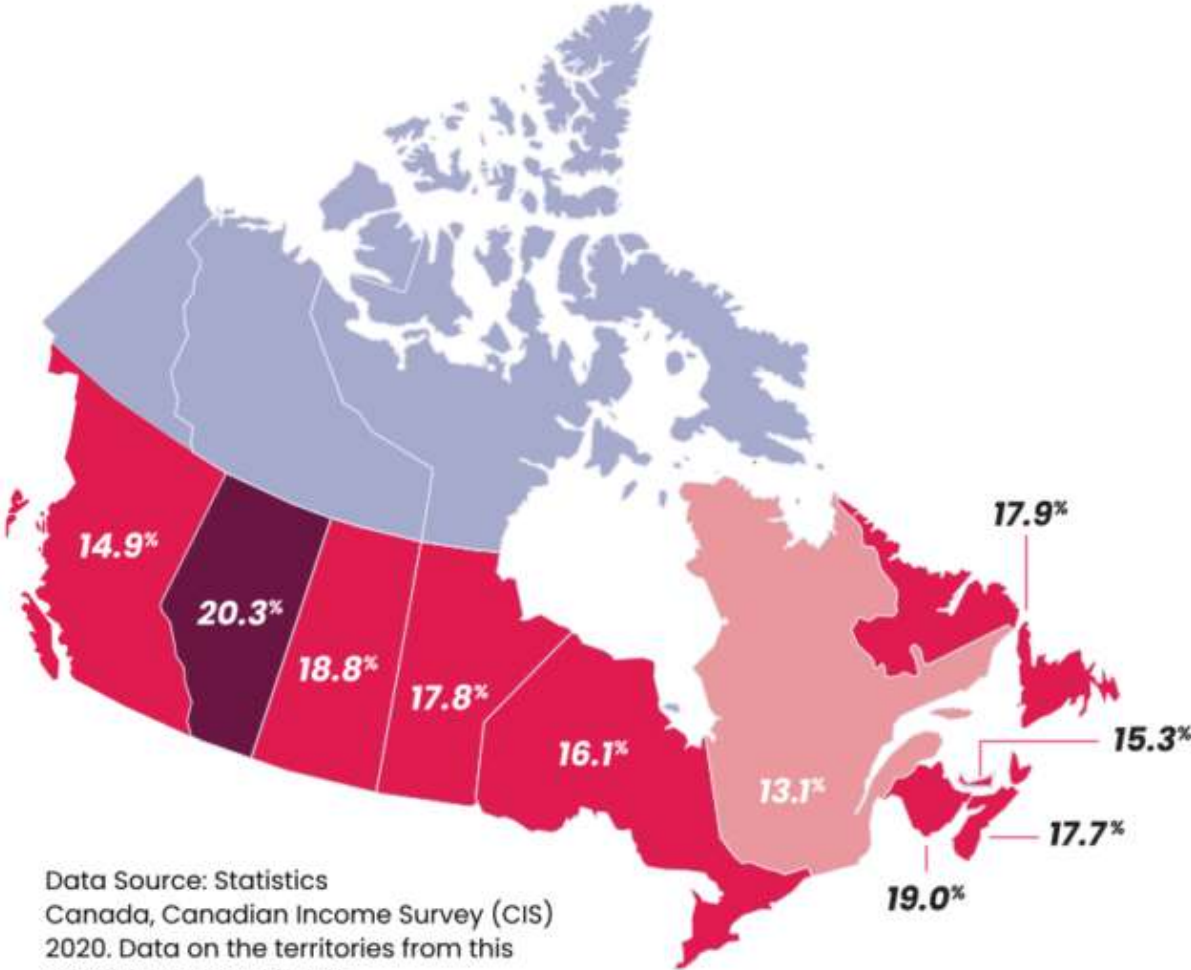
As of August 2022





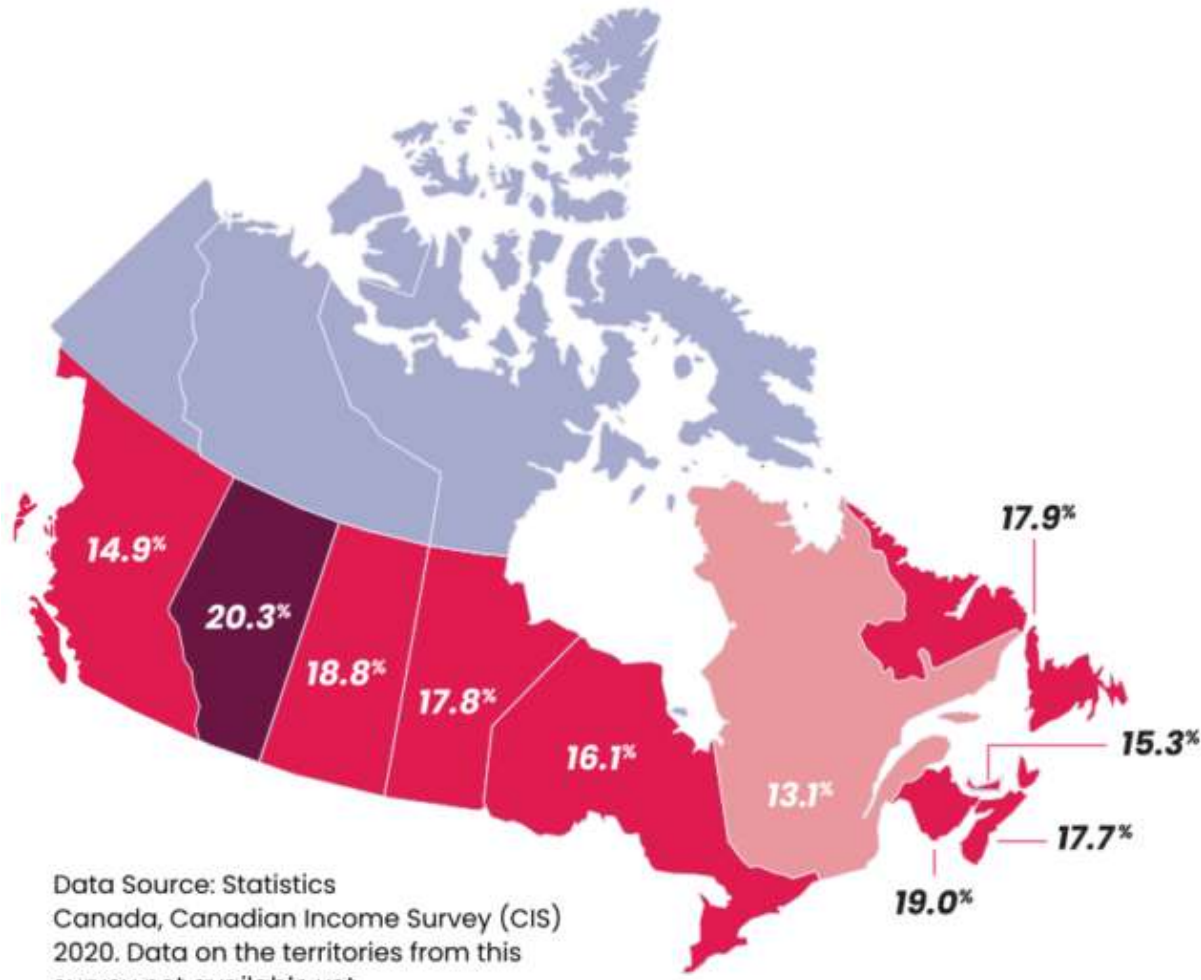
Digital twins  
for cyber-biophysical systems

# Prevalence of Household Food Insecurity by Province, 2021



Data Source: Statistics Canada, Canadian Income Survey (CIS) 2020. Data on the territories from this survey not available yet.

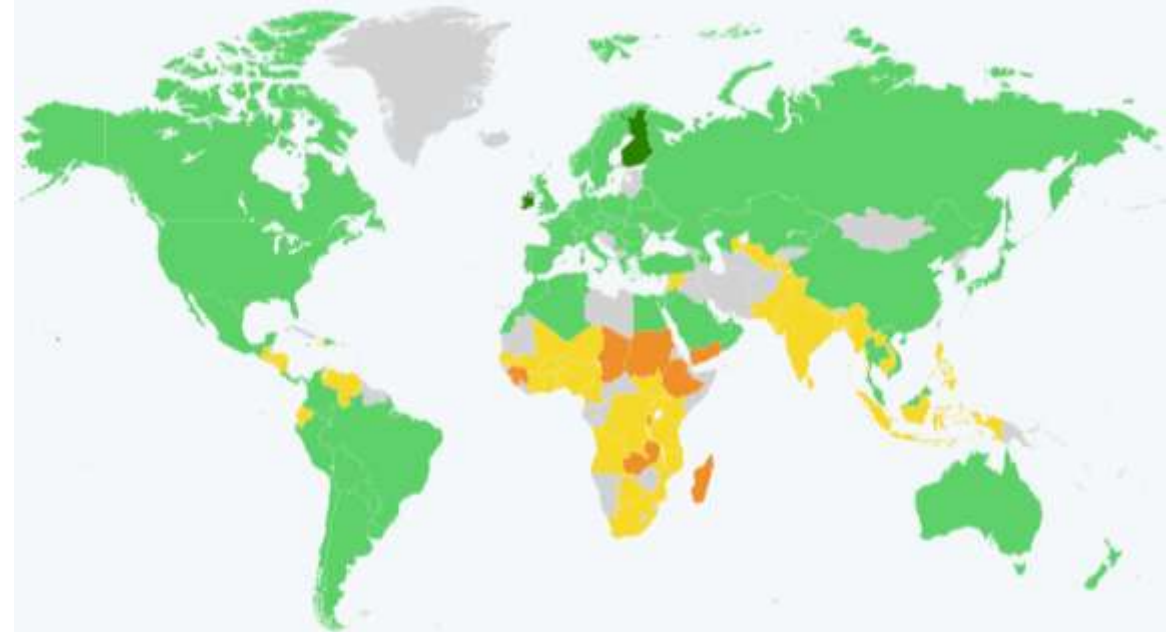
## Prevalence of Household Food Insecurity by Province, 2021



## The Global State of Food Security

Best and worst performing countries for food security in 2020\*

- Best performance
- Good performance
- Moderate performance
- Need improvement

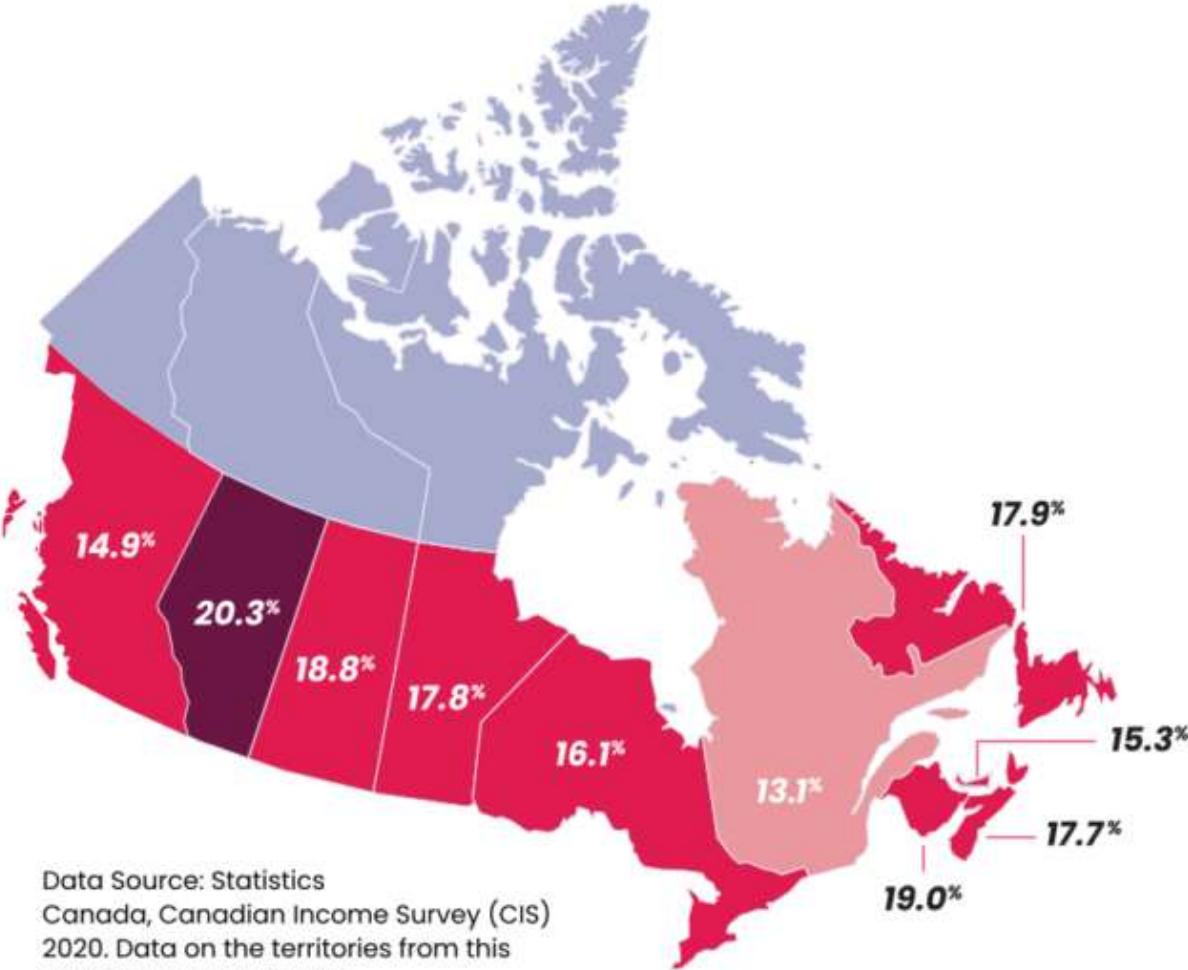


\* Affordability, availability, safety, quality and natural resources of food based on 59 unique indicators across 113 countries.

Source: Economist Intelligence Unit



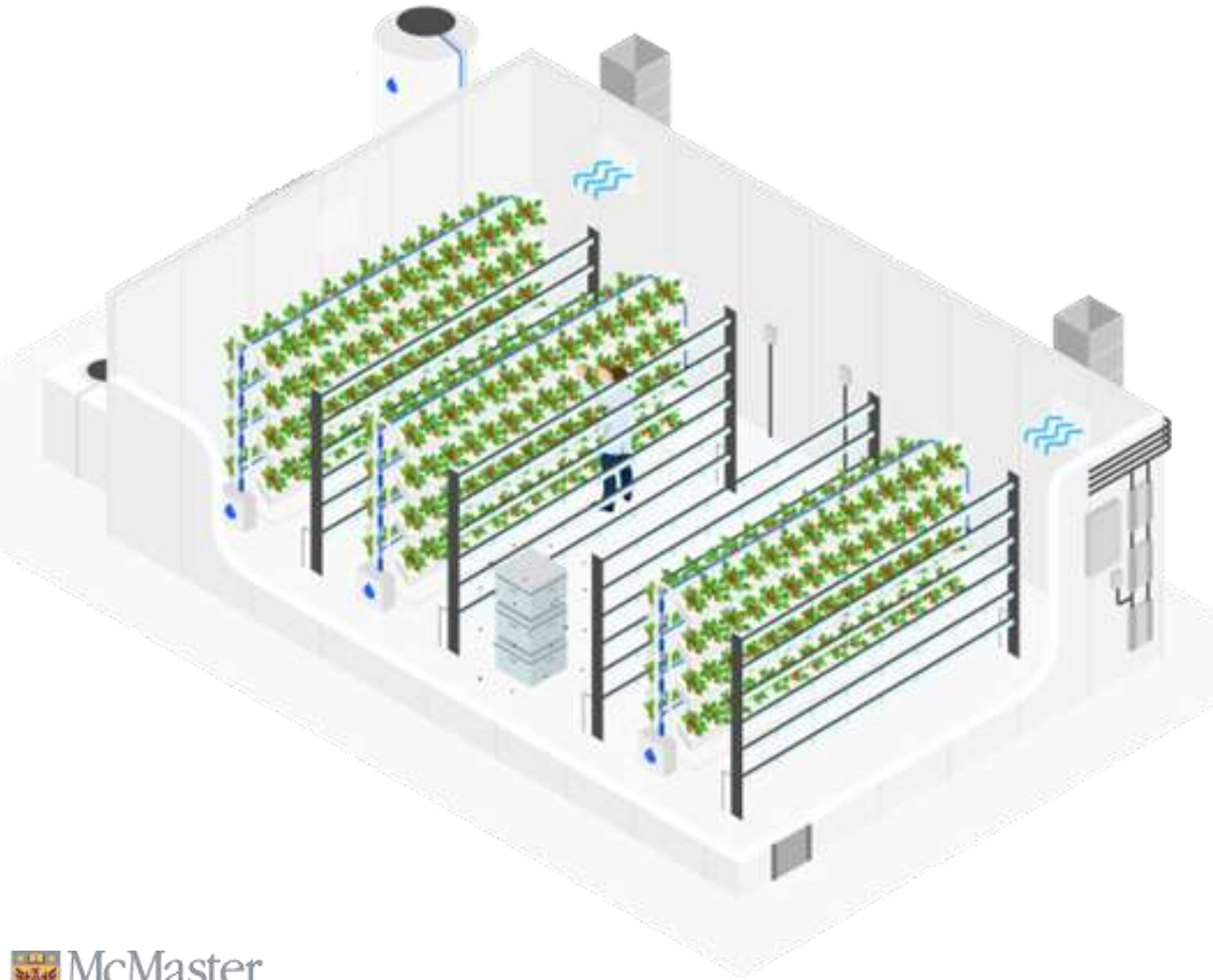
# Prevalence of Household Food Insecurity by Province, 2021



Data Source: Statistics Canada, Canadian Income Survey (CIS) 2020. Data on the territories from this survey not available yet.



# Controlled Environment Agriculture (CEA)



# The challenge of CEA: **control is hard**

*Maximize crop-to-energy ratio*

*Reduce waste*

...

*Each bush must produce 50 grams of*

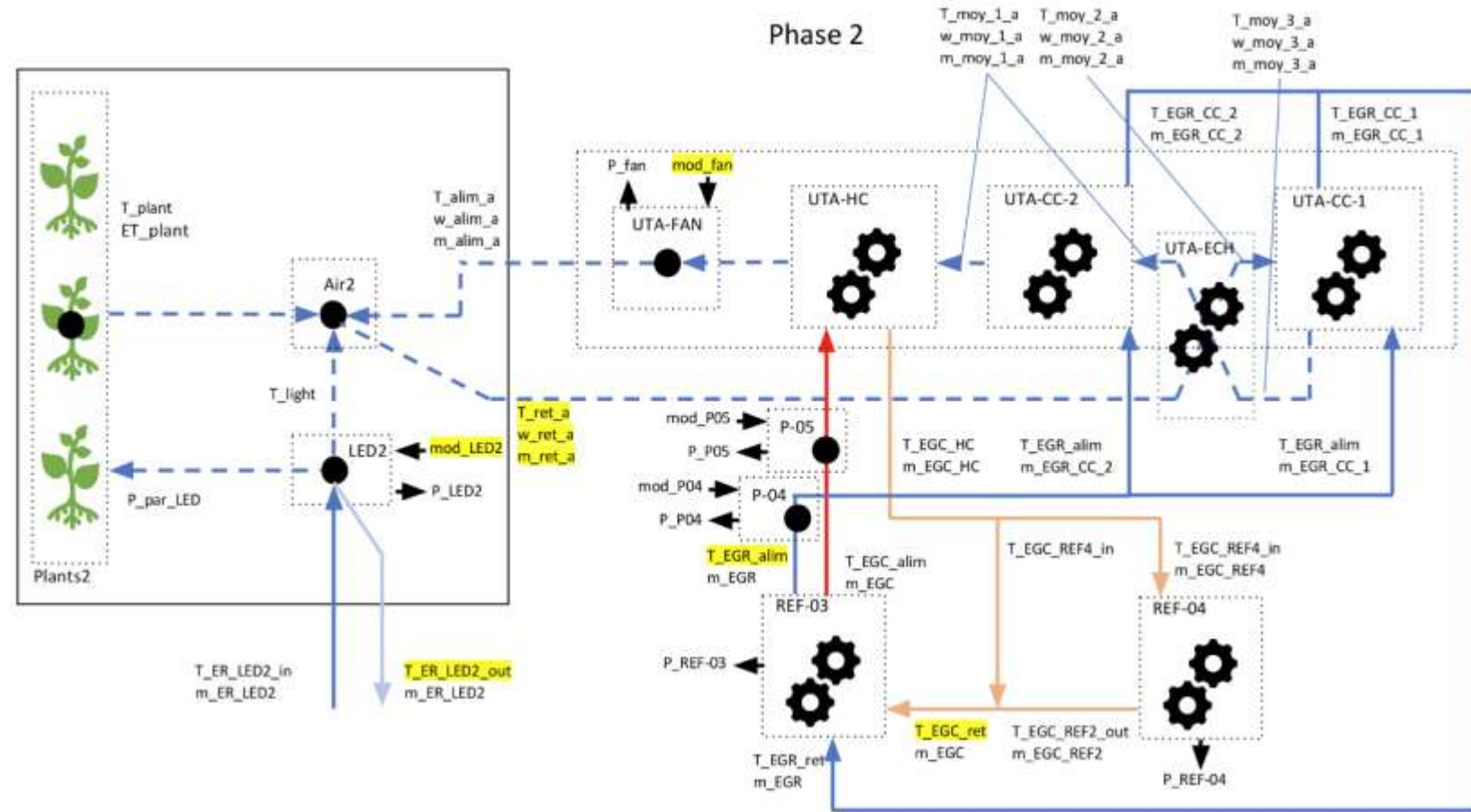


*per week*

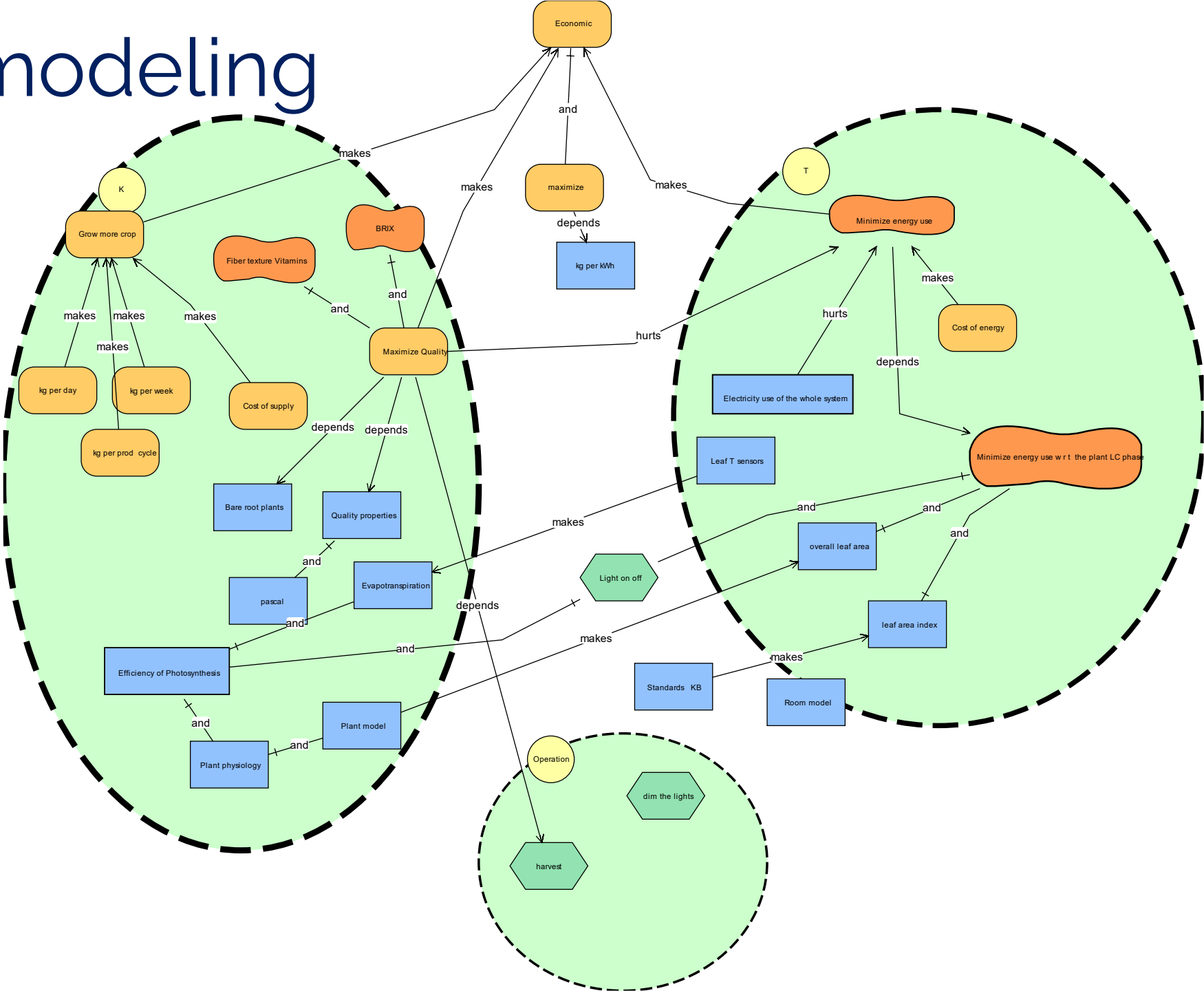
*...starting two months from now*



# Expressing expert processes

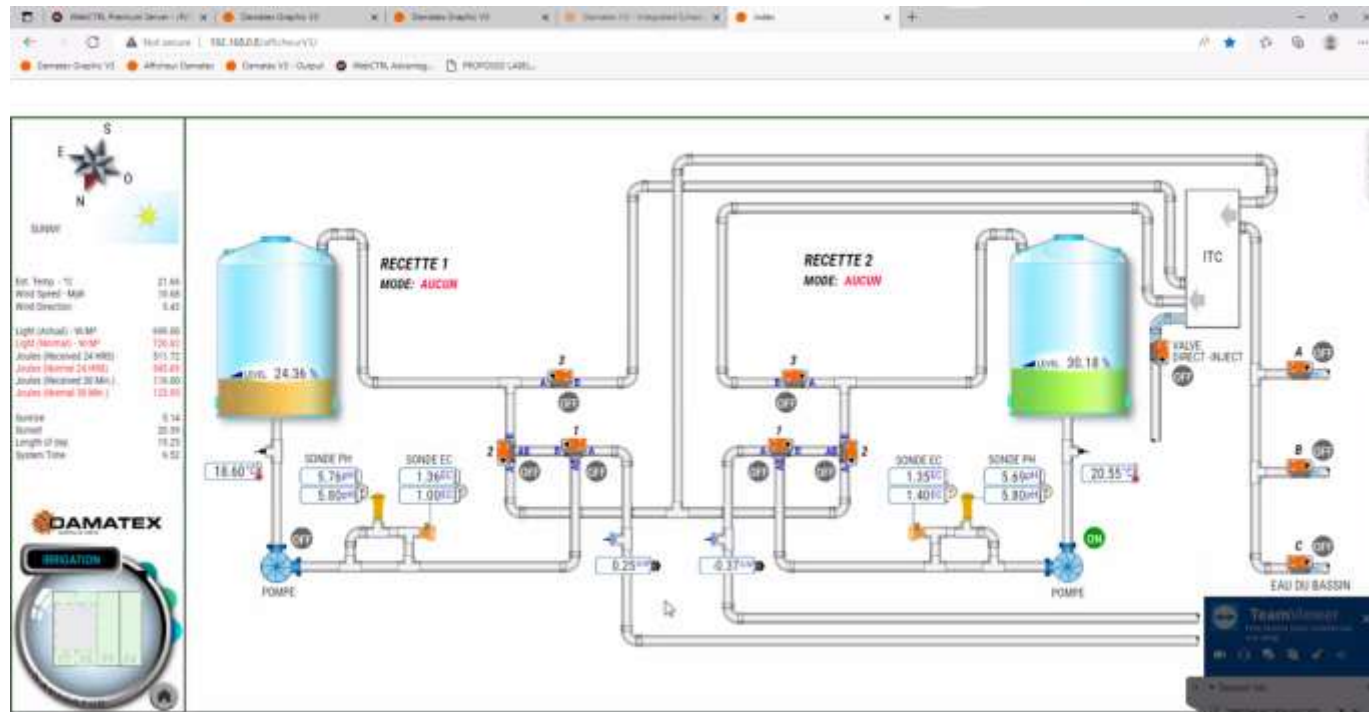
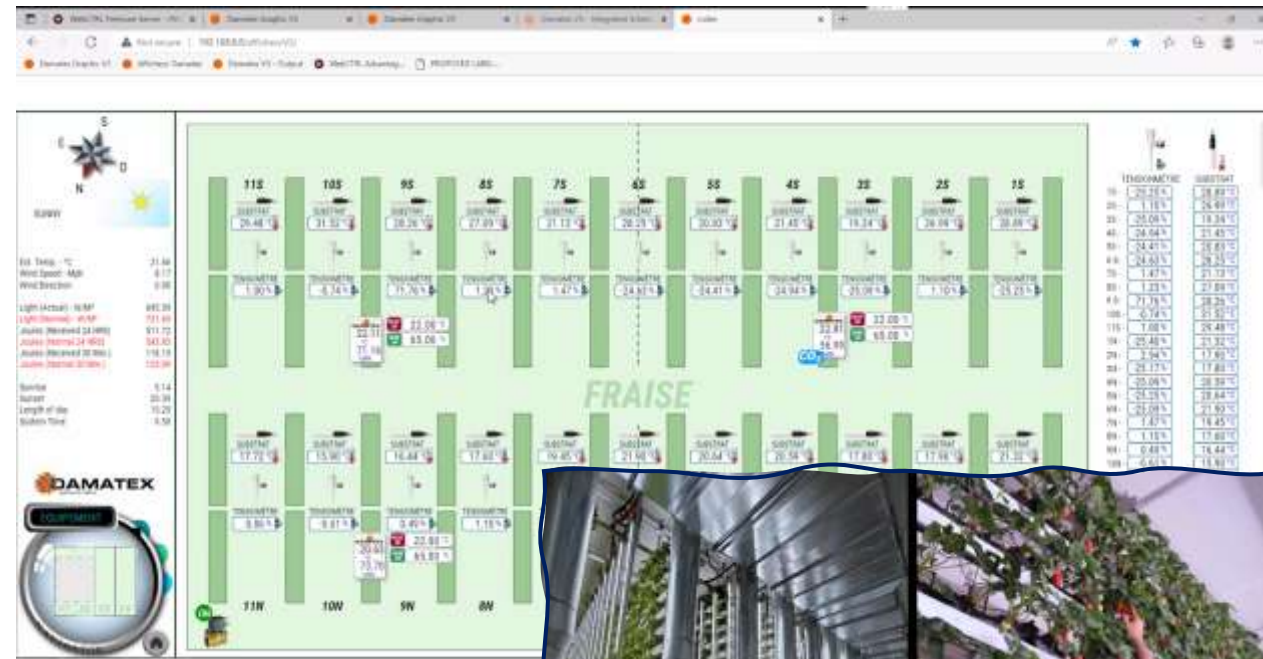
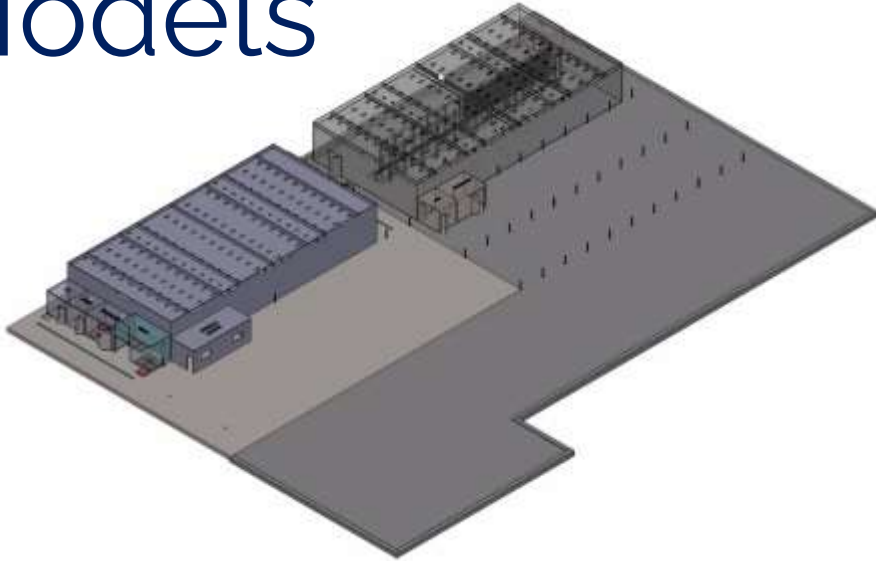


# Goal modeling

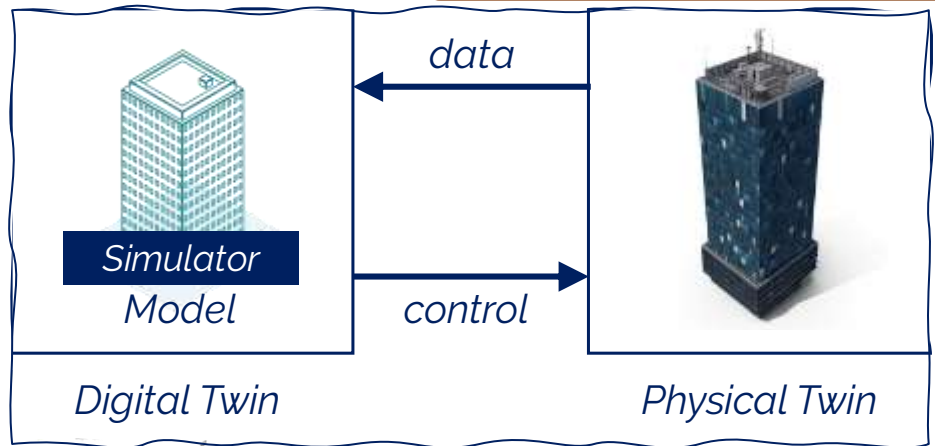
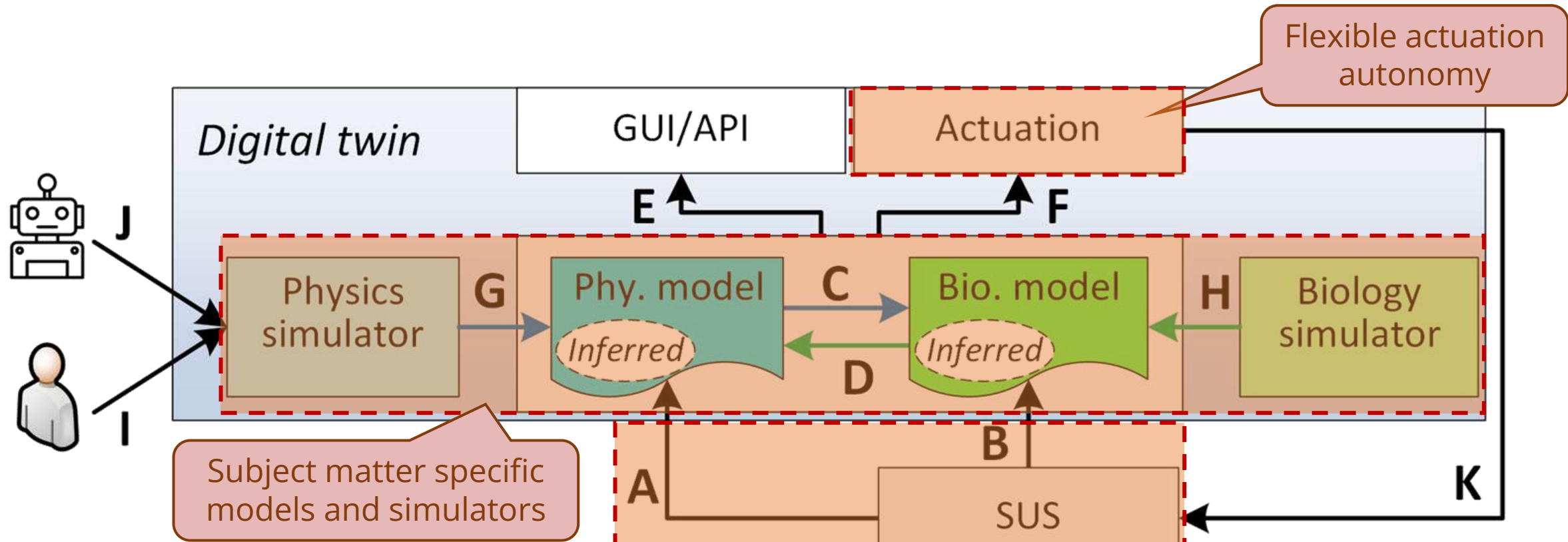




# Models



# DT4CBPS: Conceptual framework and requirements



Flexible data and information collection

# Environment

## Room

HVAC



Lights



Irrigation



Negri, E., Fumagalli, L., Cimino, C. and Macchi, M.  
 FMU-supported simulation for CPS digital twin.  
*Procedia manufacturing*: 28, pp.201-206, 2019.

**fmi** Functional Mock-Up Interface

**fmi** Functional Mock-Up Interface

$$\frac{dy}{dt} = ky$$

**MATLAB**

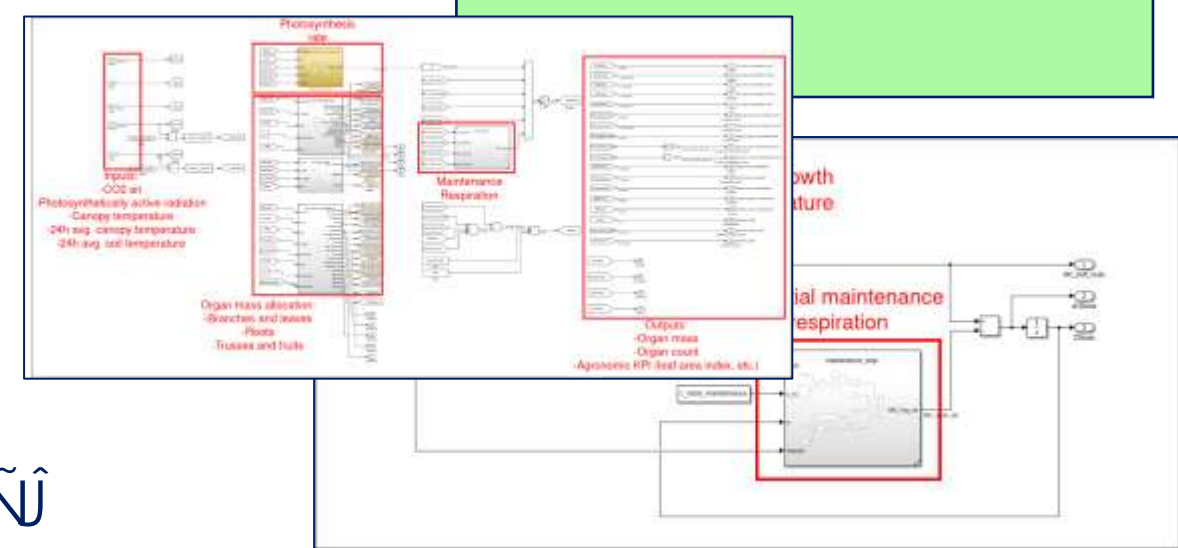
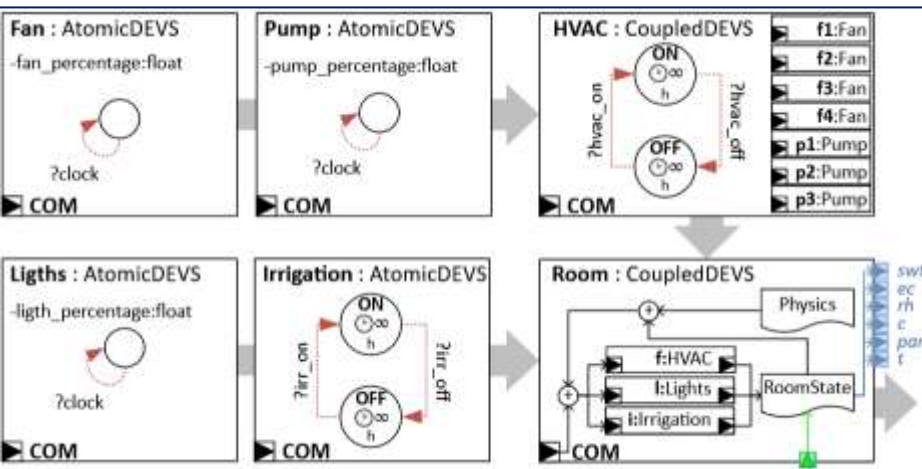
**SIMULINK**

# Crop

Energy



Growth



# Digital Twins for Cyber-Biophysical Systems: Challenges and Lessons Learned

Istvan David<sup>\*✉</sup>, Pascal Archambault<sup>\*</sup>, Quentin Wolak<sup>\*</sup>, Cong Vinh Vu<sup>\*</sup>,  
Timothé Lalonde<sup>†</sup>, Kashif Riaz<sup>†</sup>, Eugene Syriani<sup>\*✉</sup>, Houari Sahraoui<sup>\*✉</sup>  
<sup>\*</sup>DIRO, Université de Montréal, Canada – {firstname.lastname}@umontreal.ca  
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## AI on the farm: A new path to food self-sufficiency

UDEMNOUVELLES | 03/16/2022 | CAROLINE BOILY

<https://nouvelles.umontreal.ca/en/article/2022/03/16/ai-on-the-farm-a-new-path-to-food-self-sufficiency/>

## Des algorithmes pour transformer l'agriculture hivernale

L'intelligence artificielle s'invite dans fermes verticales de l'entreprise québécoise Ferme d'hiver, qui ambitionne de proposer une solution de rechange technologique et carboneutre à l'importation de fruits et légumes pendant la saison froide.

<https://lactualite.com/techno/des-algorithmes-pour-transformer-lagriculture-hivernale/>



<https://mydigitalpublication.com/publication/?m=1281&i=805712&p=22&ver=html5>

## What the future brings

Exploring sustainable solutions for greenhouse adaptation and survival



# AI Simulation by Digital Twins

# Hype Cycle for Artificial Intelligence, 2023



## AI Simulation

The combined application of AI and simulation technologies to jointly develop AI agents and the simulated environments in which they can be trained, tested and sometimes deployed.

Some of the best simulators today: in DTs

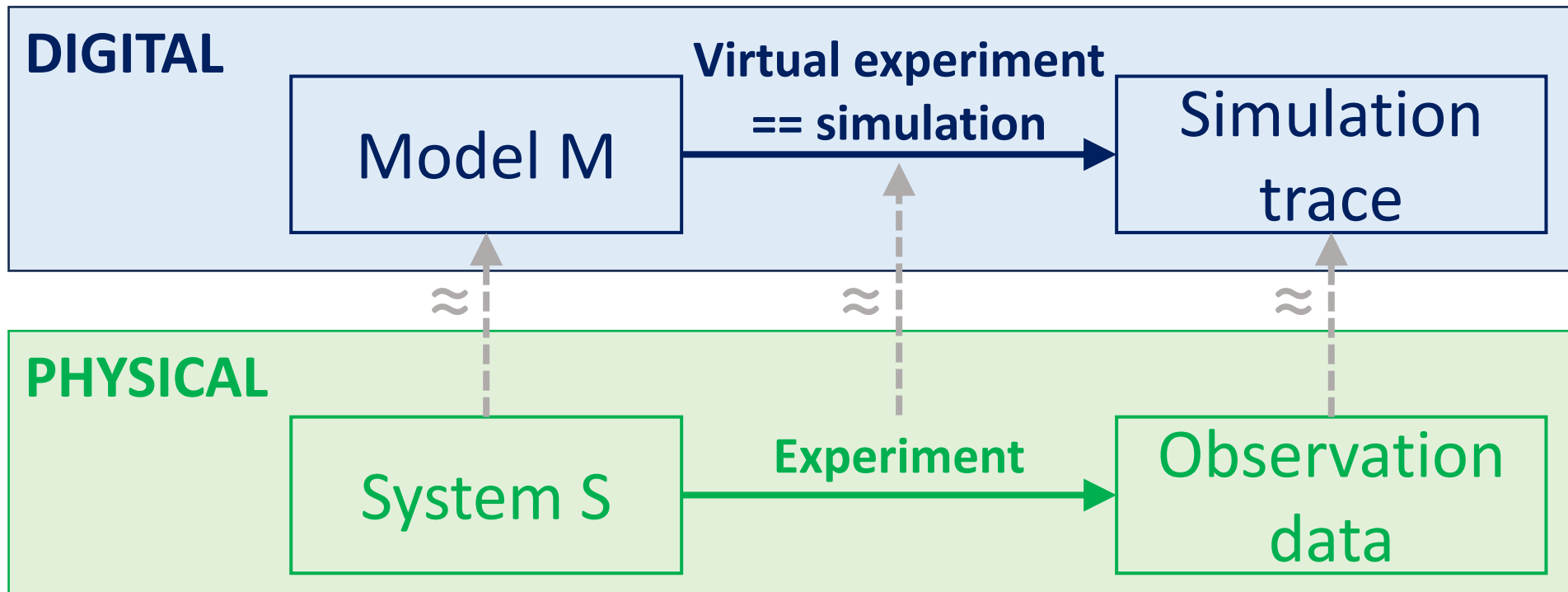


Xiaoran (Sharon) Liu

- Xiaoran Liu and Istvan David. "AI Simulation by Digital Twins: Systematic Survey of the State of the Art and a Reference Framework". In: ACM/IEEE MODELS-C / EDTConf'24. ACM, 2024.
- Xiaoran Liu and Istvan David. "AI Simulation by Digital Twins: Reference Framework and Mapping on ISO 23247". In: Soft. Syst. Model. Springer. 2025. [Under review]

Opportunity:

***purposeful*** experimentation to acquire missing data







Domains/problems



Digital twins



AI/ML



Lifecycle models



Challenges/limitations

# AI Simulation by Digital Twins

Systematic Survey of the State of the Art and a Reference Framework

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### ABSTRACT

Insufficient data volume and quality are particularly pressing challenges in the adoption of modern subsymbolic AI. To alleviate these challenges, AI simulation recommends developing virtual training environments in which AI agents can be safely and efficiently developed. Digital twins open new avenues in AI simulation, as these high-fidelity virtual replicas of physical systems are equipped with state-of-the-art simulators and the ability to further interact with the physical system for additional data collection. In this paper, we report on our systematic survey of digital twin-enabled AI simulation. By analyzing 22 primary studies, we identify technological trends and derive a reference framework to situate digital twins and AI components. Finally, we identify challenges and research opportunities for prospective researchers.

### CCS CONCEPTS

• General and reference → Surveys and overviews; • Computing methodologies → Learning settings.

### KEYWORDS

AI, artificial intelligence, data science, deep neural networks, digital twins, lifecycle model, machine learning, neural networks, reinforcement learning, SLR, subsymbolic AI, survey, training

### ACM Reference Format:

Xiaoran Liu and Istvan David. 2024. AI Simulation by Digital Twins: Systematic Survey of the State of the Art and a Reference Framework. In *Proceedings of International Conference on Engineering Digital Twins (EDTConf'24)*. ACM, New York, NY, USA, 12 pages. <https://doi.org/XXXXXXX.XXXXXXX>

### 1 INTRODUCTION

Modern artificial intelligence (AI) is enabled by massive volumes of data processed by powerful computational methods [84]. This is a stark contrast with traditional AI which is supported by symbolic methods and logic [69]. The volume and quality of available data to train AI is the cornerstone of success in modern AI. However, accessing and harvesting real-world data is a substantial barrier due to its scarcity, cost, or difficult accessibility, hindering the development of precise and resilient AI models. For example, in manufacturing,

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EDTConf'24, September 23-24, 2024, Linz, Austria  
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ACM ISBN 978-x-xxxx-xxxx-x/YY/MM  
<https://doi.org/XXXXXXX.XXXXXXX>

proprietary data, data silos, and sensitive operational procedures complicate the acquisition of data [43]. Data-related barriers, in turn, limit the applicability of otherwise powerful AI methods.

AI simulation is a prime candidate for alleviating these problems. As defined by Gartner recently, AI simulation is the technique of "the combined application of AI and simulation technologies to jointly develop AI agents and the simulated environments in which they can be trained, tested and sometimes deployed. It includes both the use of AI to make simulations more efficient and useful, and the use of a wide range of simulation models to develop more versatile and adaptive AI systems" [47]. After modeling the phenomenon or system at hand, a simulation of the model computes the dynamic input/output behavior [77], representative of the system. A simulation produces data, called the simulation trace, that represents the behavior of the simulated system over time. These traces can be used as training data for AI agents, assuming that the simulation is a faithful, valid and detailed representation of the modeled system, and that the simulation can still be executed efficiently and in a timely manner.

With the emergence of digital twins (DT) [54], the quality attributes of simulators have improved as well. Simulators are first-class components of DTs [36] and enablers of sophisticated services, e.g., real-time adaptation [73], predictive analytics [62], and process control in manufacturing [28]. These advanced services require well-performing and high-fidelity simulators—the types of simulators that align well with the goals of AI simulation.

A recent interview study on DTs with nineteen academic and industry participants by Muctadir et al. [58] mentions that "machine learning and reinforcement learning could possibly be combined with DTs in the future, to help to learn about complex systems (i.e., safety-critical systems) in a virtual environment, when this is difficult to do on the real-world system." Similar ambitions have been identified by Mihai et al. [56] as future prospects of DTs. Indeed, the improvements in simulator engineering that have been driven by DTs, are generating interest in DTs for AI simulation. It is plausible to anticipate that the next generation of AI simulation techniques will be heavily influenced by the further advancements of DT technology [51, 66]. Therefore, it is important to understand the state of affairs in digital twinning for AI simulation purposes, prepare for the related challenges, and set targeted research agendas.

This work marks a step towards converging AI simulation and DT technology. We review the state of the art on AI simulation by DTs, derive a framework, identify trends in system organization, AI flavors, and simulation, and outline future avenues of research.

*Context and scope.* In this work, we focus on AI simulation by digital twins. We acknowledge the utility of the other direction, i.e., simulators of DTs being enabled by AI [55]; however, we consider such works outside the scope of the current study.

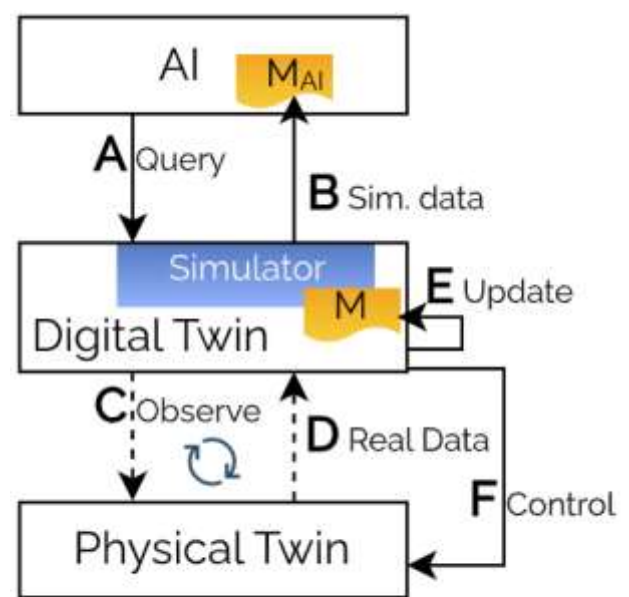
 Domains/problems

 Digital twins

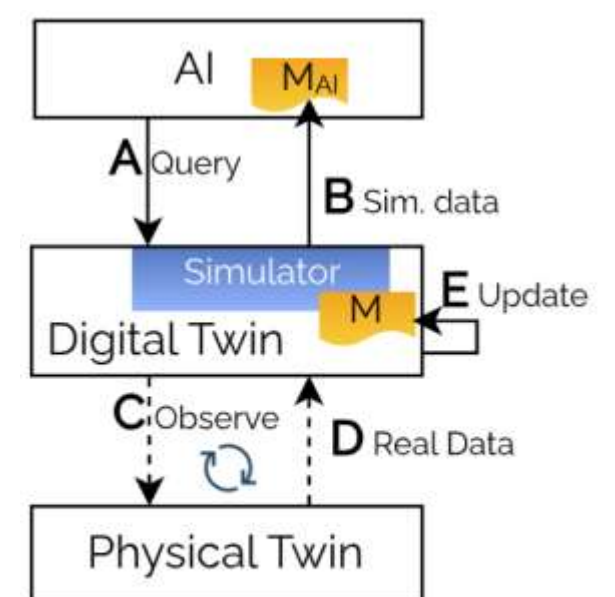
 AI/ML

 Lifecycle models

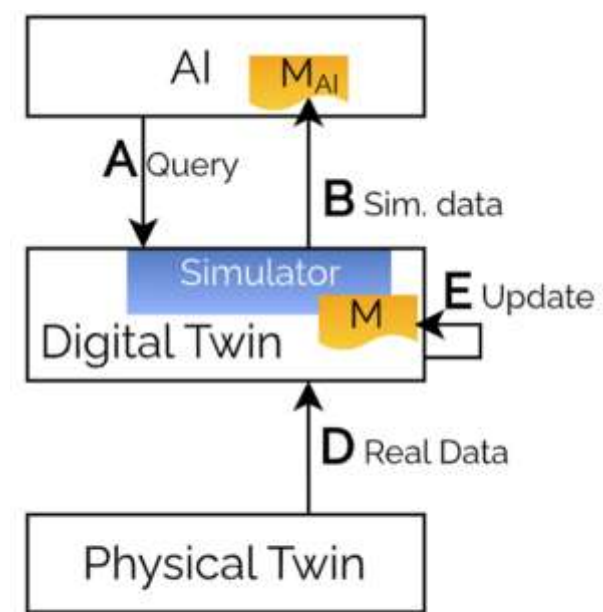
 Challenges/limitations



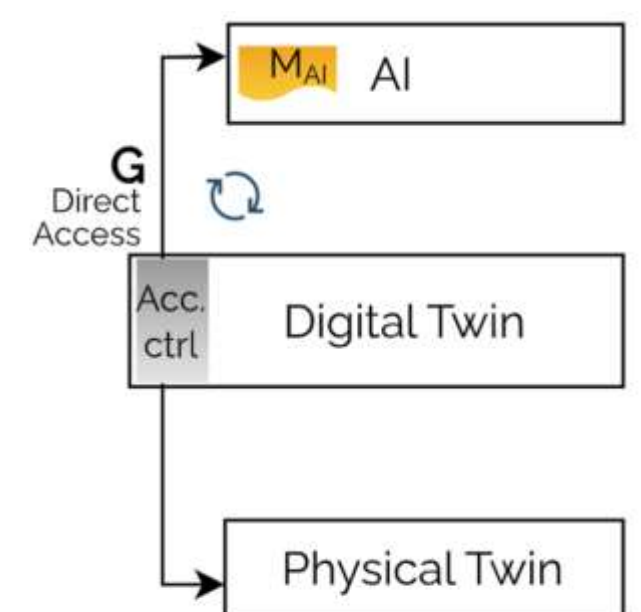
(a) Experimentable DT



(b) Experimentable DS



(c) Experimentable Model



(d) Policy DT

 Domains/problems

# Deep learning proliferates

 Digital twins

**Table 7: AI methods**

 AI/ML

AI	#Studies	Studies
RL	18 (81.8%)	
↳ DRL	13 (59.1%)	
↳ Value	8 (36.4%)	[2, 10, 14, 15, 18, 19, 21, 22]
↳ Policy	5 (22.7%)	[6, 8, 9, 11, 13]
↳ Vanilla	5 (22.7%)	[4, 7, 16, 17, 20]
DL	4 (18.2%)	[1, 3, 5, 12]
TL	1 (4.5%)	[16]

 Lifecycle models

 Challenges/limitations



Domains/problems

# Deep learning proliferates



Digital twins



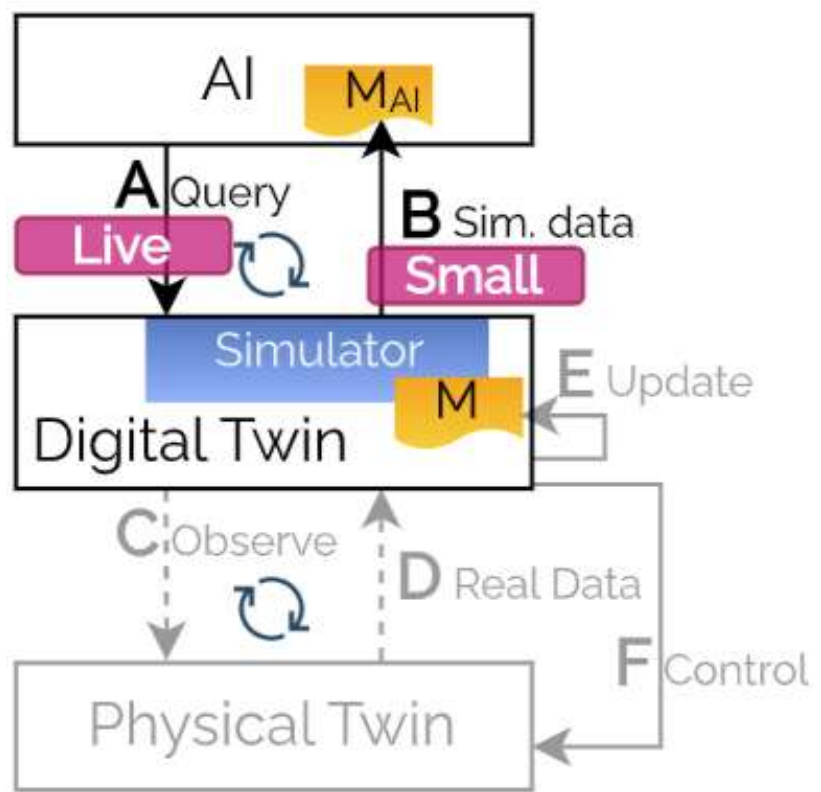
AI/ML



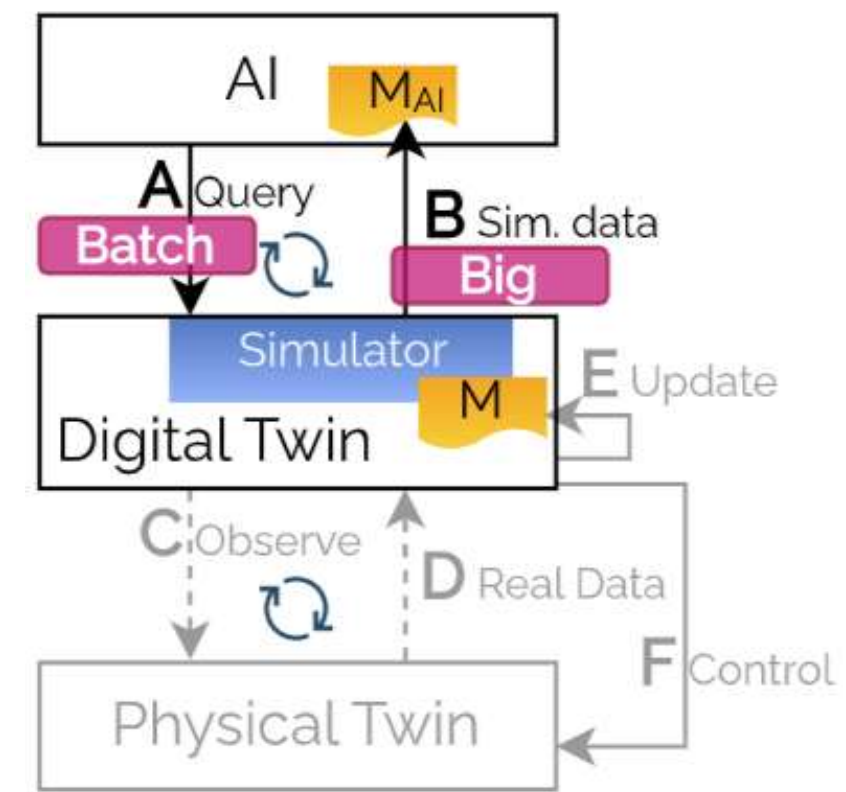
Lifecycle models



Challenges/limitations



**(a) Reinforcement Learning**



**(b) Deep Learning**

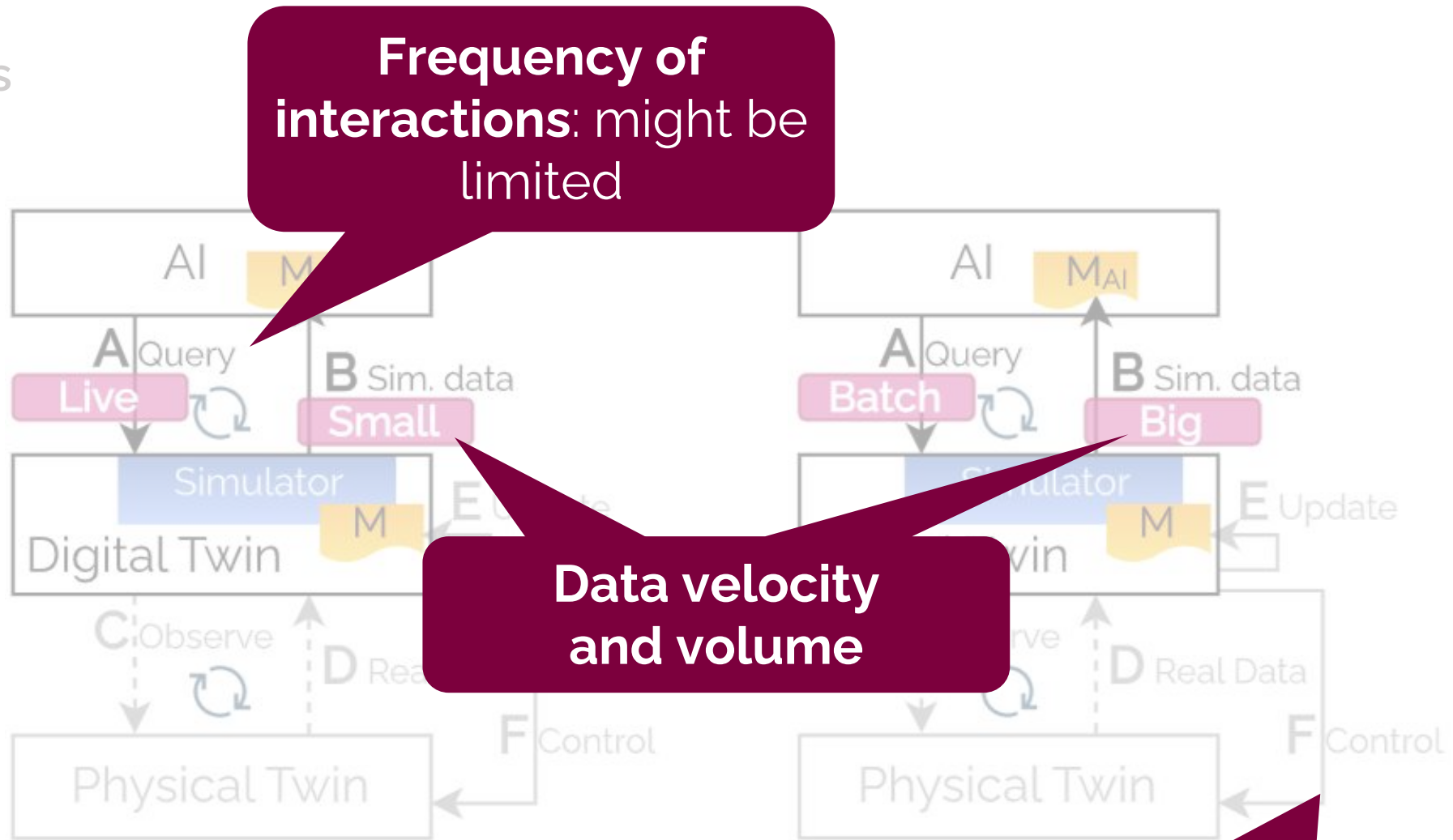
 Domains/problems

 Digital twins

 AI/ML

 Lifecycle models

 Challenges/limitations



**Frequency of interactions:** might be limited

**Data velocity and volume**

**Instrumentation**

(a) Reinforcement Learning



Sustainability  
*by and of* Digital twins

# Problem: our systems are not sustainable



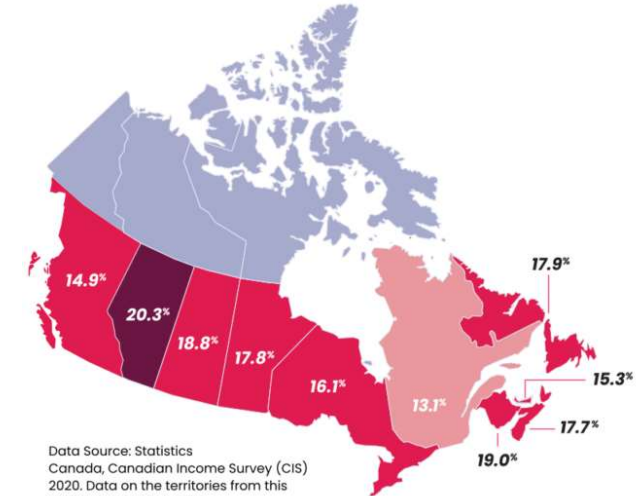
Systems Engineering

“meeting the needs of the present without compromising the ability of future generations to meet their own needs (Brundtland)

“Technical sustainability addresses the **long-term use** of software-intensive systems and their appropriate **evolution** in a constantly changing execution environment

P. Lago, S. A. Koçak, I. Crnkovic, and B. Penzenstadler. Framing Sustainability as a Property of Software Quality, Commun. ACM, vol. 58, no. 10, pp. 70–78, Sep. 2015.

Prevalence of Household Food Insecurity by Province, 2021



Data Source: Statistics Canada, Canadian Income Survey (CIS) 2020. Data on the territories from this survey not available yet.

NATURE AND ENVIRONMENT | EUROPE

## Air pollution: Nearly everyone in Europe breathing bad air

Rodrigo Menegat Schiavinski

04/10/2023

With the EU voting on new air quality rules, satellite data shows 98% of people face pollution above limits recommended by the World Health Organization.

98%



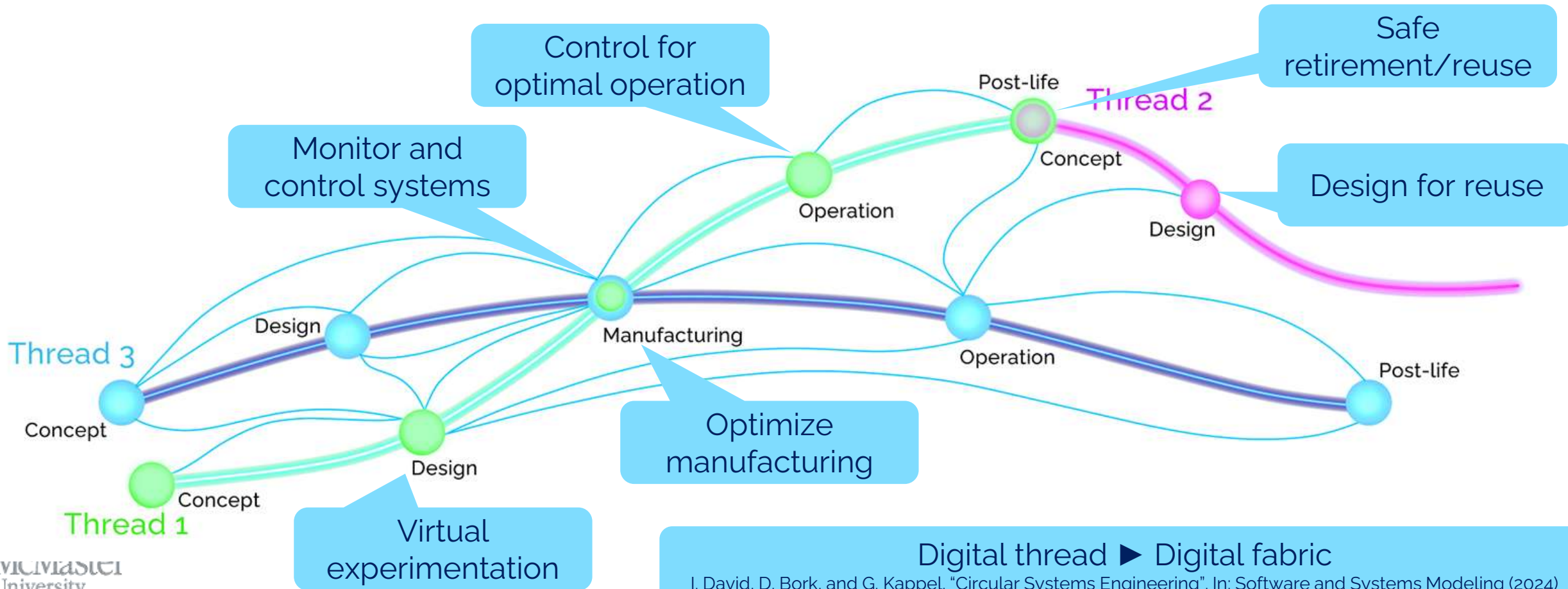
*Digital technology to the rescue?*



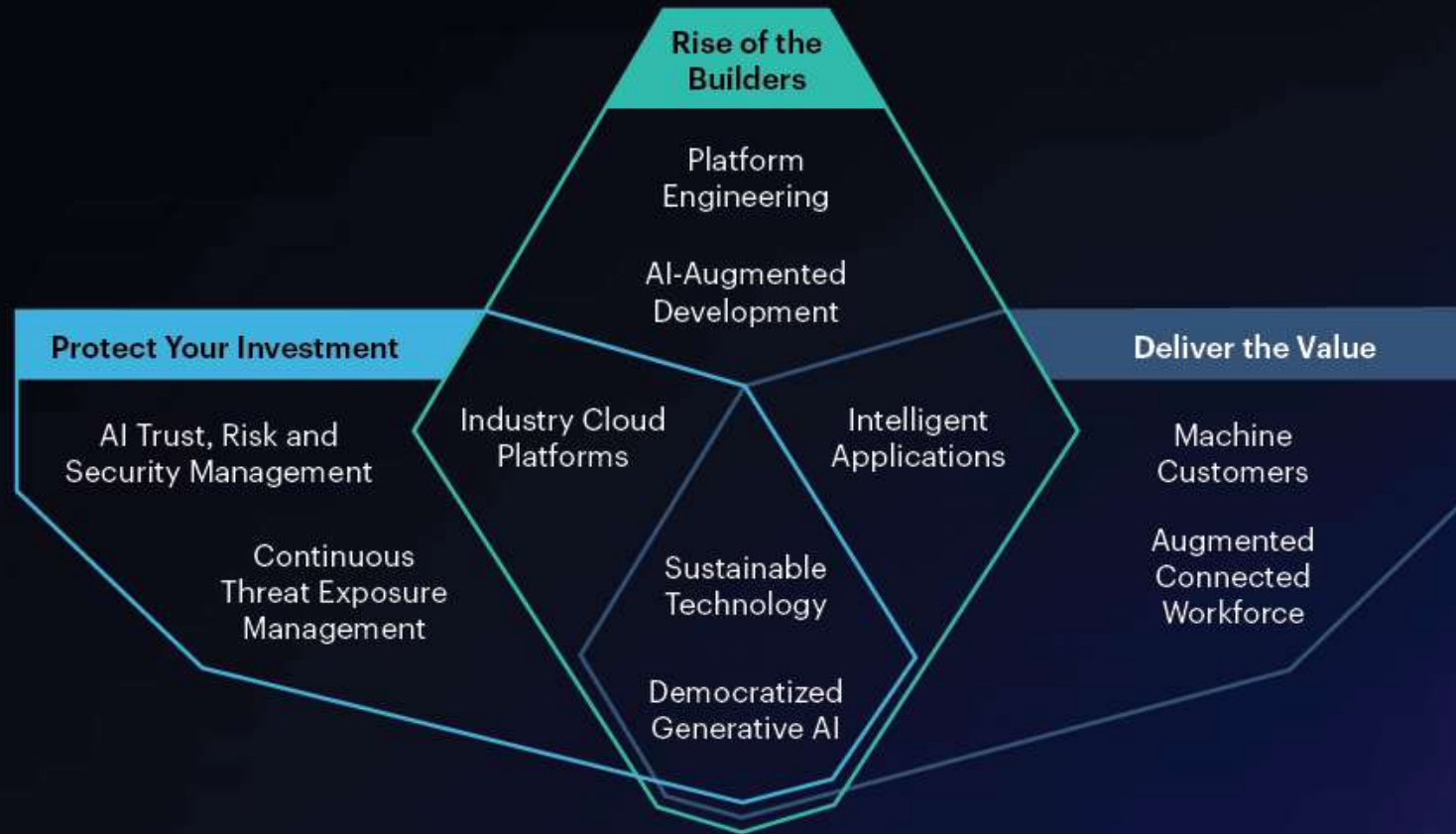
# Digital Twins for Sustainable Systems

**60% of organizations believe** Digital Twin technology is critical to improving sustainability efforts.

(CapGemini, 2022)



# Top Strategic Technology Trends 2024



# Top Strategic Technology Trends 2024

Rise of the  
Builders

By 2027, 80% of CIOs will have performance metrics tied to the sustainability of the IT organization.

Source: Gartner

Threat Exposure  
Management

Technology

Workforce

Democratized  
Generative AI



EMERGING TECHNOLOGIES

# 9 ways AI is helping tackle climate change

Feb 12, 2024

WORLD  
ECONOMIC  
FORUM



*Digital technology to the rescue!*





EMERGING TECHNOLOGIES

## 9 ways AI is helping tackle climate change

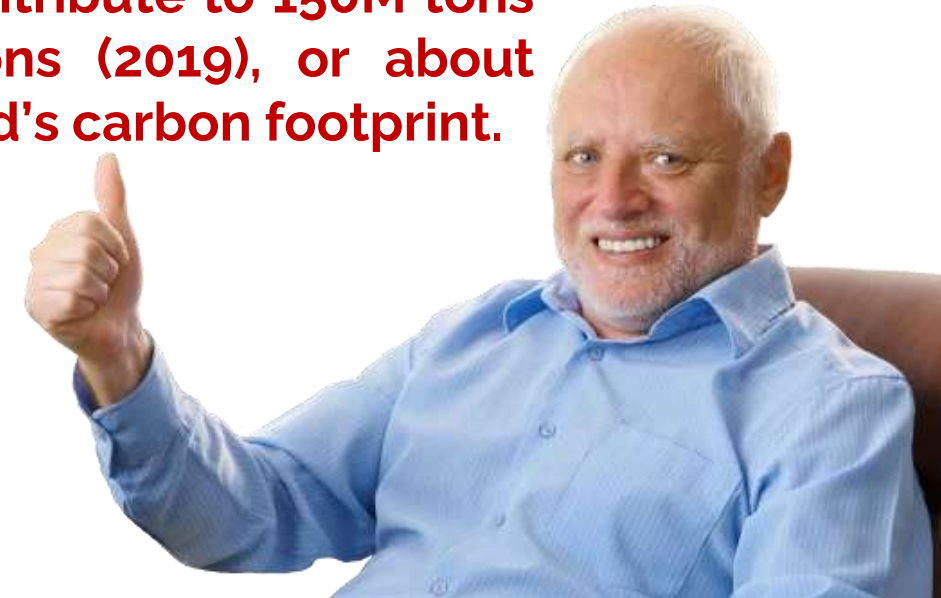
Feb 12, 2024



*Digital technology to the rescue!*



**It is estimated that emails sent globally can contribute to 150M tons of CO<sub>2</sub> emissions (2019), or about 0.3% of the world's carbon footprint.**



# Unsustainable computing and ICT



## Manufacturing

- 33% of global energy consumption
- 36% of global CO<sub>2</sub> emissions



## Hardware and e-devices

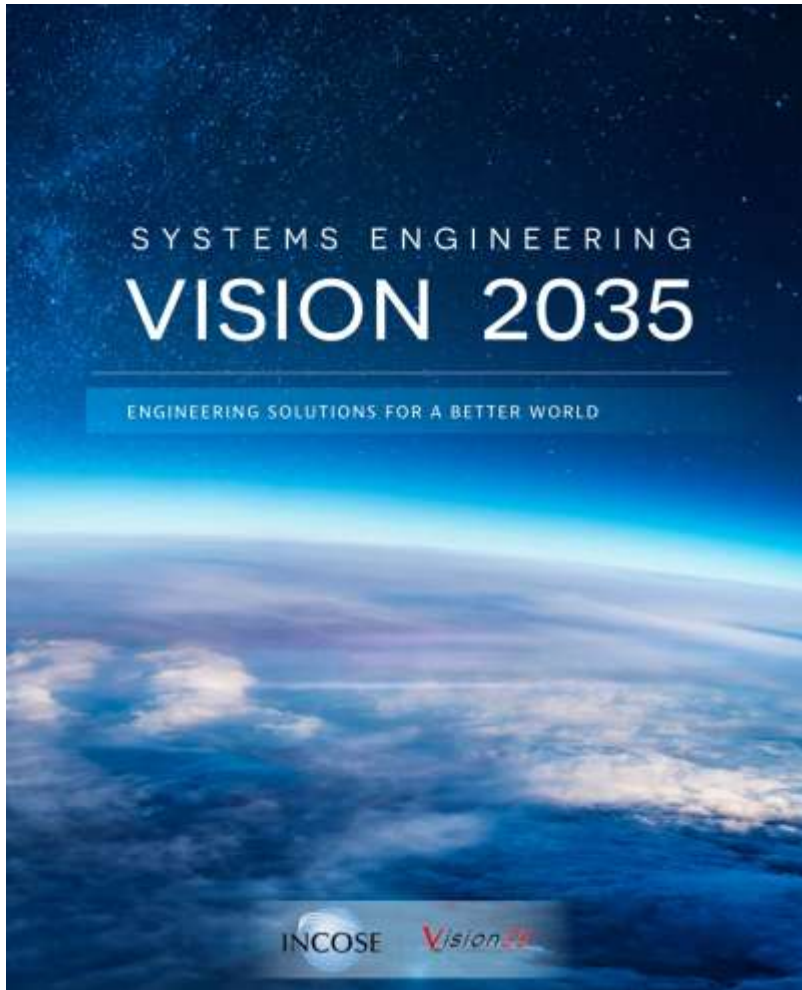
- *E-waste* is recognized by the World Economic Forum as the fastest-growing category of waste

## ICT

- 2-4% of global CO<sub>2</sub> emissions (~avionics sector)
- 2040 prediction: ~14%
- To follow suit, decrease CO<sub>2</sub> emissions by
  - 42% by 2030
  - 72% by 2040
  - 91% by 2050



We need to engineer sustainable systems.  
By sustainable methods.

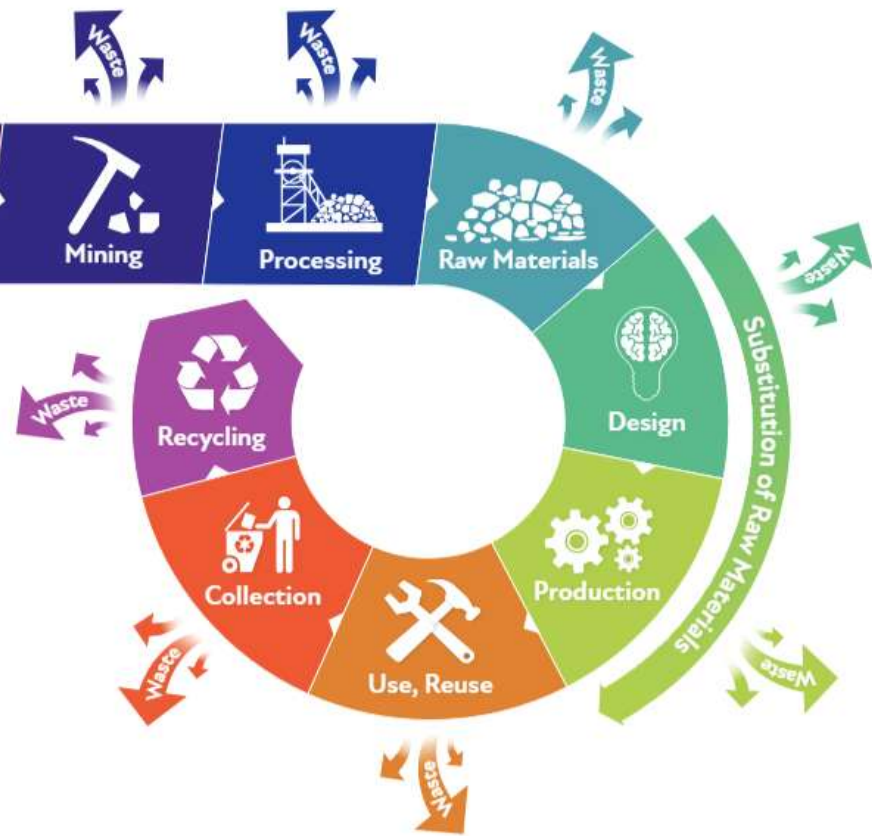




# GLOBAL MEGATREND 1

ENVIRONMENTAL SUSTAINABILITY BECOMES A HIGH PRIORITY

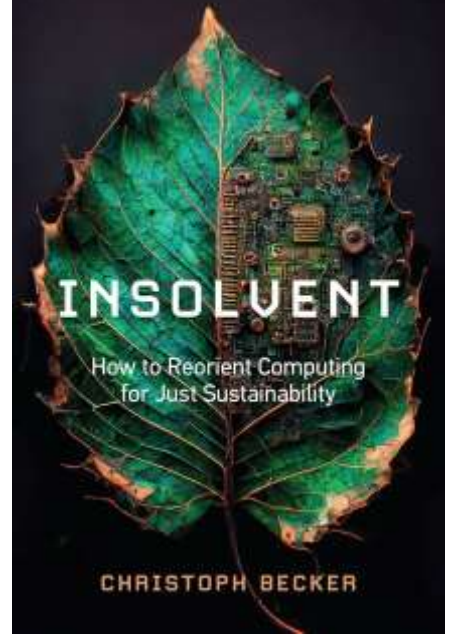
► Society will place great importance on reuse, giving rise to Circular Economies.



- Impacts of human activity on climate will be ingrained in assessments of engineered systems and public/private policies.
- Many systems, both straightforward and novel, will arise to mitigate the deleterious impacts of climate change, such as global warming.

THE GLOBAL PUBLIC WILL TRUST AND REWARD SYSTEMS PROVIDERS AND OPERATORS THAT PRODUCE SUSTAINABLE SYSTEMS AND BEHAVE IN A SUSTAINABLE MANNER.

# Demand for sustainable systems



**GLOBAL MEGATREND 1 ENVIRONMENTAL SUSTAINABILITY BECOMES A HIGH PRIORITY**

Consumption of non-renewable resources resulting from economic activity will increasingly require better global management, recycling strategies, sustainable policies, local actions, and supporting systems, such as energy conversion and infrastructure for clean transportation and manufacturing.

Environmental change will result in shifts in living conditions, and impacts bio-diversity, climate, global heat transport, the availability of fresh water, and other natural resources necessary for human sustenance and well-being.

Overall environmental quality will be a priority, requiring global cooperation. The trend toward greater concern for environmental sustainability will result in several key societal and system imperatives.

Engineering for sustainability, a system characteristic, will create a new generation of engineers who routinely assess the societal impacts of engineered systems.

► Impacts of human activity on climate will be ingrained in assessments of engineered systems and public/private policies.

► Many systems, both straightforward and novel, will arise to mitigate the deleterious impacts of climate change, such as global warming.

► Priority will be placed on systems that are more efficient at resource utilization and responsible waste disposal. Though enterprises will continue to struggle with business and consumer pressures to increase consumption, versus environmental prerogatives to reduce waste.

► The global fossil-fuel based energy economy will be transformed to one based on clean and renewable sources.

► Society will place great importance on reuse, giving rise to Circular Economies.

Bus Inf Syst Eng 65(1):1-6 (2023)  
<https://doi.org/10.1007/s12599-022-00784-6>

**EDITORIAL**

**Sustainable Systems Engineering  
 Opportunities and Challenges**

Wil M. P. van der Aalst · Oliver Hinz · Christof Weinhardt

**Modelling Sustainability in Cyber-Physical Systems:  
 A Systematic Mapping Study**

Ankica Barišić<sup>a</sup>, Jácome Cunha<sup>a,b</sup>, Ivan Ruchkin<sup>b</sup>, Ana Moreira<sup>f,e</sup>, João Araújo<sup>f,e</sup>, Moharram Challenger<sup>c</sup>, Dušan Savić<sup>d</sup>, Vasco Amaral<sup>f,e</sup>

**White Paper DRAFT for Discussion**  
 by concerned Education and Enterprise Communities  
 This Paper is a Compendium to the Systems Engineering Vision 2035

**Building the Systems Engineering Workforce of the Future  
 Education, Training and Development of System Engineers**

*Compiled and edited by Prof. Dipl.-Ing. Heinz Stoewer, M.Sc. SAC GmbH and TU Delft, and co-edited by David Nichols, JPL/Caltech.*

“ computing's dominant frame of thinking is **conceptually insufficient** to address our current challenges (...) and computing continues to incur societal debts it cannot pay back

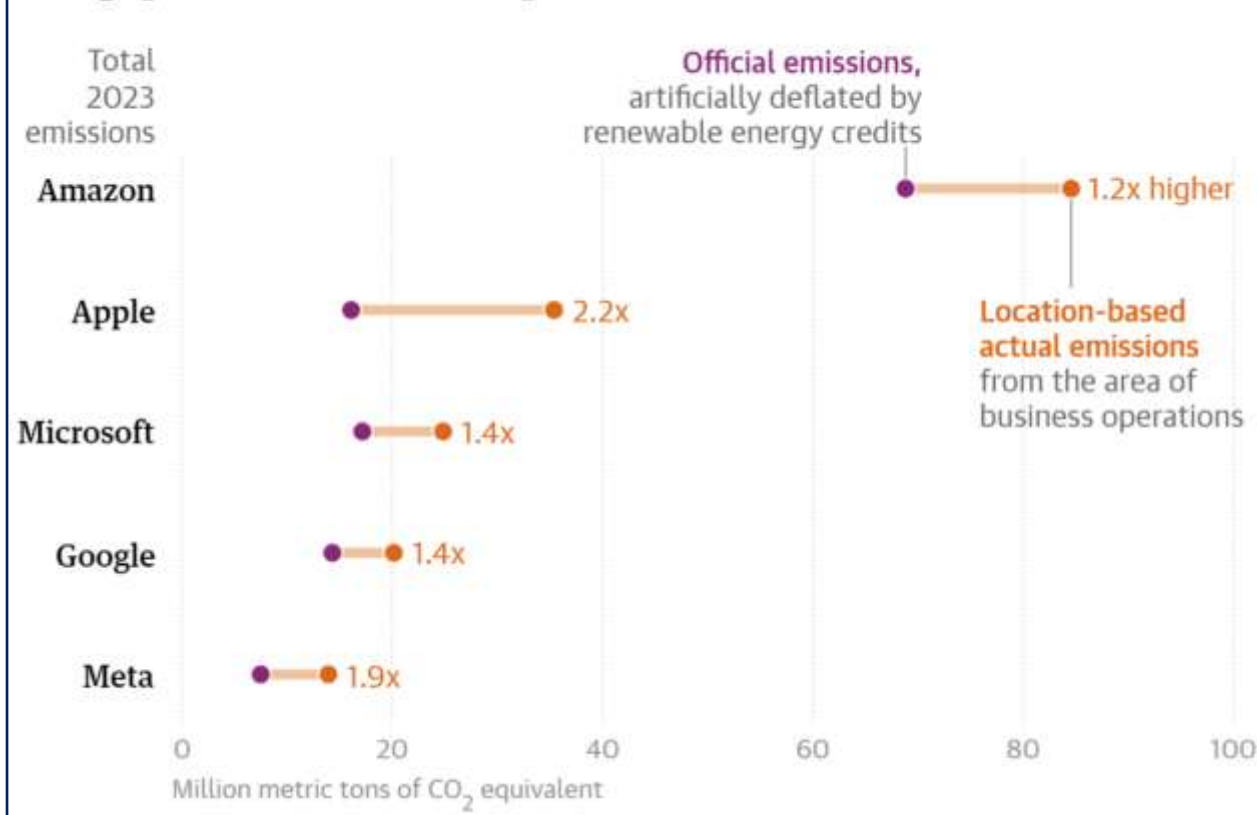
“ I believe, similarly to the EDI statements we have to write today for project proposals and applications, soon we will have to write sustainability statements. And this is not fifteen years from now, but closer to five.

INCOSE Systems Engineering Vision 2035

# We need a better understanding of “sustainability”

**Data center emissions probably 662% higher than big tech claims. Can it keep up the ruse?**

**The gap between tech companies' official and actual emissions**

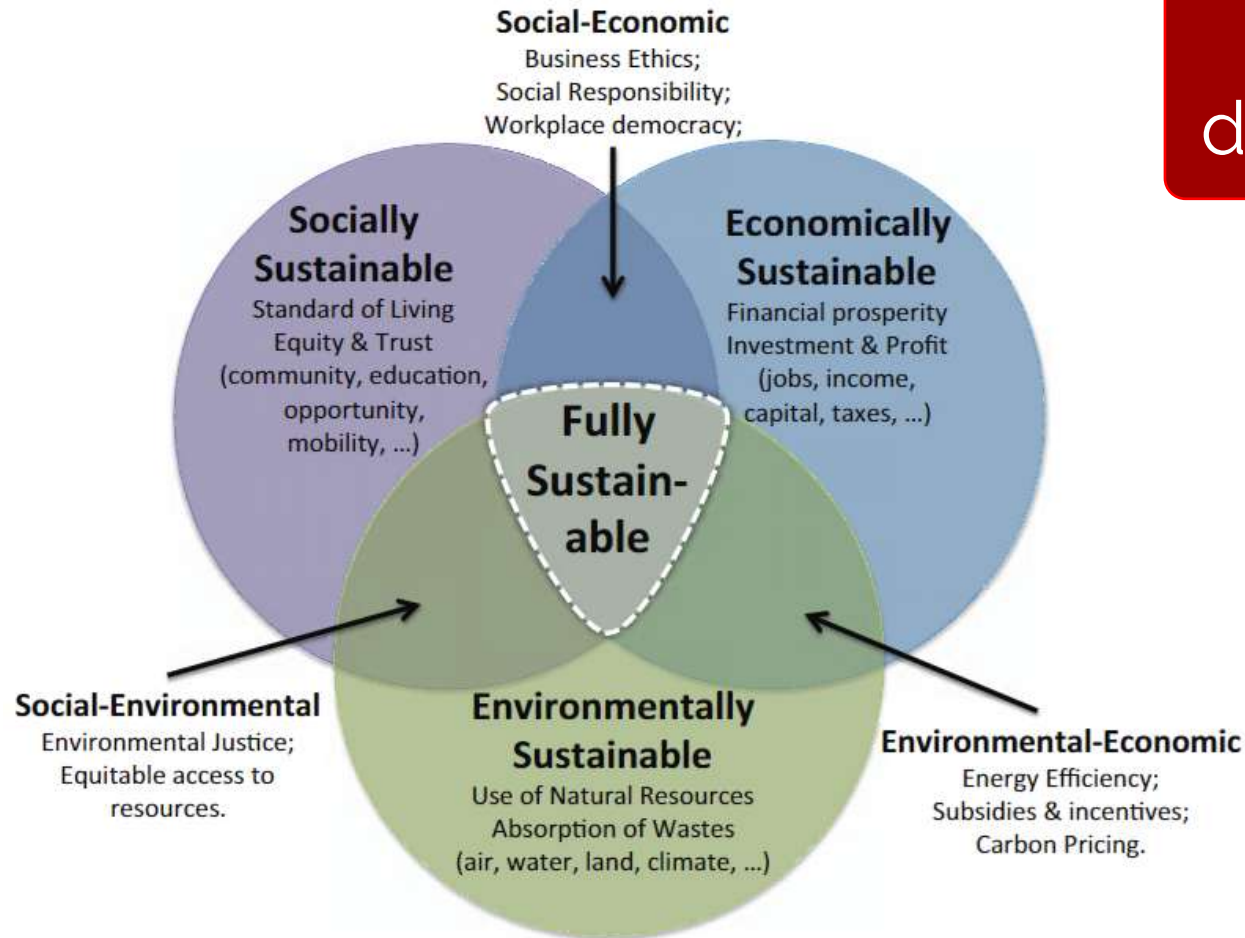


Emissions from in-house data centers of Google, Microsoft, Meta and Apple may be 7.62 times higher than official tally

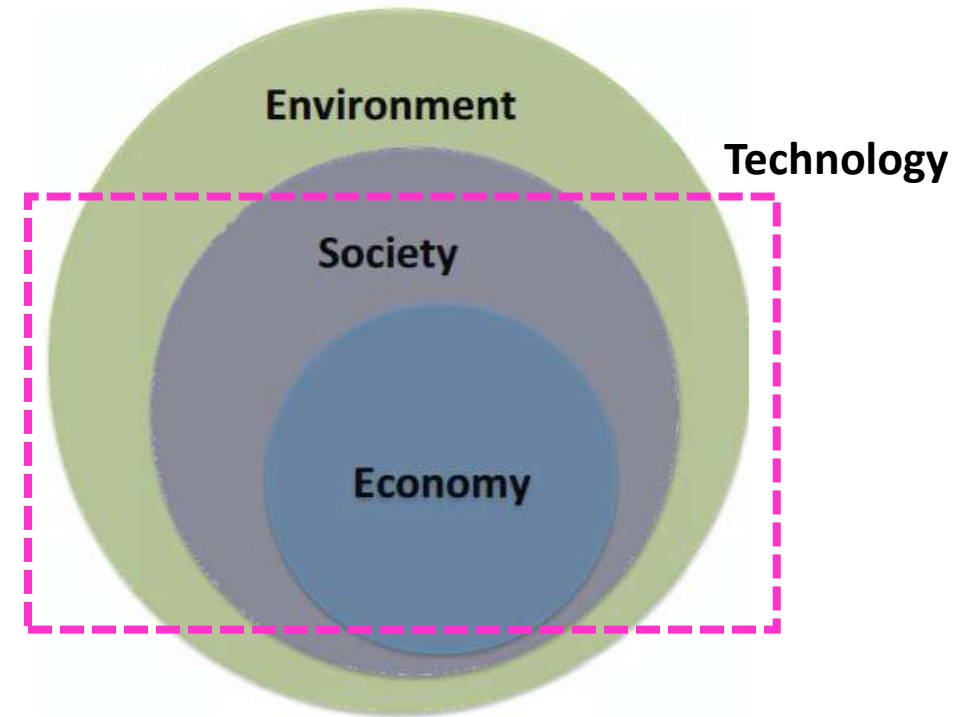


# We need to **model** “sustainability”

Complexity →  
diverse expertise is needed



**Weak notion of sustainability**



**Strong notion of sustainability**

# Participatory and Collaborative Modeling of Sustainable Systems: A Systematic Review

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## ABSTRACT

Sustainability has become a key characteristic of modern systems. Unfortunately, the convoluted nature of sustainability limits its understanding and hinders the design of sustainable systems. Thus, cooperation among a diverse set of stakeholders is paramount to sound sustainability-related decisions. Collaborative modeling has demonstrated benefits in facilitating cooperation between technical experts in engineering problems; but fails to include non-technical stakeholders in the modeling endeavor. In contrast, participatory modeling excels in facilitating high-level modeling among a diverse set of stakeholders, often of non-technical profiles; but fails to generate actionable engineering models. To instigate a convergence between the two disciplines, we systematically survey the field of collaborative and participatory modeling for sustainable systems. By analyzing 24 primary studies (published until June 2024), we identify common challenges, cooperation models, modeling formalisms and tools; and recommend future avenues of research.

## CCS CONCEPTS

• General and reference → Surveys and overviews; • Social and professional topics → Sustainability.

## KEYWORDS

collaboration, MDE, model-driven, model-based, participatory modeling, survey, sustainability, systematic literature review

## ACM Reference Format:

Rajitha Manellanga and Istvan David. 2024. Participatory and Collaborative Modeling of Sustainable Systems: A Systematic Review. In *ACM/IEEE 27th International Conference on Model Driven Engineering Languages and Systems (MODELS Companion '24)*, September 22–27, 2024, Linz, Austria. ACM, New York, NY, USA, 10 pages. <https://doi.org/10.1145/3652620.3688557>

## 1 INTRODUCTION

Sustainability is the capacity to endure [57] and preserve a system's functionality over time [52]. Sustainability has become one of the key characteristics and a major concern in modern systems [39]. An apt demonstration of this trend is the position the European Commission takes in identifying sustainability as one of the two central

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<https://doi.org/10.1145/3652620.3688557>

topics for a resilient European industry within the framework of Industry 5.0 [44]. Expert voices are also calling to action in developing more sustainable systems and engineering methods [56, 78].

Unfortunately, design for sustainability is significantly challenged by the stratified and multi-systemic nature of sustainability [39], i.e., having different meanings at different levels of abstraction and having different meanings for stakeholders of different domains. Various forms of *cooperative modeling* offer a treatment for these challenges. Modeling allows for treating the problem of stratified meanings by the mechanisms of multi-abstraction and multi-semantics [80]. As such, the role of modeling in the analysis and design of sustainable systems is clearly recognized [28]. Cooperation allows for treating multiple meanings by involving a diverse set of stakeholders at strategic points of the design process.

In the absence of sufficiently diverse cooperation, complex endeavors inevitably fail. For example, Nutt [64] reports that about half of policy decisions fail to achieve the desired results as ignored stakeholder knowledge and interests lead to erroneous decision-making. In response to the need for a diverse involvement of stakeholders, *participatory modeling* [51] facilitates a high-level modeling approach, e.g., through systems dynamics [63], in which non-experts and non-technical stakeholders can be part of the decision-making and design process. While the high level of abstraction and informal modeling benefit diversity, they limit the technical depth modeling can achieve, preventing such cooperative endeavors from shifting into an effective *design* phase. The need for combining participatory modeling with a more technical cooperative modeling paradigm for the design of sustainable systems has been clearly articulated before, e.g., by Midgley [60] and Nabavi et al. [63].

*Collaborative modeling* [36, 46] is a prime candidate to become the cooperative modeling approach required in the design of sustainable systems. Collaborative modeling is a method or technique in which multiple stakeholders manage, collaborate, and are aware of each others' work on a set of shared formal models [46]. While the benefits of collaborative modeling in technical problems have been demonstrated in academia and industry alike, state-of-the-art collaborative modeling techniques are severely limited in their human facets and communication aspects [37], forming a serious barrier for non-technical stakeholders to participate in collaborative modeling endeavors. This, in turn, restricts collaborative modeling to technical problems and limits the potential of collaborative modeling to be applied in sustainability decisions.

There is a synergy between participatory and collaborative modeling that can benefit the design of sustainable systems. Collaborative modeling can support the detailed design of sustainable systems, but it needs to become stakeholder-focused and inclusive

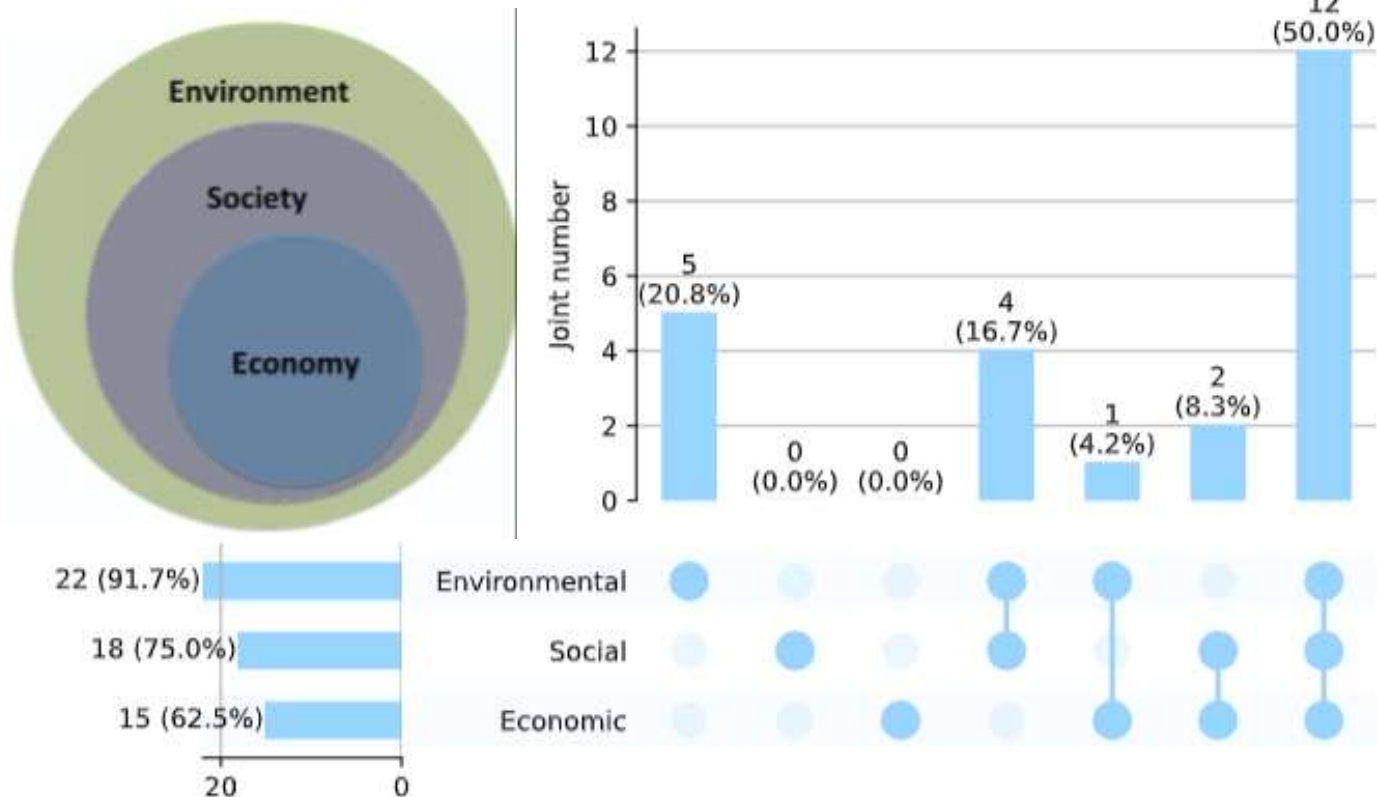
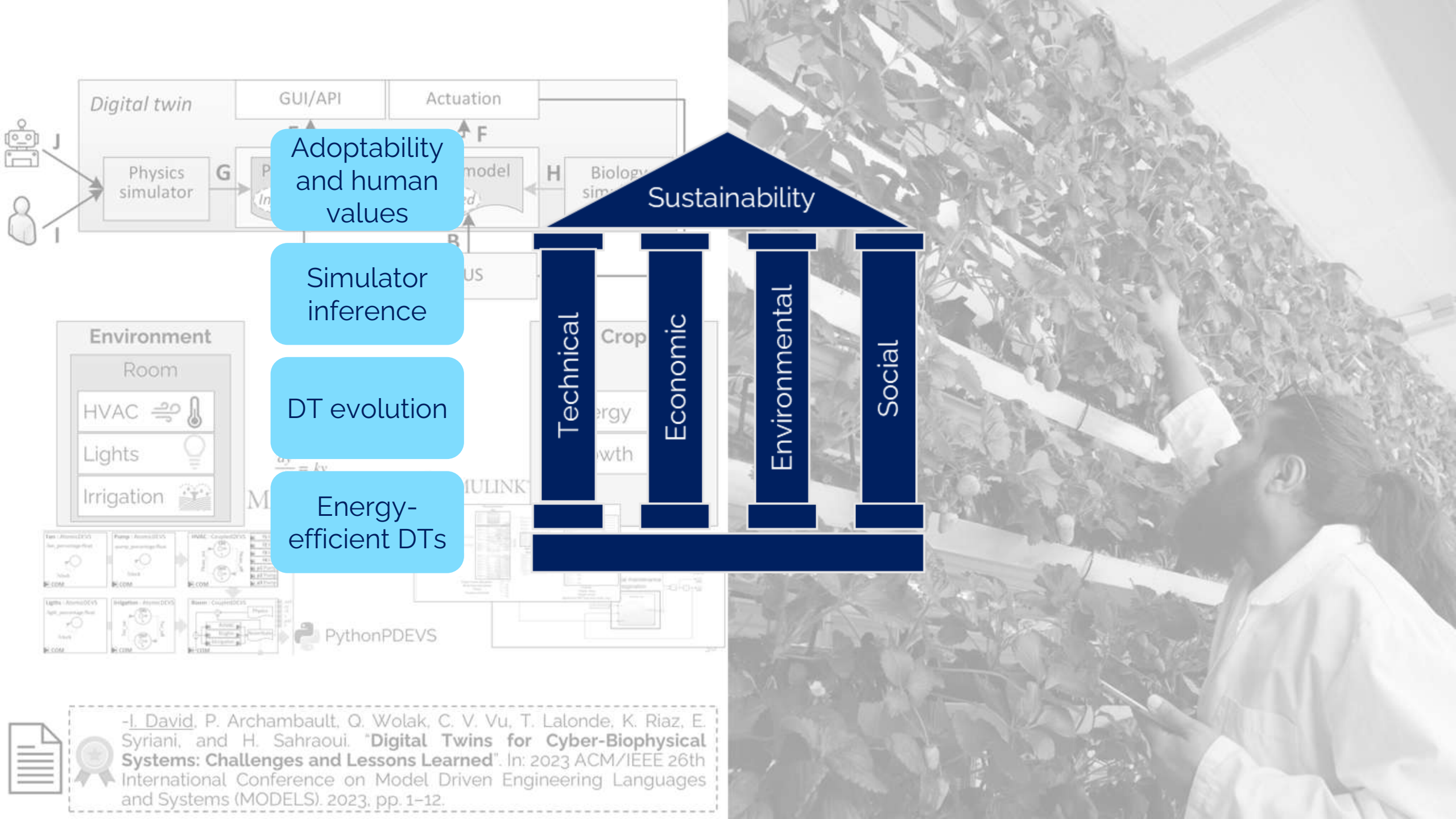


Figure 2: Breakdown of joint sustainability dimensions

So, what about the sustainability *of* digital twins?



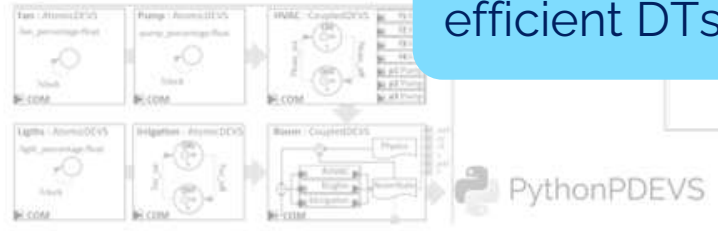
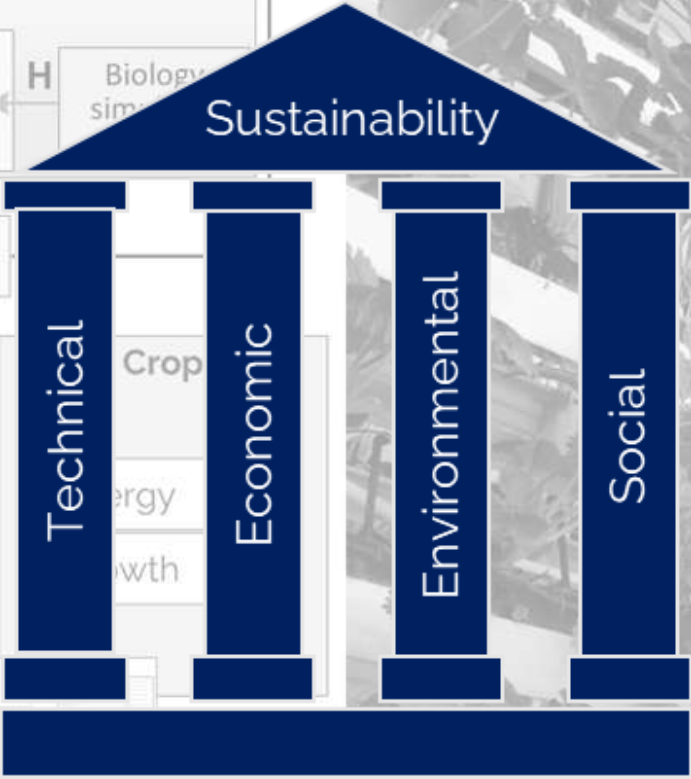
Digital twin    GUI/API    Actuation



Adoptability and human values

Simulator inference

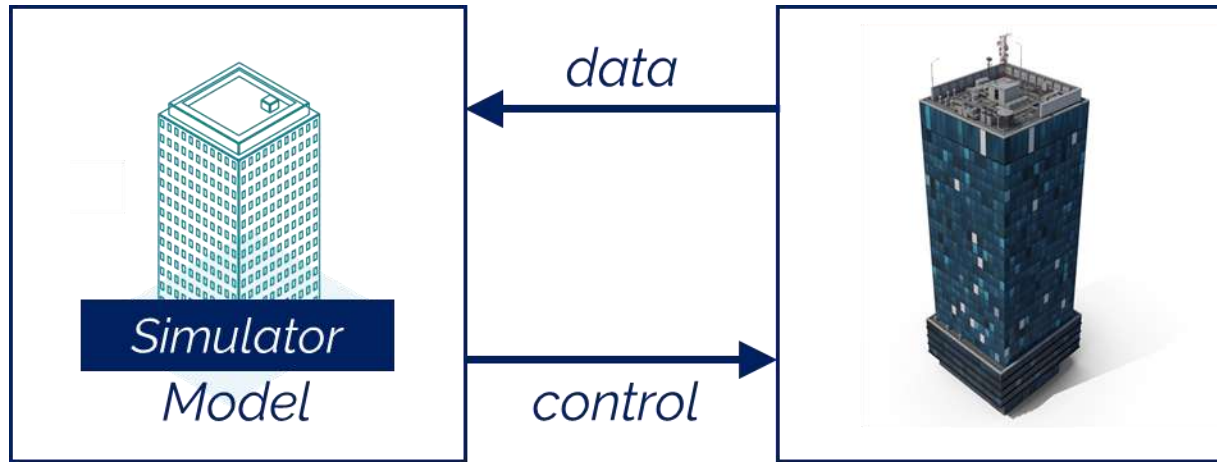
DT evolution

Energy-efficient DTs

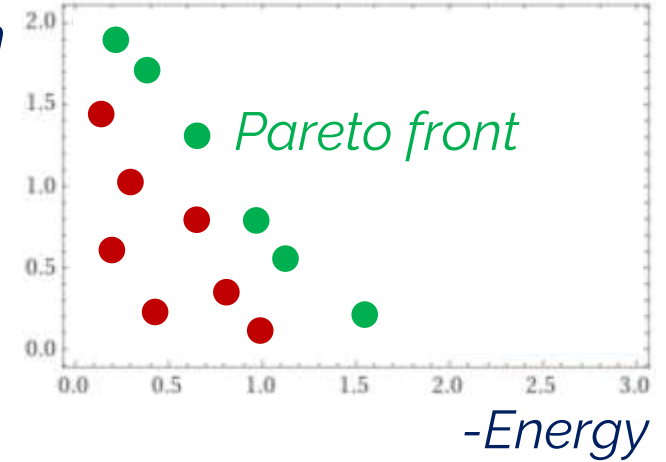


  -J. David, P. Archambault, Q. Wolak, C. V. Vu, T. Lalonde, K. Riaz, E. Syriani, and H. Sahraoui. "Digital Twins for Cyber-Biophysical Systems: Challenges and Lessons Learned". In: 2023 ACM/IEEE 26th International Conference on Model Driven Engineering Languages and Systems (MODELS). 2023, pp. 1–12.

# Energy-efficiency of Digital Twins

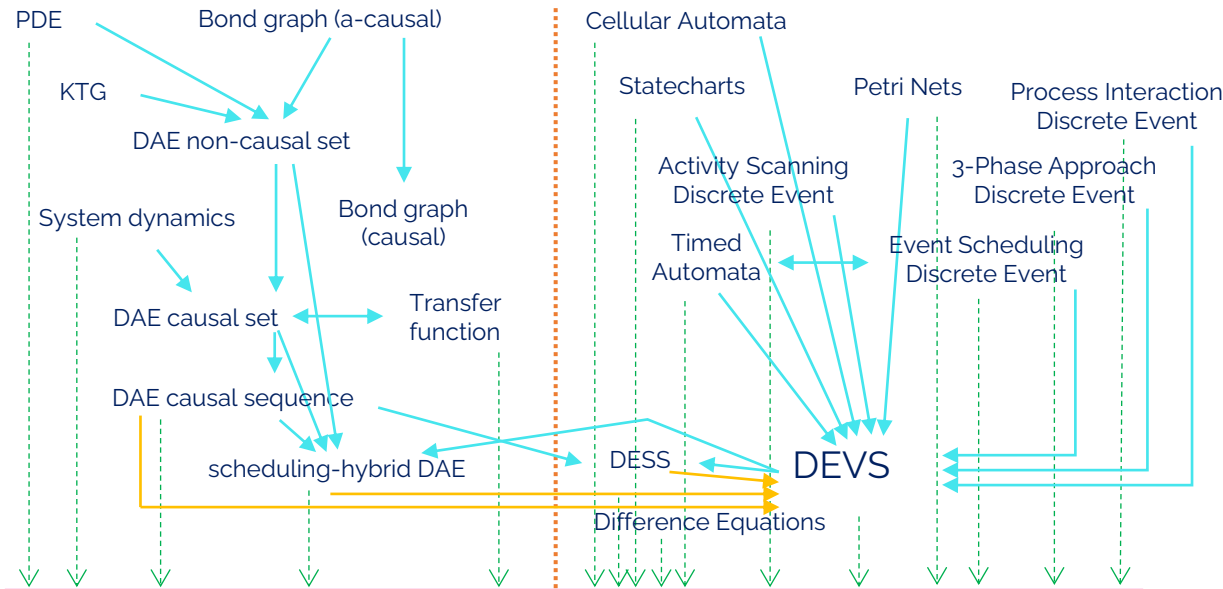


Precision



Digital Twin

Physical Twin



## A Survey of Techniques for Approximate Computing

SPARSH MITTAL, Oak Ridge National Laboratory

Approximate computing trades off computation quality with effort expended, and as rising performance demands confront plateauing resource budgets, approximate computing has become not merely attractive, but even imperative. In this article, we present a survey of techniques for approximate computing (AC). We discuss strategies for finding approximable program portions and monitoring output quality, techniques for using AC in different processing units (e.g., CPU, GPU, and FPGA), processor components, memory technologies, and so forth, as well as programming frameworks for AC. We classify these techniques based on several key characteristics to emphasize their similarities and differences. The aim of this article is to provide insights to researchers into working of AC techniques and inspire more efforts in this area to make AC the mainstream computing approach in future systems.

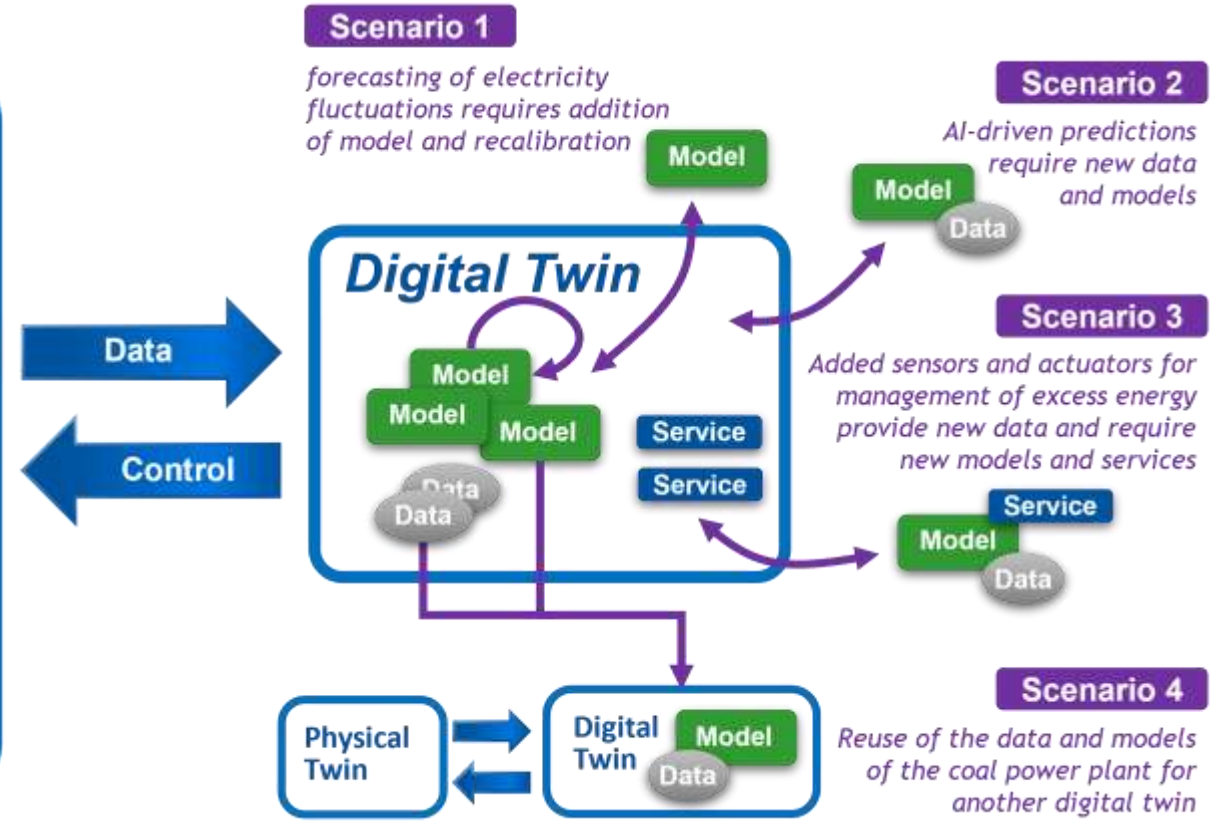
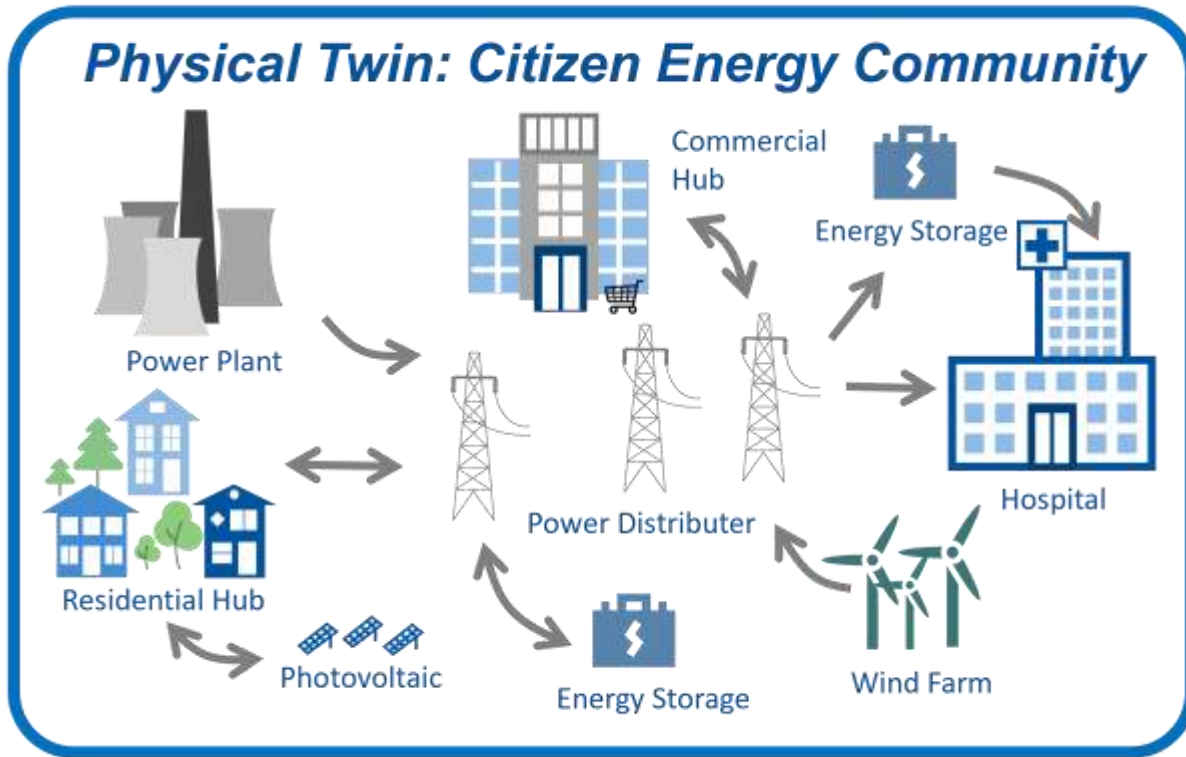
Does not trivially translate to energy savings!

More abstract approximations

State trajectory (observation frame)



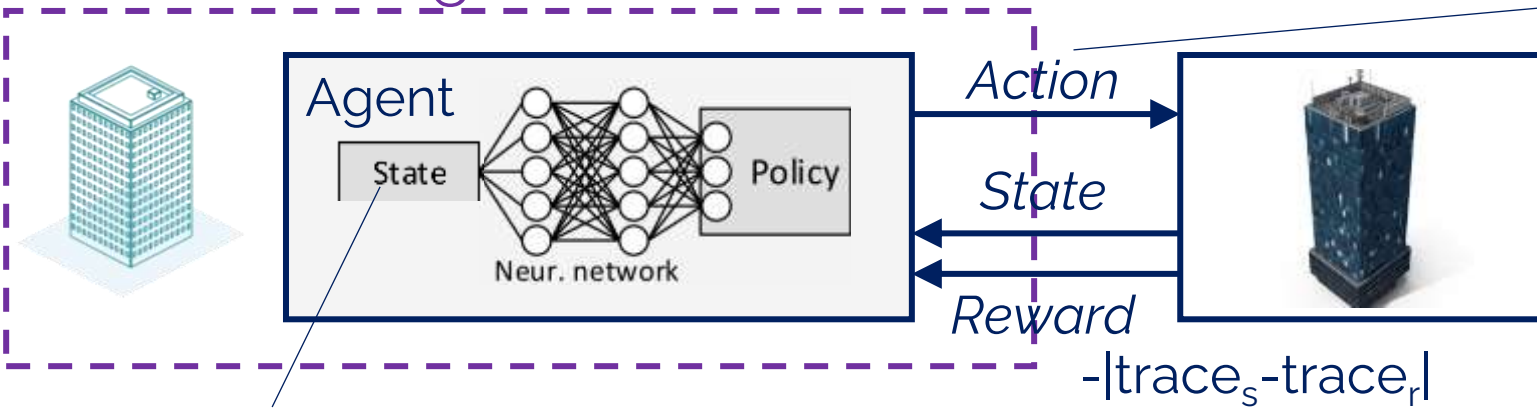
# Evolution of Digital Twins



- Istvan David and Dominik Bork. "Towards a Taxonomy of Digital Twin Evolution for Technical Sustainability". In: MODELS-C 2023 Companion, Vasteras, Sweden. IEEE, 2023
- Judith Michael, Istvan David, and Dominik Bork. "Digital Twin Evolution for Sustainable Smart Ecosystems". In: MODELS-C 2024 Companion, Linz, Austria. ACM, 2024

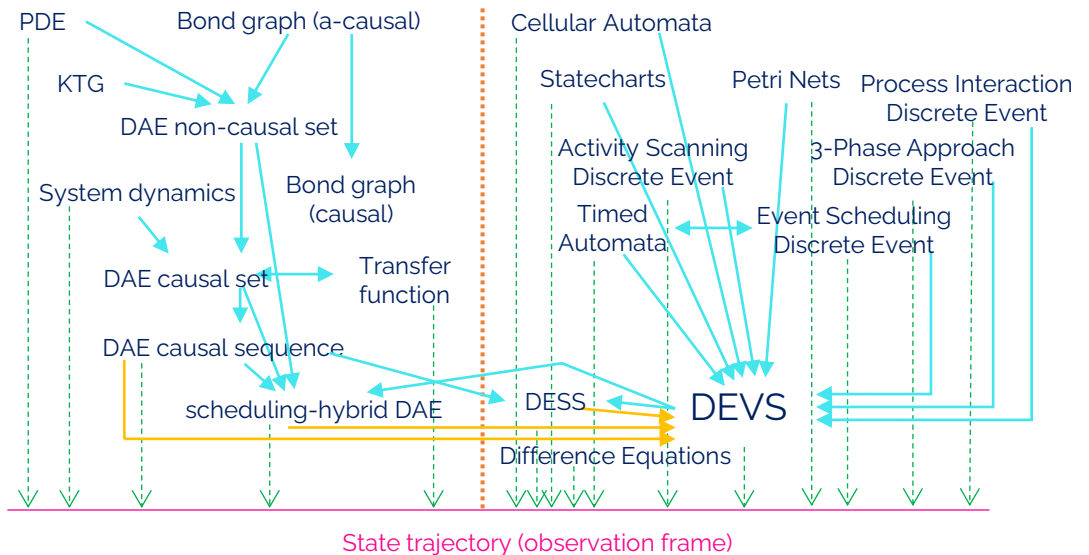
# Rapid development of Digital Twins: Simulator inference by reinforcement learning

## Digital Twin



State actions		
Add DEVS state	$addState(M, s) : M' = M \mid S \cup \{s\}, ta(s) = inf$	(8)
Remove DEVS state	$removeState(M, s) : M' = M \mid S \setminus \{s\}$	(9)
State transition actions		
Add internal transition	$addIntTransition(M, \delta = (s_i, s_j)) : M' = M \mid \Delta_{int} \cup \{\delta\}$	(10)
Remove internal transition	$removeIntTransition(M, \delta \in \Delta_{int}) : M' = M \mid \Delta_{int} \setminus \{\delta\}$	(11)
Add external transition	$addExtTransition(M, \delta = (s_i, s_j)) : M' = M \mid \Delta_{ext} \cup \{\delta\}$	(12)
Remove external transition	$removeExtTransition(M, \delta \in \Delta_{ext}) : M' = M \mid \Delta_{ext} \setminus \{\delta\}$	(13)
Time advance actions		
Set time advance	$updateTa(M, s, t' \in \mathbb{R}_{0,+\infty}^+) : M' = M \mid ta(s) = t'$	(14)
Interacting with other models		
Add output	$addOutput(M, y) : M' = M \mid Y \cup \{y\}$	(15)
Remove output	$removeOutput(M, y \in Y) : M' = M \mid Y \setminus \{y\}$	(16)
Add input	$addInput(M, x) : M' = M \mid X \cup \{x\}$	(17)
Remove input	$removeInput(M, x \in X) : M' = M \mid X \setminus \{x\}$	(18)
Model initialization		
Set initial state	$setInitial(M, s \in S, \theta \leq t \leq ta(s)) : M' = M \mid q_{init} = \{s, e\}$	(19)

a specific DEVS configuration  
(structure and parameters)



- Istvan David and Eugene Syriani. "Automated Inference of Simulators in Digital Twins". In: Handbook of Digital Twins. To appear. CRC Press, 2023. isbn: 978-1-032-54607-0
- Istvan David and Eugene Syriani. "DEVS Model Construction as a Reinforcement Learning Problem". In: 2022 Annual Modeling and Simulation Conference (ANNSIM). IEEE. 2022, pp. 30–41.
- Istvan David, Jessie Galasso, and Eugene Syriani. "Inference of Simulation Models in Digital Twins by Re-inforcement Learning". In: MODELS-C 2021, IEEE, pp. 221–224.

# Social sustainability of Digital Twins



## Individual

Whose decisions are twinned anyways?

- Inclusive partnerships are key in fostering societally sustainability
- Include those who may be affected by the Digital Twins that govern socio-technical systems



## Society

Adoption in lower-income economies?

- Digital solutions that might not be viable in another context
- Variability, product families, validity frames



## Organizations

Who will adopt these solutions?

- Higher-digitalized domains: lack of agility, lack of understanding of benefits
- Lower-digitalized domains: lack of expertise, lack of trust

## VIENNA MANIFESTO ON DIGITAL HUMANISM

VIENNA, MAY 2019

»The system is failing« — stated by the founder of the Web, Tim Berners-Lee — emphasizes that while digitalization opens unprecedented opportunities, it also raises serious concerns: the monopolization of the Web, the rise of extremist opinions and behavior orchestrated by social media, the formation of filter bubbles and echo chambers as islands of disjoint truths, the loss of privacy, and the spread of digital surveillance. Digital technologies are disrupting societies and questioning our understanding of what it means to be human. The stakes are high and the challenge of building a just and democratic society with humans at the center of technological progress needs to be addressed with determination as well as scientific ingenuity. Technological innovation demands social innovation, and social innovation requires broad societal engagement.

**This manifesto is a call to deliberate and to act on current and future technological development.** We encourage our academic communities, as well as industrial leaders, politicians, policy makers, and professional societies all around the globe, to actively participate in policy formation. Our demands are the result of an emerging process that unites scientists and practitioners across fields and topics, brought together by concerns and hopes for the future. We are aware of our joint responsibility for the current situation and the future — both as professionals and citizens.

**Today, we experience the co-evolution of technology and humankind.** The flood of data, algorithms, and computational power is disrupting the very fabric of society by changing human interactions, societal institutions, economies, and political structures. Science and the humanities are not exempt. This disruption simultaneously creates and threatens jobs, produces and destroys wealth, and improves and damages our ecology. It shifts power structures, thereby blurring the human and the machine.

**The quest is for enlightenment and humanism.** The capability to automate human cognitive activities is a revolutionary aspect of computer science / Informatics. For many tasks, machines surpass already what humans can accomplish in speed, precision, and even analytic deduction. The time is right to bring together humanistic ideals with critical thoughts about technological progress. We therefore link this manifesto to the intellectual tradition of humanism and similar movements striving for an enlightened humanity.

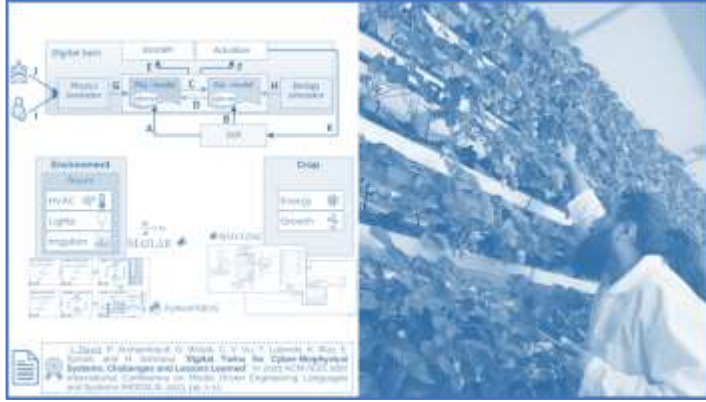
**Like all technologies, digital technologies do not emerge from nowhere.** They are shaped by implicit and explicit choices and thus incorporate a set of values, norms, economic interests, and assumptions about how the world around us is or should be. Many of these choices remain hidden in software programs implementing algorithms that remain invisible. In line with the renowned Vienna Circle and its contributions to modern thinking, we want to espouse critical rational reasoning and the interdisciplinarity needed to shape the future.

**We must shape technologies in accordance with human values and needs, instead of allowing technologies to shape humans.** Our task is not only to rein in the downsides of information and communication technologies, but to encourage human-centered innovation. We call for a Digital Humanism that describes, analyzes, and, most importantly, influences the complex interplay of technology and humankind, for a better society and life, fully respecting universal human rights.

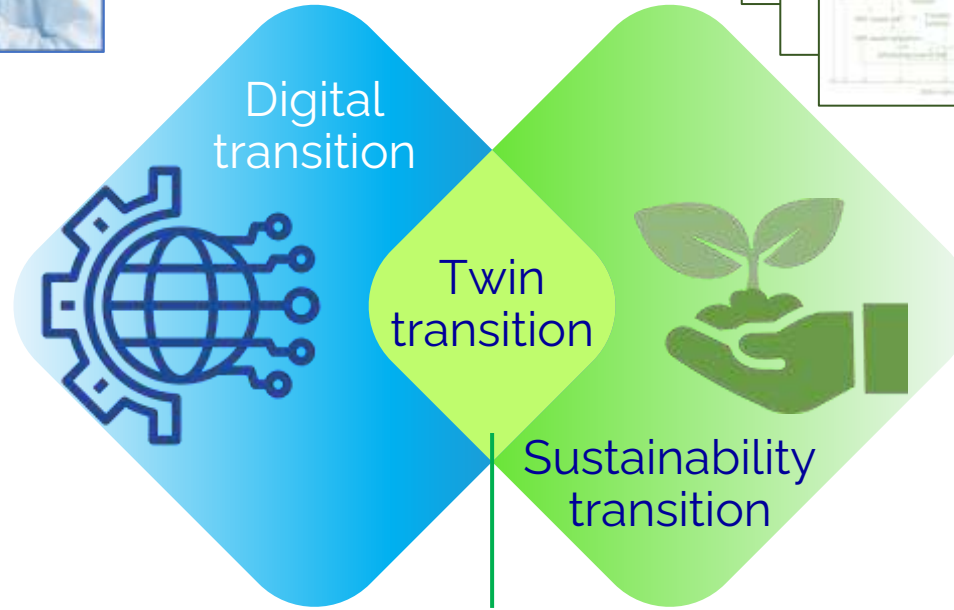
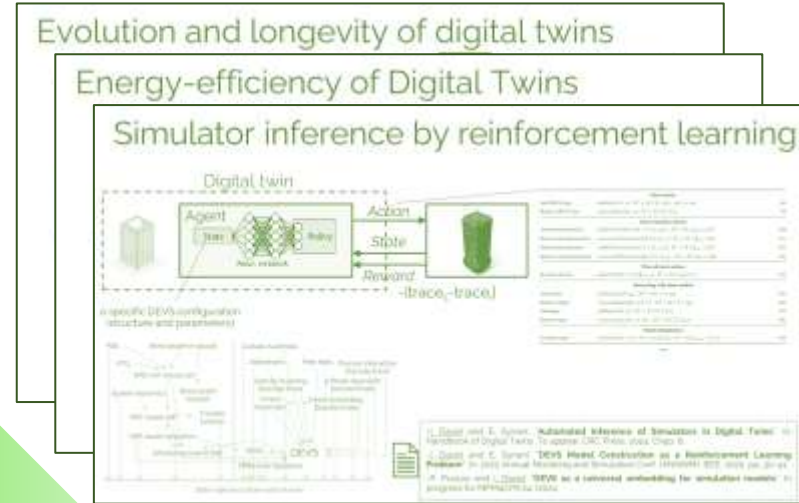
We need to engineer sustainable systems.  
By sustainable methods.

Sure. But how?


# Sustainable systems



# ...by sustainable methods



Sustainability of and by digital tech

 **Sustainable**  
Systems and Methods



Baran Shajari

# Problem: social sustainability is missing from TT

- Social sustainability?
  - Trust in digitalization, transparency, explainability...
- Ignoring social sustainability leads to unwanted consequences
  - Unskilled workforce, fear of AI (e.g., losing jobs)
- HCI as the universal design framework in support of social sustainability of digital tech
  - Human-centered design, trust
  - Accessibility
  - Transparency

## Bridging the Silos of Digitalization and Sustainability by Twin Transition: A Multivocal Literature Review

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\*McMaster University, Canada – {shajari, istvan.david}@mcmaster.ca  
<sup>1</sup>McMaster Centre for Software Certification (McSCert), Canada

**Abstract**—Twin transition is the method of parallel digital and sustainability transitions in a mutually supporting way or, in common terms, “greening of and by IT and data”. Twin transition reacts to the growing problem of unsustainable digitalization, particularly in the ecological sense. Ignoring this problem will eventually limit the digital adeptness of society and the problem-solving capacity of humankind. Information systems engineering must find ways to support twin transition journeys through its substantial body of knowledge, methods, and techniques. To this end, we systematically survey the academic and gray literature on twin transition, clarify key concepts, and derive leads for researchers and practitioners to steer their innovation efforts.

**Index Terms**—digital transformation, multivocal literature review, sustainability

### I. INTRODUCTION

Digital transformation has become an essential tool for companies to improve their operational excellence [39]. Digitalization enables an array of competitive advantages, including enhanced data collection and management, improvements to products and services, and cost reduction.

Unfortunately, the benefits of improved digitalization at the price of increased environmental footprint (PEF). Information and Communications Technology (ICT) currently contributes to about 2-4% of global CO<sub>2</sub> emissions—counting the carbon emissions of the avionics sector—number is projected to increase to about 14% by 2025 due to computation-heavy digital enablers, such as big data. This growth is unsustainable. To follow the rest of the economy, the ICT sector should—directly—decrease its CO<sub>2</sub> emissions by 42% by 2030, by 2040, and 91% by 2050 [34]. Recently, companies become more cognizant of the value of becoming environmentally sustainable, and the ways digitalization can help them do so [4]. While digitalization exerts increasing environmental pressure, it also opens opportunities for understanding, assessing, and enforcing sustainability improvements, e.g., through targeted data collection and process optimization [19]. Sustainability and digitalization seem to be dependent and inextricably linked [24]. This poses challenges for companies that strive to be competitively (environmentally and socially) responsible at the same time. Pursuing such joint innovation agendas requires novel ways to strike a balance between green and digital transition.

Originally suggested in the European Green Deal [36], twin transition is the paradigm of “greening of and by IT& data” [17], i.e., the fostering mutually reinforcing relationships between digital and sustainability transitions. As such, twin transition helps bridge the silos of digitalization and sustainability. While its benefits are clear, the concept of twin transition is not well-understood currently, and supporting methods are in their infancy.

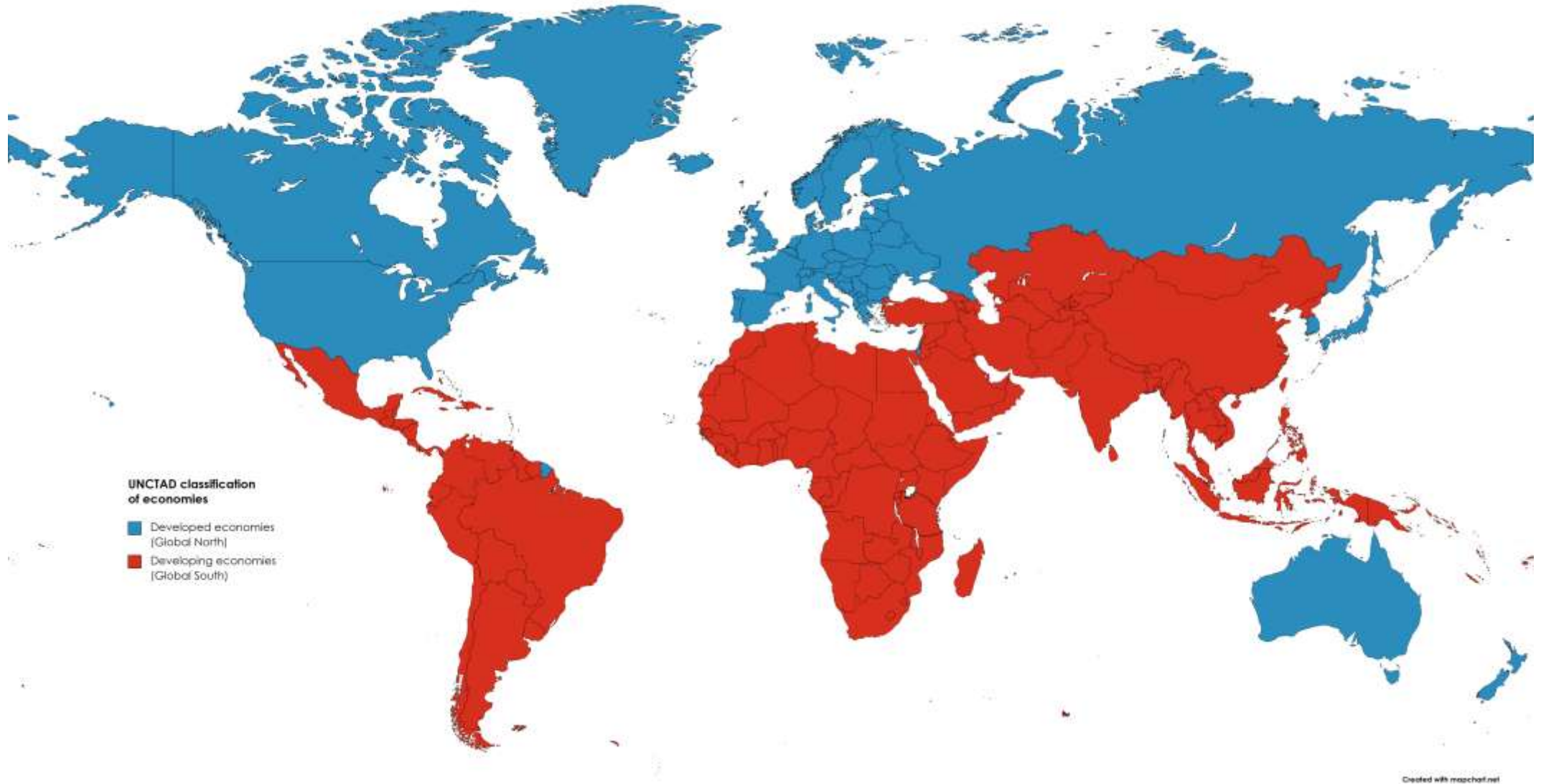
### Contributions

To understand the correct notions, enablers, and challenges of twin transition, we conduct a systematic literature review and elicit actionable leads for information systems researchers. Motivated by the early stage of research and limited academic literature, we opt for a multivocal study [30], i.e., we include non-academic (“gray”) literature in our study, e.g., pre-prints



Baran Shajari

# Societal divides



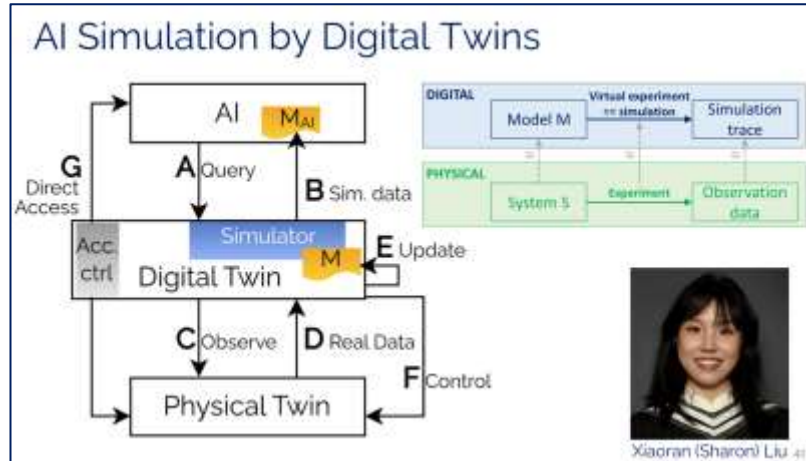
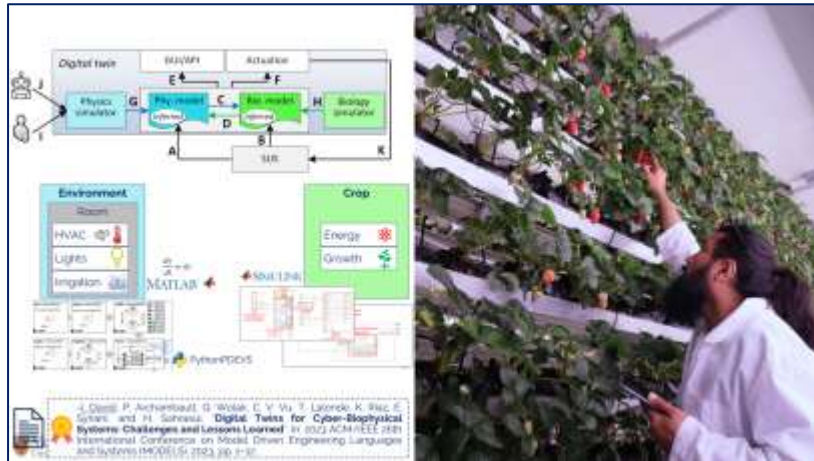
**Digitalization can, in fact, widen the gap between developed and developing countries**



By way of conclusion

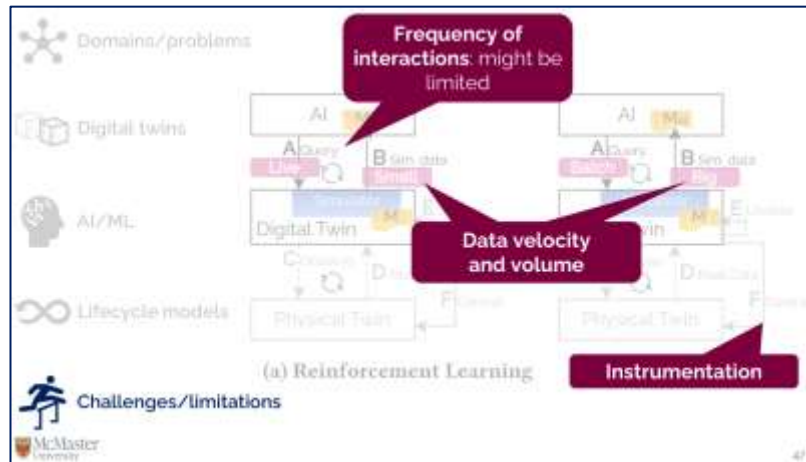
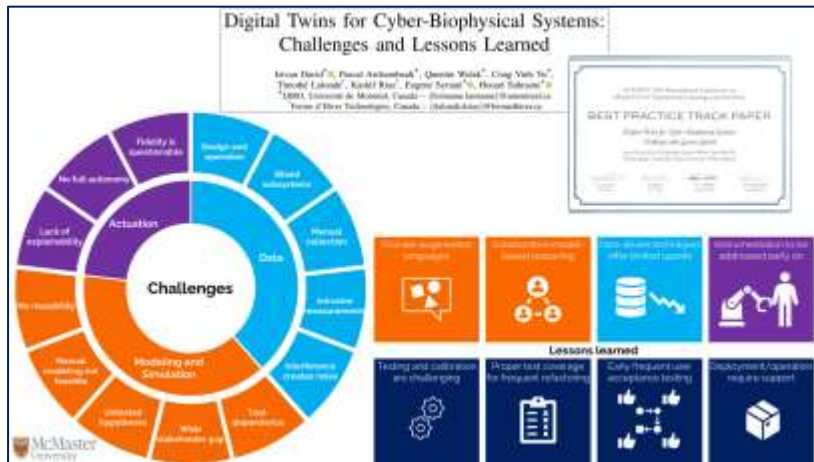


# Engineering complex and sustainable systems through digital twins



### Unsustainable computing and ICT

- Manufacturing**
  - 33% of global energy consumption
  - 36% of global CO2 emissions
- Hardware and e-devices**
  - E-waste is recognized by the World Economic Forum as the fastest-growing category of waste
- ICT**
  - 2-4% of global CO2 emissions (-aviation sector)
  - 2040 prediction: -14%
  - To follow suit, decrease CO2 emissions by -72% by 2040
  - 91% by 2050



### Sustainable systems ...by sustainable methods

**Sustainable Systems and Methods**

**Sustainability of and by digital tech**

**Sustainable Systems and Methods**

Baran Shajari