

Use of Artificial Intelligence and Digital Twins in Building Energy Modeling

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Agenda

- Building and Energy Industry Trends
- Use of Simulation in Buildings and Energy Industry
- Use of AI/ML and Simulation in Buildings and Energy Industry
- Use of Digital Twins in Buildings and Energy Industry

Energy Transition Requires Demand Side and Supply Side Action

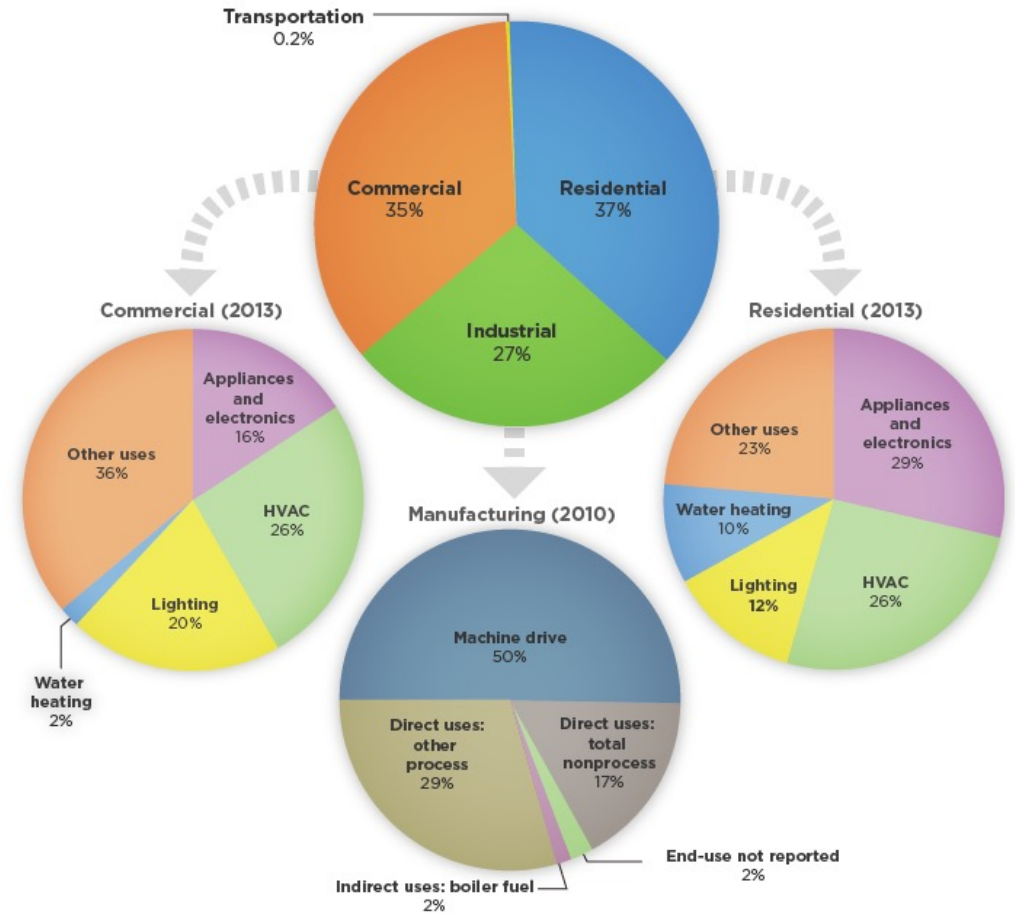
SUPPLY SIDE

- Move away from fossil fuels to renewables
- Wind, solar, hydrogen, fuel cells, microgrids
- Carbon capture

DEMAND SIDE

- More efficient usage; US has energy efficiency of 42%; 58% of energy is wasted
- More efficient energy consumption in HVAC, lighting, appliances

Electricity Consumption by Sector (2013):
Commercial, Industrial, and Residential



Future of Energy and Buildings



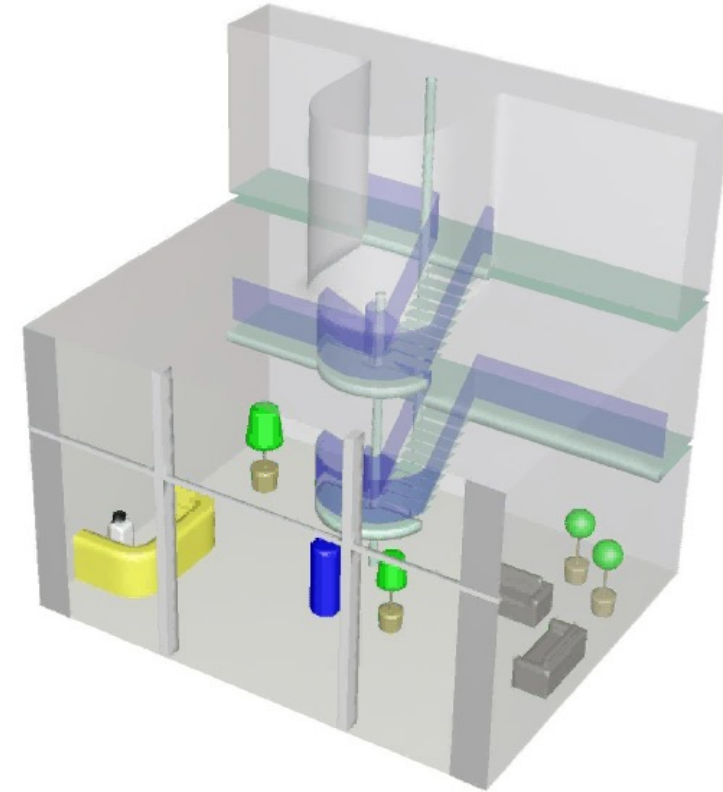
Virtual Building

If

- A building could be modeled with a great accuracy
- Origin of the event reproduced
- Transient fire and smoke behavior described in a reliable way
- Structural behavior of the building predicted with accuracy

Then

- Different scenarios could be studied
- Best solutions identified
- Buildings designed to resist extreme situations better



Fire at the front desk of ANSYS UK in Sheffield. Smoke dispersion

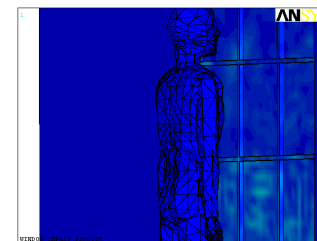
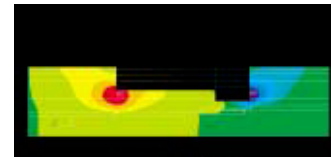
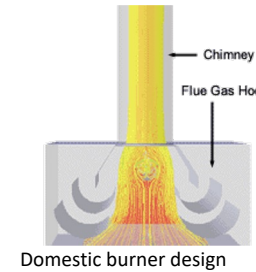
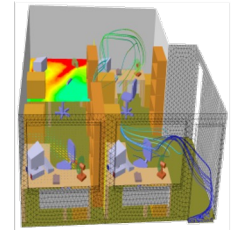
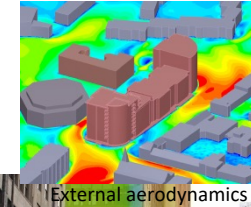


Engineering Simulations

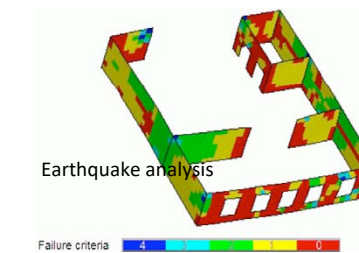


Simulation Driven Building Design

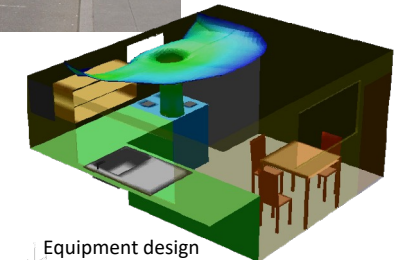
- Designing new, innovative buildings raises numerous challenges
 - Selection and location of safety equipment
 - Behavior under extreme conditions
 - Stress and fatigue of parts
 - Environmental impact
 - Material selection
 - Sustainability
 - Stability



Impact of objects on window

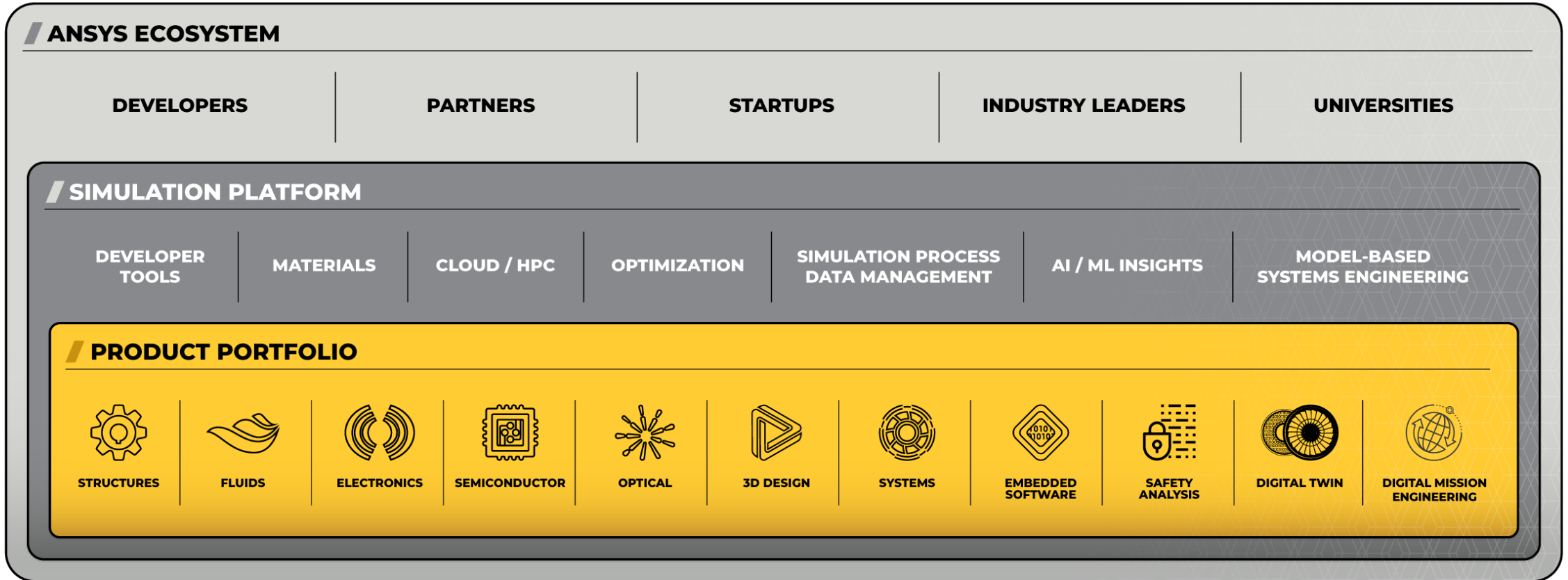


Failure criteria 4 5 6



Depth and Breadth of Ansys Simulation Portfolio

Unique design of the Ansys product portfolio, platform, and ecosystem fits our customer's development processes



ANSYS Simulation Solutions Across Industry Verticals

High-tech & Semiconductor



Bespoke silicon, semiconductor & manufacturing, intelligent connectivity

Automotive & Transportation



Electro-mobility, software defined vehicles, ADAS/autonomy, Vehicle development

Aerospace & Defense



Digital engineering, autonomous systems, safety reliability & cybersecurity, spectrum operations, propulsion systems

Energy & Industrials & Buildings



Efficiency improvements, improving reliability & affordability, scaling & maturing low carbon solutions, accelerate all w/digital transformation

Healthcare



Digitization of healthcare, democratization of simulation, digital evidence for regulatory

SUSTAINABILITY: Clean Environment, Materials and Circularity, Energy Systems, Operational Efficiency



/ Ansys Energy Focused Pillars - Delivering Industry Solutions

1

Efficiency Improvements

- Fuel production
 - Upstream
 - Midstream
 - Downstream
- Energy conversion
 - Electricity
 - Heat
 - Motion
- Operational efficiencies, assets & processes health

2

Improving Reliability & Affordability

- Safe energy production/transmission
- Energy storage solutions
- Integrated energy systems
- Life & performance prediction
- Grid resilience security, stability
- GHG leak mitigation
- Electrification
- New Energy Infrastructure

3

Scaling & Maturing Low Carbon Solutions

- Renewable energy
- Nuclear energy
 - Fusion
 - Fission
- Hydrogen value chain
- Offshore wind
- Carbon capture/storage
- Material circularity

Digital engineering will accelerate all three pillars and drive maturity scaling for new low-carbon tech.

Efficiency Improvements



Efficient Electric Motor

Customer Goals

- Improve Electric motor thermal reliability and life
- Minimize package size

Seamless Physics Coupling

Magnetic Field | Thermal

Motor EM Loss Components | Losses

Oil Spray Cooling

Fired Heater Efficiency | Fuel Burn and Emissions Reduction

Customer Goals

- Improve heaters' efficiency to decrease energy consumption
- Reduce harmful emissions into the atmosphere and meet environmental regulations

Modified Fan Design Shows Less Maldistribution

Maldistribution

Cell-4, Cell-3, Cell-1, Cell-2

Burner-A, Burner-B

Flame Interaction in a Complex Radiant Heater

Gas Turbine Combustion Efficiency

Customer Goal

- Identify operational range with confidence:
- Flashback (even in lean condition in fuel blend)
- Lean Blow-off (LBO) – avoid extinction

Solution

- Effect of flame stretch rate and heat loss for the prediction of premixed flame dynamics
- Accurate laminar flame speed for CH₄/H₂ blends

Benefits

- Control operability loss due to LBO improving efficiency
- Achieve stable operation over full range of engine condition even with fuel flexibility

Ch₄ : No Flashback

H₂ : Flashback

NETL Simulation Validation (SimVal) :
Predict experimental trends to Flashback with increasing concentration of H₂ in fuel

Cambridge Burner : Predict the lean blow off limit with Flamelet Generated Manifold (FGM) for a Premix burner operating under lean conditions

Strained-FGM
LBO @ Equivalence ratio=0.55 (=Exp)

Standard-FGM
LBO @ Equivalence ratio=0.52

2 Improving Reliability & Affordability



Reliable & Safe Hydrogen Storage

Customer Goals

- Design lightweight, safe and reliable composite vessel for hydrogen storage
- Optimize performance while reducing total materials used and autoignition due to sudden release of hydrogen

Solution

- Accurate structural propagation
- High-fidelity chemisorption/refinement
- Material embrittlement
- Efficient artifact algorithms
- Design from data
- Reduced and faster
- Operational hazard

Reliable Battery Aging | Predictive Health Management

Customer Goal

- Create a digital twin of the EV battery running on the cloud that predicts the life degradation due to battery aging, based on real-life field operating history

Solution

- High-fidelity Order Model
- Real-time
- Dynamic implementation by field
- Connect
- Accurate
- Critical prediction

Solar Power: Reliability of Supporting Structures

Customer Goals

- Optimize support structure design for 20-year lifetime of photo-voltaic (PV) based solar panels
- Accelerate testing and certification processes

Solution

- Analyze resulting stresses in support structure at various positions considering aerodynamics, inertial and thermal loads.
- Fatigue as well as single event static (ie, snow, wind gust) and dynamic (ie, modal) failure are explored.
- Fluid-thermal-structural coupling

Benefits

- Minimize support structure materials reduces material and labor costs for installation as well as reduces safety and maintenance costs.

3 Scaling & Maturing Low Carbon Solutions



Biofuel Production From Algae Oil in Raceway Ponds

Customer Goals

- Improve energy efficiency in algae ponds
- Reduce paddle wheel energy required to circulate and grow micro-algae

Original Raceway Design
dP = 44.5 Pa

25% Improvement
dP = 33.4 Pa

Hydrogen Production: Water Electrolysis

Customer Goals

- Achieve highly efficient electrolyzer design
- Ensure better partial load behavior
- Increase the power density of the stacks
- Ensure sufficient electrical conduction to the reaction points and to dissipate the heat generated during the electrolysis

PEM Electrolyzer

PEM Electrolyzer: V_{H_2} and V_{O_2}

Hydrogen fraction in an electrolyzer cell at different flow rates with straight parallel channels.

"These results from Ansys tools will contribute to the future manufacture of better-performing, lower-cost, energy-efficient PEM electrolyzers."

Direct Air Capture

Customer Goals

- Develop and design collector system for capturing, filtering and concentrating CO₂ gas from the atmosphere, to help mitigate global warming
- Vision: scale up to 1Gt/yr of CO₂ removed

Challenge

- Minimize energy need:
 - removing 1t of CO₂ requires filtering ~2M m³ of ambient air → system must be highly optimized, air resistance minimized
- Cost and time to build system prototypes

Solution

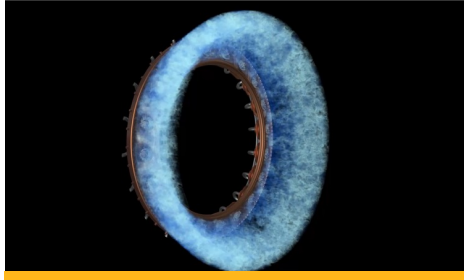
- Modeled 500 geometry variations of the air collector to perfect design

Benefit

- Modeling 5-10x faster than building 1 physical prototype
- Time savings is cost savings for Climeworks, its customers, and for the planet by avoiding the high price of climate change

Simulation Technology Strategy Pillars

Core Physics Numerical Methods and Models



- Solver methods
- Geometry and meshing
- Shape and topology optimization
- Advanced analysis
- Multi-physics
- Multi-scale

High-Performance Computing



- Shared-memory
- Message-passing
- Fine-grained GPUs
- New architectures: FPGAs & AI hardware
- Quantum computing

Artificial Intelligence and Machine Learning



- Solver acceleration
- Solver settings
- Top-down methods
- Bottom-up methods
- Reduced order models
- Generative AI

Cloud, Platform, and User Experience



- Cloud Enabled
- Cloud Native
- Platform, Collaboration
- Open APIs and developer ecosystem
- Common user experience

Digital Engineering



- Model Based System Eng.
- Requirements & architecture Connections
- Safety, security, & software
- Digital twins
- Simulation process & data management
- Mission engineering

Building new pillars of technology on more than 50 years of innovation ...



AI/Machine Learning



AI/ML Approaches to Simulation: *Solver Acceleration and Solver Settings*



Bottom-Up Methods

Organic ML Initiatives

- ML-based fast high-fidelity solvers
- ML-enabled enhancements
 - Starter meshing
 - Sub-modeling
 - User-Experience
 - Solution steering / settings
- ML-services across physics
 - C++ based deployment model



Top-Down Methods

Fast, Accurate, Predictive

- Design Optimization
 - Performance
 - Operation
- Model Calibration
 - MIL-SIL integration
- Virtual Validation
 - Digital certification
- Lifecycle Exploration
 - Digital Twins



Reduced Order Models

ROMs, Twins, Cloud

- 0-3D Reduced Order Models
- Multi-Fidelity models
 - Physics + Data
- Digital Twins
 - Analytical, Data-driven, Hybrid
- Feature extractions from data
- API and UX for ecosystems
 - Partner, customer solutions
 - Cloud services
 - Autonomy and Swarms



Generative AI

Large Language Models

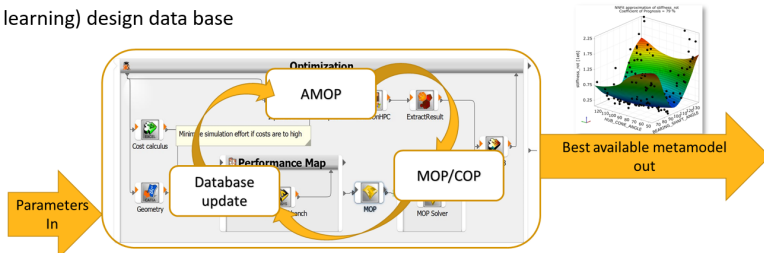
- Generative design and optimizations
- Make solvers easier to use
 - Setup-tuning, AI-Bots:
- Improve customer interactions
 - Virtual agents
 - AnsysGPT

Top-Down: AI/ML Using optiSLang + Ansys Solvers

AI/ML integrated in optiSLang PIDO toolbox and workflows

- MOP (Metamodel of Optimal Prognosis): Workflow for automatic detection and training of best possible metamodel out of large range of surrogate models including
- AI/ML (Neural Network) integrated into metamodel competition = auto ML
- AMOP (Adaptive Metamodel of Optimal Prognosis): Framework for automatic refinement

- Automatic generation of best metamodel training and verification regarding forecast quality of response variation
- Based on the most efficient (self learning) design data base
- Simulation Platform physics and vendor independent
- After Simulation workflow setup democratize as App
- Enables EVERYBODY to use it



Thermal MetaModeling – optiSLang + Icepak

Package Material Level

As-is process/Challenges

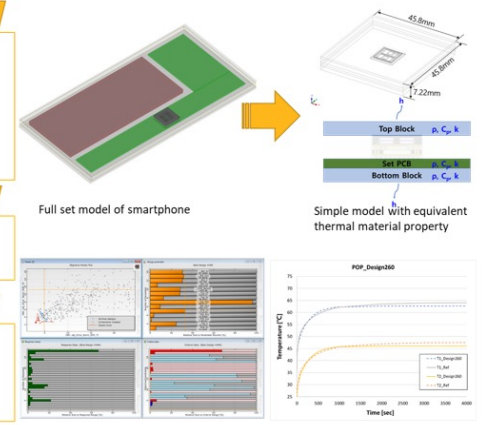
- Sensitivity analysis of thermal material properties of mobile AP
- Fast and Accurate equivalent virtual thermal testing model → Simple Model
- Trial & Error approach for fine tuning material → Expensive!
- Too many trials (1000+) need to be performed for 10+ parameters
- Challenges:
 - Significant manual effort for 1000+ trials
 - Accurate simple model for transient thermal analysis
 - Reduced Dependency on package type

Ansys Value Stream

- Robust workflow integration and optimization with optiSLang-AEDT Icepak
- Reduced input BC conditions and material properties (h,K,CP and Den)
- Sensitivity analysis with thermal material parameter of components.

Outcome

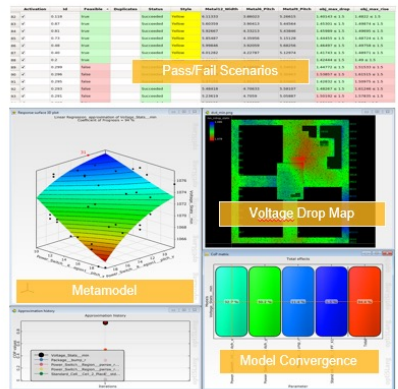
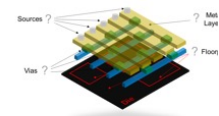
- Extract optimized equivalent properties of Simple model that is well matched with reference data
- Automatic DOE reduction to reduce the overall time for optimization.
- Reduced time for optimization and increased accuracy
 - 2~4 Weeks → 4~5 Days



"Thermal Model Simplification of Mobile Device with Adaptive Metamodel of Optimal Prognosis (AMOP)", V. Krishna, et al., JTherm, 2022

Power Integrity MetaModeling – optiSLang + RHSC

SoC Level



- 1 **Seascope** - Elastic Computing & Database
PDNs prototype generation (Python based), Analysis & Metrics Storage
- 2 **RHSC** - Power Integrity Solvers to generate hierarchical CPMs
Voltage Drop, Electro Migration, Effective Resistance
- 3 **optiSLang** - Parameters generation & Sensitivity Analysis
MetaModel, Pre-processing of parameters, Post-processing of Simulations

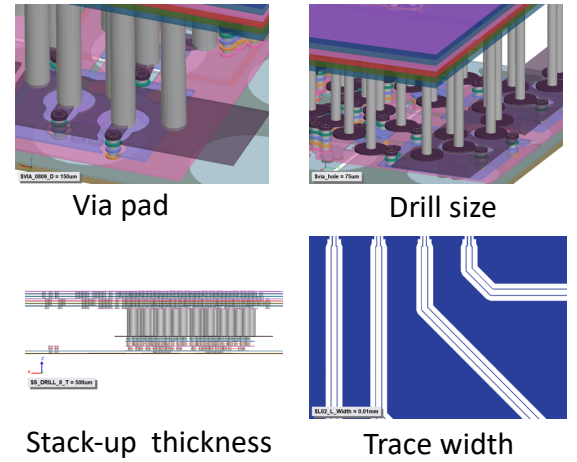
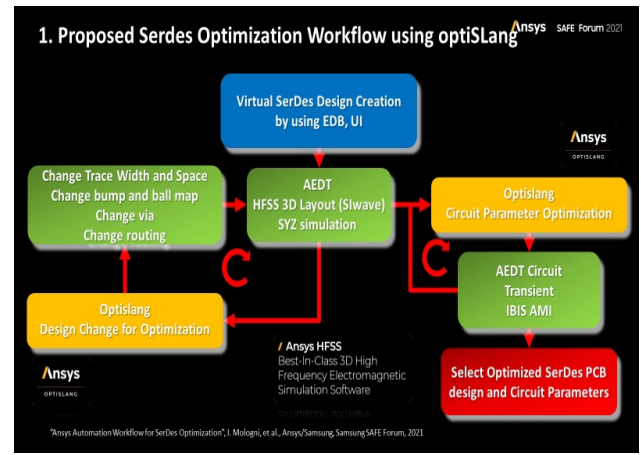
Proof of Concept

- Design Case: 5nm, 80mm2 design, 3 big Macros, 10M Instances
- PDN: 12 Metal Layers and Constant power distribution
- Variables: Metal6 & Metal9 pitch, Metal12 width, PowerSwitches
- Criteria: Static Voltage Drop
- Metamodel: Analytics of 958 PDN scenarios (total runtime: 1Hour)

OptiSLang Power Integrity Parameters Analytics

SoC Power Delivery Networks Exploration & Analytics

EMag MetaModeling – optiSLang + HFSS + AEDT Circuit Simulation



"Ansys Automation Workflow for SerDes Optimization", J. Molegri, et al., Ansys/Samsung, Samsung SAFE Forum, 2021



Bottom Up: ML-based Fluids Solver

Motivation:

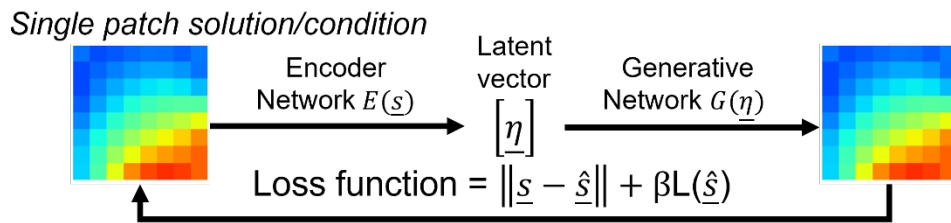
- Geometries and physics have lots of patterns!
- Do we need to solve from scratch?

Key insights:

- Learn lower dimensional representation (η) of solutions on subdomains
- Learn how these patches connect in η space
- Deploy trained networks to solve on entire domains iteratively

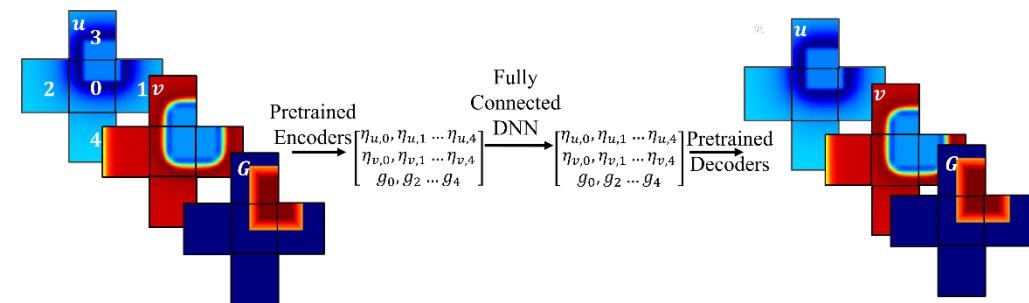
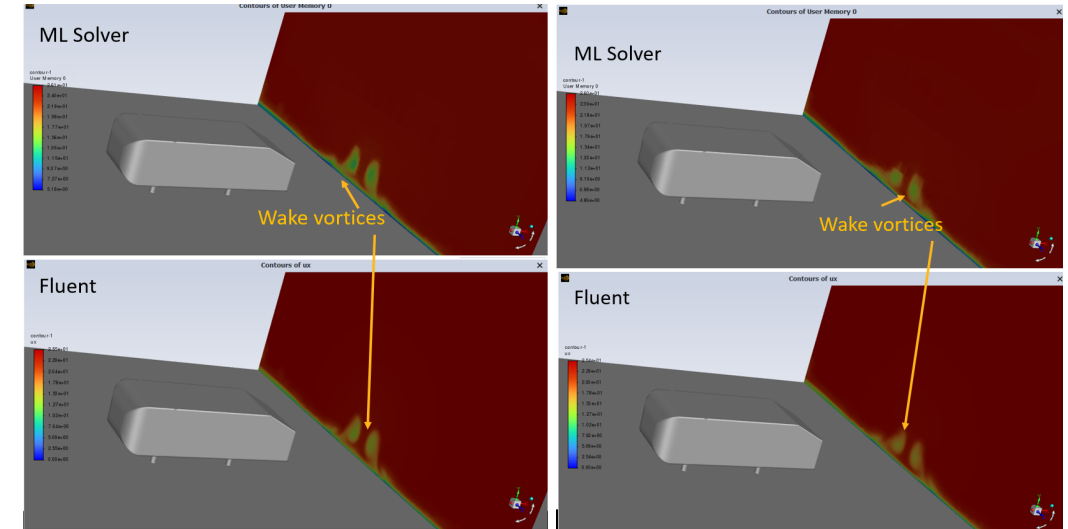
What is new

- Faster solve (100X speedup)
- Lower memory footprint (1/1000 of Fluent)



Learn lower dimensional (η) of solutions on subdomains

ML Solver: Ahmed Body Improved U-Velocity Profiles



Learn how these patches connect in η space.

Reduced Order Models (ROM)

Reduced-order models are critical enablers for data-driven learning & engineering design



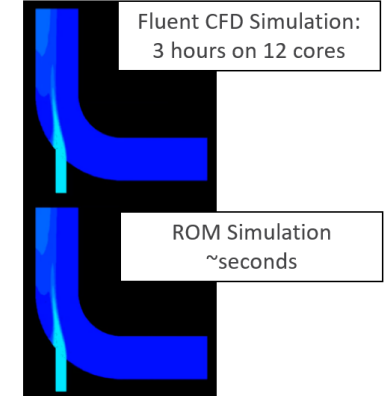
- 1 **Train:** Solve PDEs to generate training data (snapshots)
- 2 **Identify structure:** Compute a low-dimensional basis
- 3 **Reduce:** Project PDE model onto the low-dimensional subspace

ROMs are compact, auto-generated representations of full-3D models that are useful for control design/validation

- ROMs are usually compute and license intensive to create but fast once built
- ROMs have well-defined inputs and outputs (electrical ports, parameters, etc)
- ROMs match the steady-state and/or dynamic responses of the original model, within specified tolerance
- Benefits
 - Reuse: Easily and automatically generate accurate & validated component models
 - Process Compression: Simulate accurate models in 1/10th to 1/100th of the time
 - System Verification and Optimization: Perform rapid design optimization and tradeoff analysis at system level



INCREASED PHYSICS USAGE



ROM Type	Training Data Generation		ROM Extraction		Consumption Format	Ease of Use	Complexity
	Type of Data Required	Source Tools	ROM Technique	Tools			
Response Surface	Parameter Sweep/DoE	All/Any	Lookup table + interpolation	OSL, MC, TB	FMU for Scalars, Twin for Fields	Integrated end-to-end workflow	No states/steady-state only
LT1/LPV/State-space ROMs	Response to step inputs	Mech, CFD and EM	State-space Model with Interpolation	TB, Fluent	FMU for Scalars, Twin for Fields	Mostly integrated. Manual transfers.	Linear or weakly non-linear and stateful
DynaROM	Response to various input waveforms	All/Any	Nonlinear ODE	TB	FMU for Scalars, Twin for Fields	Experience needed to generate training	Non-linear and stateful

Digital Twins

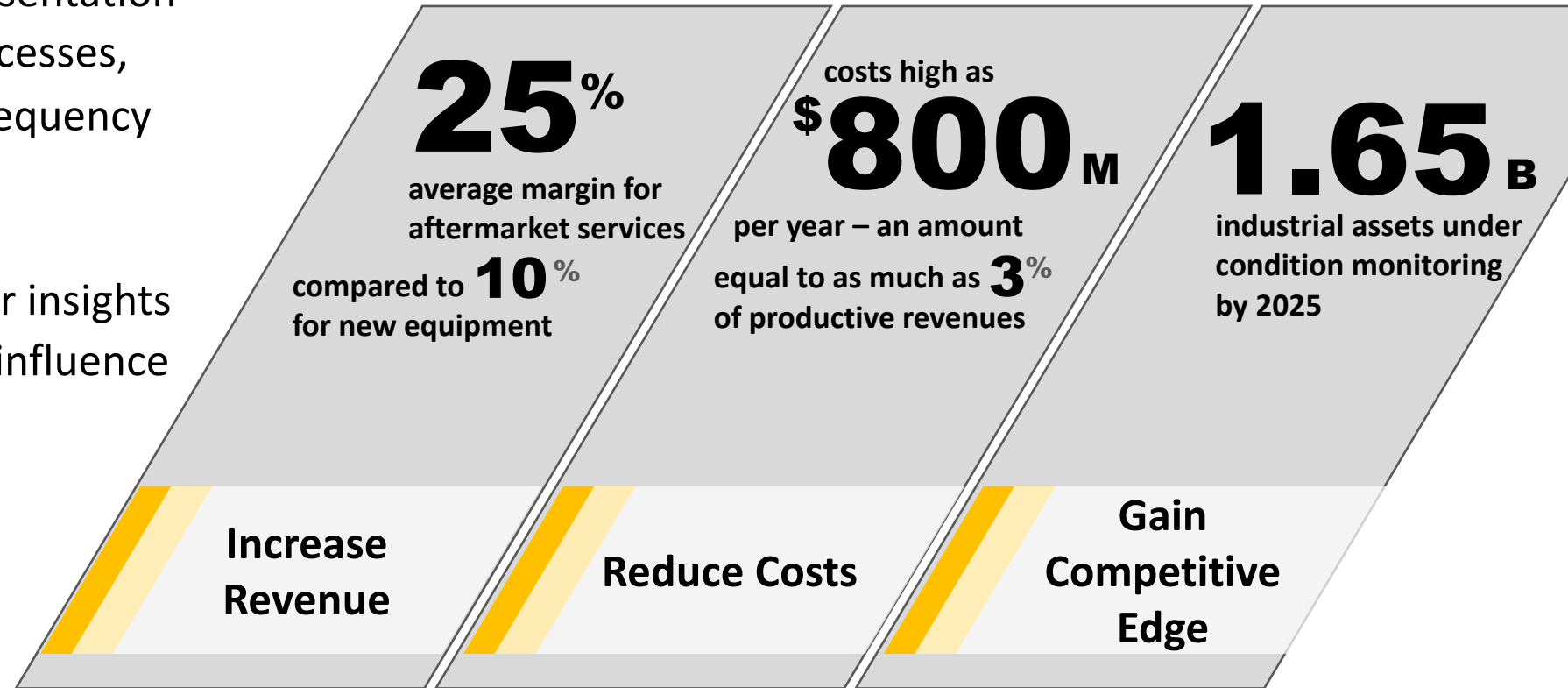


What is a Digital Twin?

Past, Present, Future, Simulate!

digital twin™ : “Virtual representation of real-world entities and processes, synchronized at a specified frequency and fidelity”
CONSORTIUM

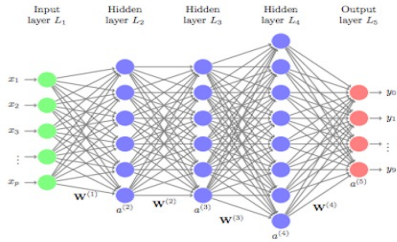
Track the past, provide deeper insights into the present, predict and influence future behavior



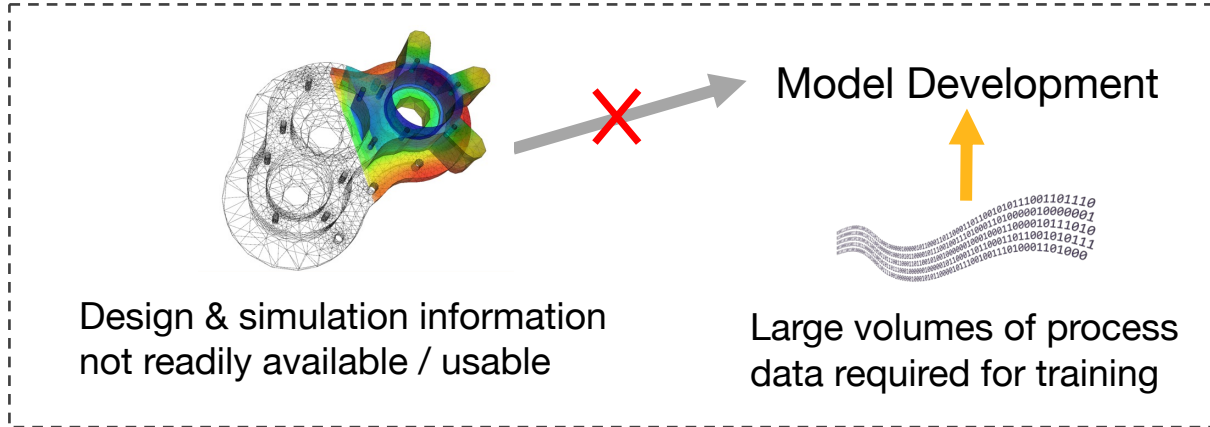
Sources:

1. “Industrial aftermarket services: Growing the core.”, McKinsey.com
2. “Controlling Warranty Costs by Preventing No Fault Found”, WIKA Group
3. Total addressable market (TAM) and compound annual growth rate (CAGR) information throughout presentation is based on third party study completed by Evaluserve Inc. in 2019 commissioned by ANSYS. Study was based on customer and industry expert interviews and review of industry analyst reports and commentaries. Refer to Cautionary Statement for a discussion of factors that could impact future financial results.

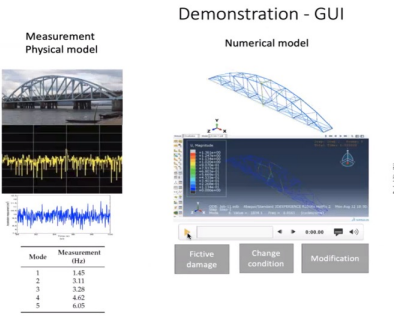
Digital Twin Challenge: Accuracy, Time & Cost



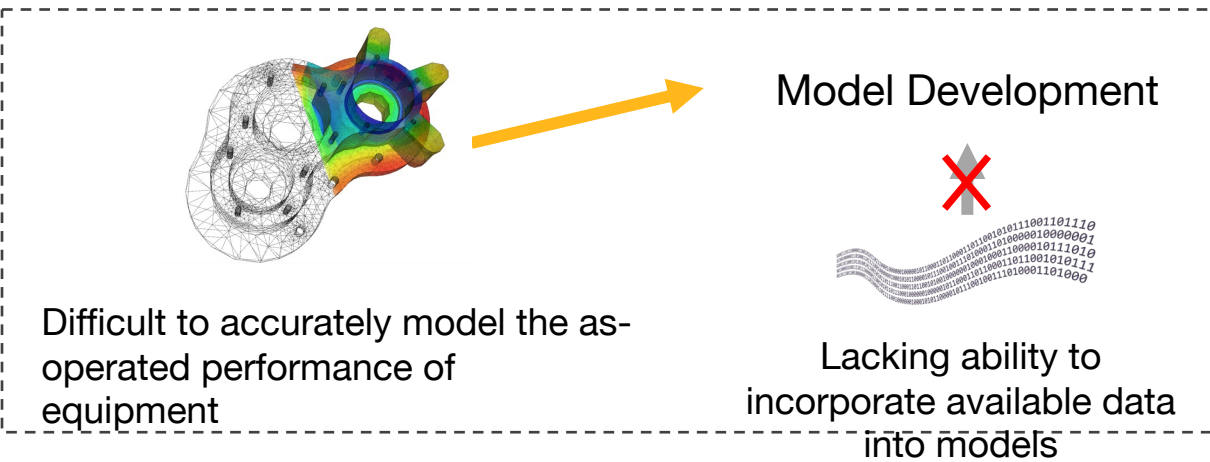
Data-Driven Modeling



Insufficient accuracy, limited by observed data

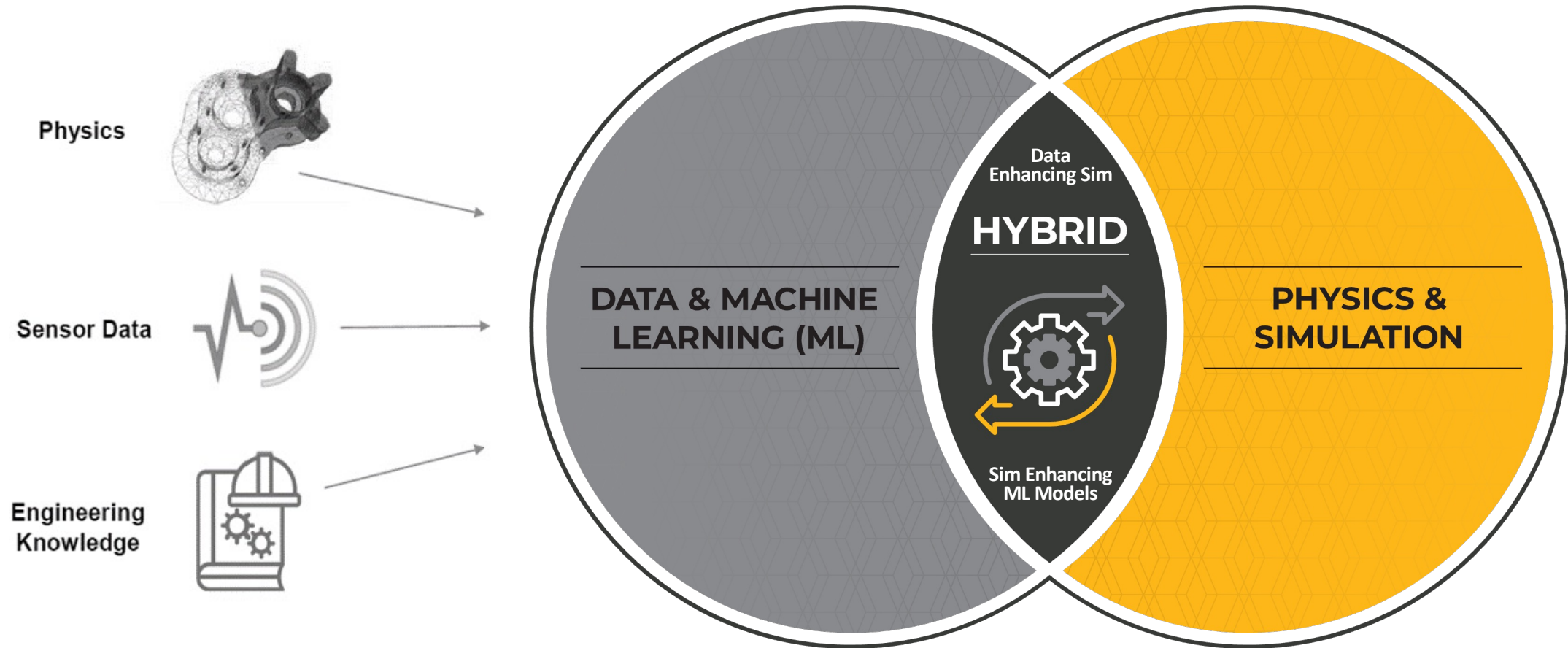


Simulation-Based Modeling



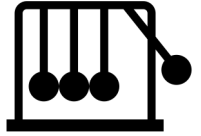
Long, expensive time scales to develop & deploy

Hybrid Digital Twins: Combining simulation and data

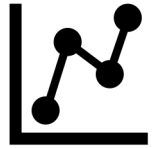


Hybrid Analytics combines data and physics to build Hybrid Digital Twins

Elements of the Digital Twin ecosystem



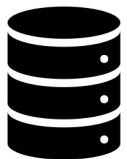
Simulation-Based and Hybrid Analytics



Data-Based Analytics



IoT/Edge Platform



Assets and Infrastructure

Open Ecosystems and
Key Announced Partners

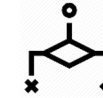


Customers are putting simulation at the center of their Digital Twin implementations

Simulation-Based & Hybrid Analytics



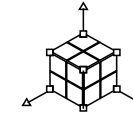
Create virtual sensors to “measure” missing data



Perform what-ifs before applying a solution

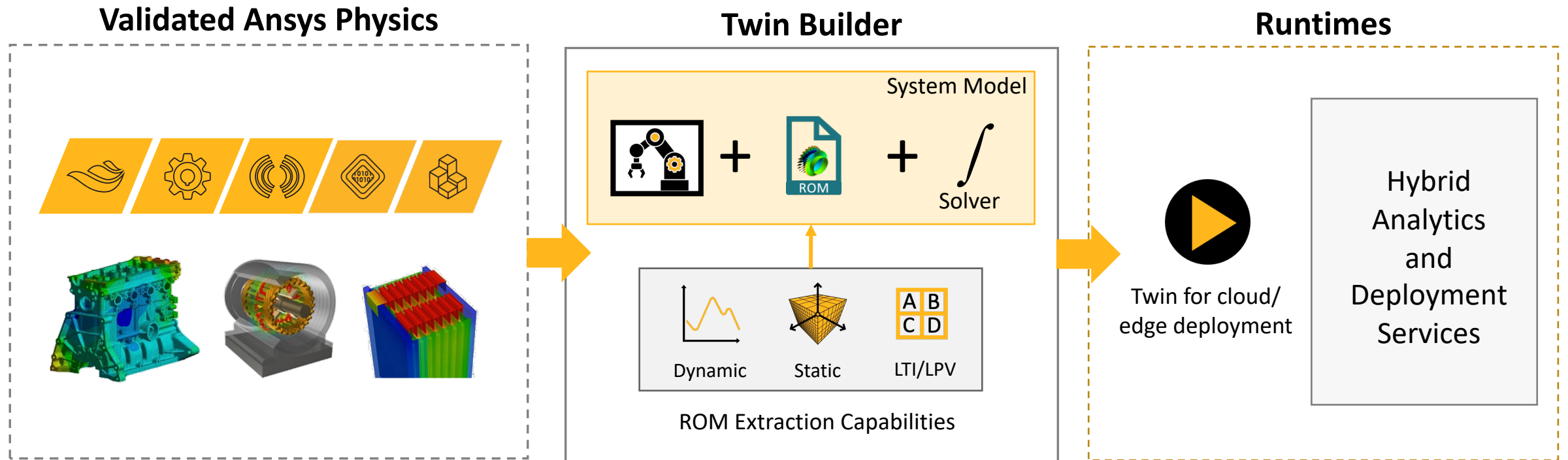


Analyze accurate and deterministic predictions based on physical principles



Explore causality and failure modes using physics

Our solution architecture fits seamlessly into our customers' stack



1. Best in class Reduced Order Modeling capabilities → Reuse
2. Hybrid Calibration → Accurate, evolving models
3. Unique runtime model and open architecture → Scalability



Digital Twin Partners



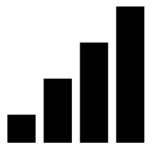
/ Typical use cases for Digital Twins



Virtual Commissioning and System Configuration

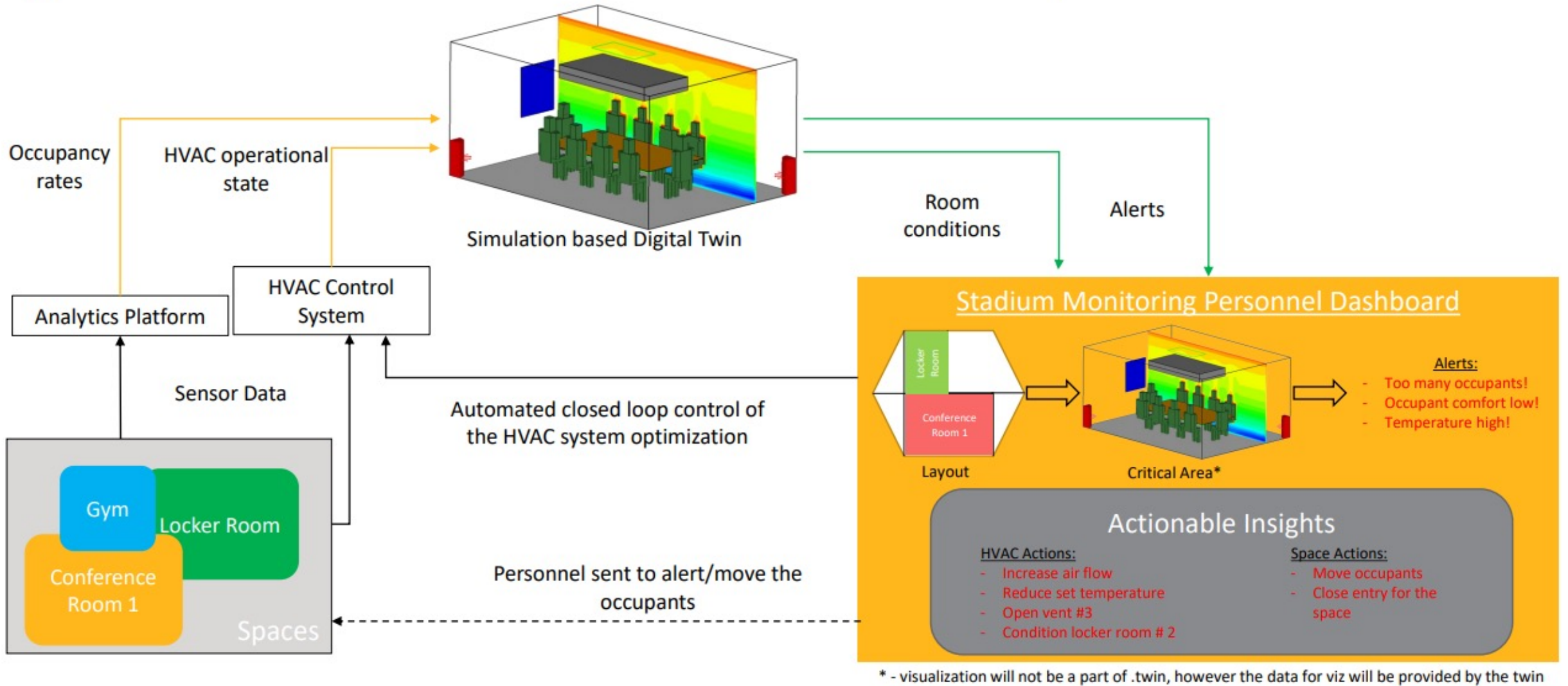


Predictive and Prescriptive Maintenance

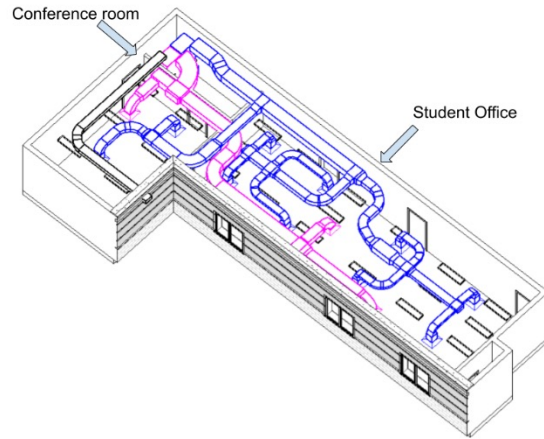


Production Optimization

Optimizing HVAC | Stadium Application with a Real Time Digital Twin

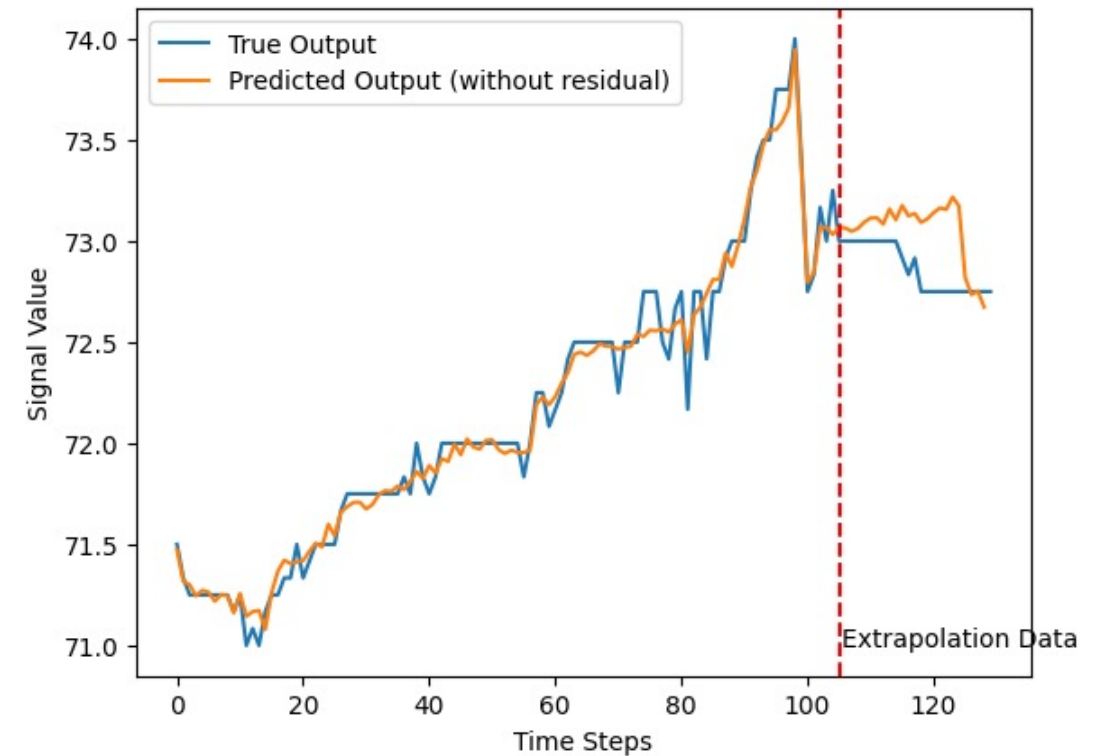


CMU Porter Hall HVAC Digital Twin



Component	Content	Method
Physical Space	Porter Hall office rooms & HVAC systems	/
Digital Space	Information model	Building information model (Revit model)
	Simulation model	Co-simulation model (Modelica & CFD model in Ansys Twin Builder)
Hybrid Analytics model (Python-based ML model in Ansys Twin Deployer)		

Training	Test	Extrapolation	RMSE (test)	RMSE (extrapolation)
1x83x7	1x22x7	1x25x7	0.11	0.24

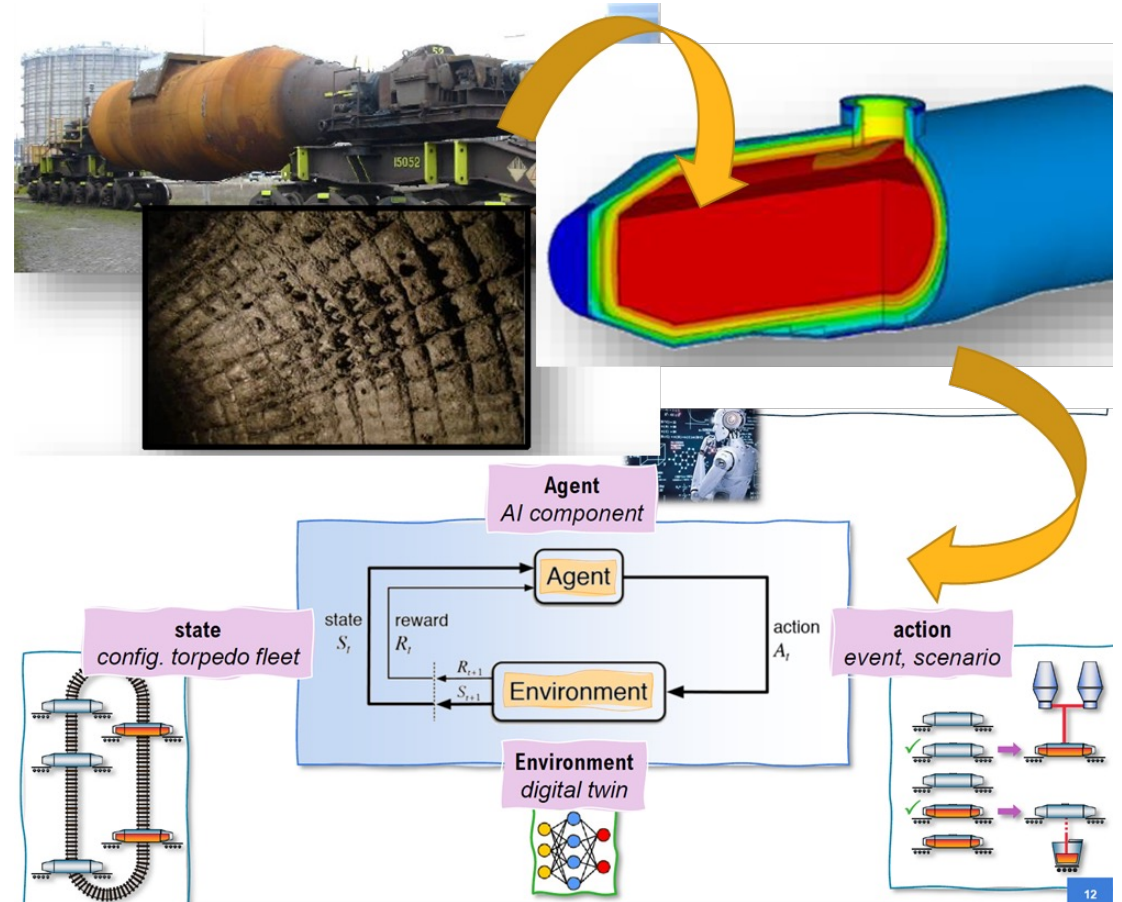


Energy efficiency optimization of iron & steelmaking by digitalization of thermal process management

Challenge: Torpedo refractory maintenance costs exceeded by several M€. Higher hot metal temperatures help with yield losses and CO2 emissions but lead to higher wear of insulation of torpedo car linings and higher energy usage.

Solution: A comprehensive (thermal) digital twin for the entire hot metal (HM) production route. AI based controls to optimize for refractory wear rate.

Result: Savings of ~M€ due to improved maintenance. Additionally, can optimize number of ladles and torpedo cars in use with respect to temperature and select best possible refractory lining



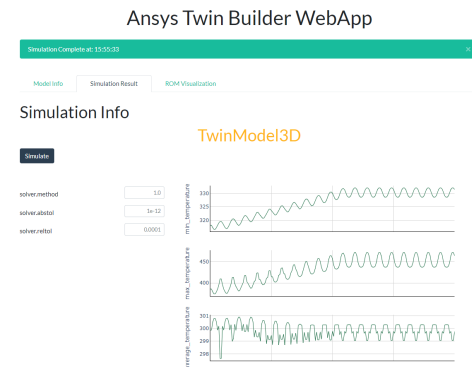
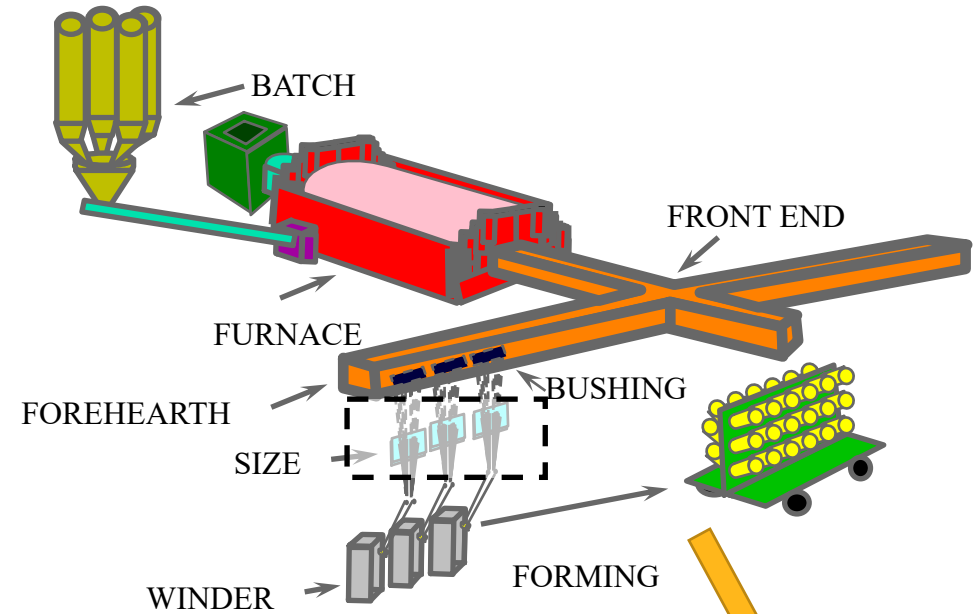
<https://www.ansys.com/webinars/how-digital-twin-is-a-game-changer-for-tata-steel-nederland-to-achieve-their-targets>

Improving production at global glass manufacturer

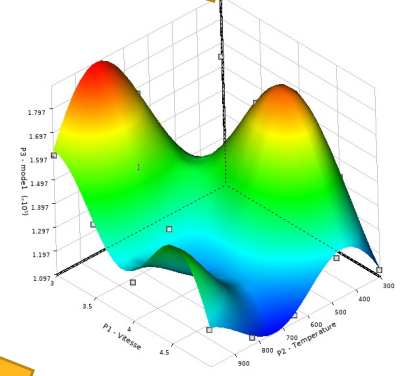
Challenge: For glass fiber manufacturing, consistent temperature (within 2-3 degrees at temperatures in excess of 1400C) in the glass flow path is vital to the quality of the output product. Positioning sensors along the entire flow path is infeasible.

Solution: A reduced order model based digital twin to predict the entire temperature flow field of the forehearth. The reduced order model was created based on available non-linear CFD model and predicts temperatures

Results: Digital twin is deployed on the customer's asset, giving alerts to operators when temperatures are out of bounds. Twin runs in < 5 s, well under the budget allowed for the model execution. Real-time product optimization based on the temperature virtual sensor output in the pilot stage



Operator Dashboard

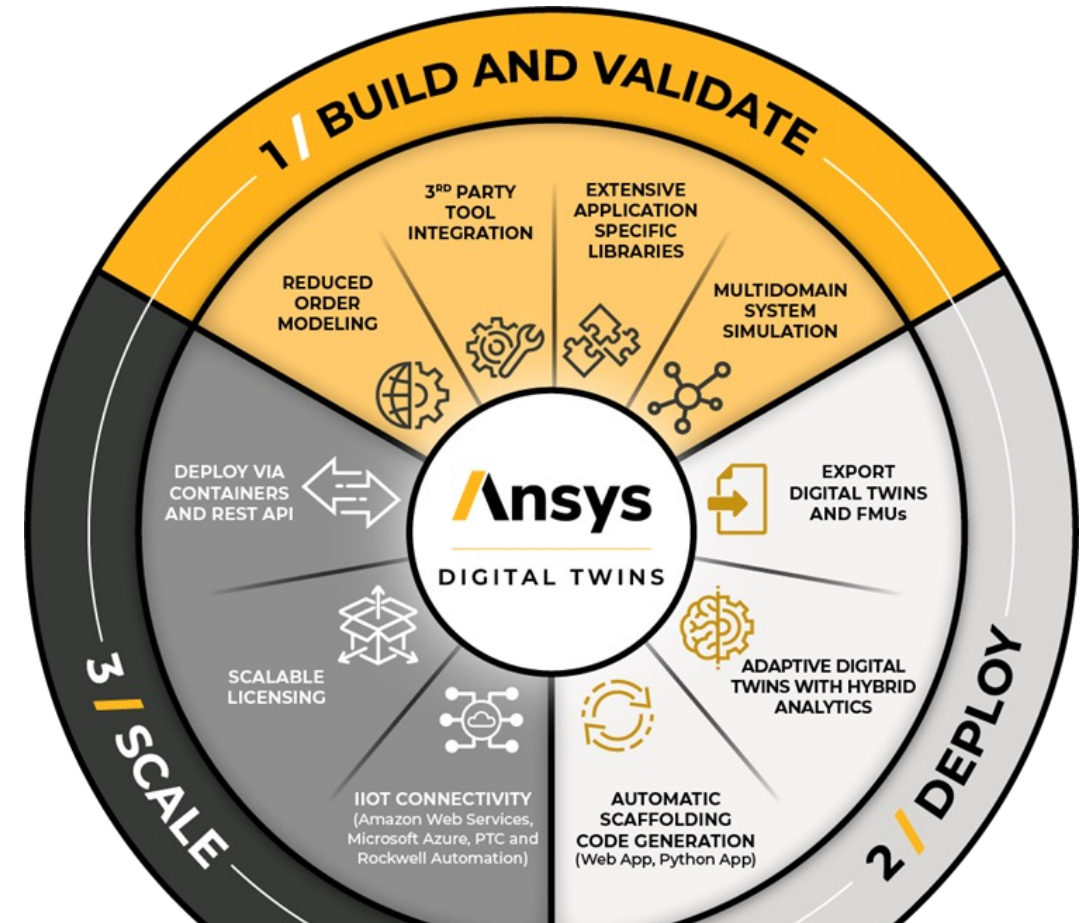


Field ROM



Summary

- Simulation is at the center of virtual prototyping
 - Moving from hardware prototyping and testing to software prototyping and validation and verification
- Simulation allows our customers to grow top-line revenue and bottom-line savings
 - Rapid innovation, lower cycle time, lower risks, increase quality, manage complexity
- Ansys provides the broadest and deepest simulation platform in the industry with the leading physics solvers
- We discussed the latest technology trends for simulation including AI/ML, HPC, Cloud and Digital Twins
- Ansys has a robust Hybrid Digital Twin solution that combines the benefits of physics-based simulation with data-based ML techniques to create accurate, evolving representations of real-world assets.



Let's work together and capitalize on tools that can assist you in your ultimate goals and objectives faster.

/ Summary

- Building and Energy industry face increasing pressures:
 - Cost, time, safety, comfort, environment
- Virtual Building Design offers an attractive alternative
 - Validate proposed designs against regulations
 - Optimize buildings to maximize occupant safety and minimize operating / maintenance cost
 - Communicate and illustrate benefit to protagonists
- Simulation is at the center of virtual prototyping
 - Moving from hardware prototyping and testing to software prototyping and validation and verification
- Simulation allows our customers to grow top-line revenue and bottom-line savings
 - Rapid innovation, lower cycle time, lower risks, increase quality, manage complexity
- Ansys provides the broadest and deepest simulation platform in the industry with the leading physics solvers
- Ansys has a robust Hybrid Digital Twin solution that combines the benefits of physics-based simulation with data-based ML techniques to create accurate, evolving representations of real-world assets
- Simulation is key to Digital Twin implementations, providing critical capabilities such as virtual sensors, what-if analysis and causality and failure mode analysis
- Ansys has demonstrated successful deployment of Digital Twins via several real-world use cases

 **Ansys**

