1st Accelerators Technology Sector Workshop

Engineering Design Tools and Processes
Project Management Methodologies and Tools

Chair: Mike Lamont

Interconnecting knowledge, experience, methods, people & data to foster learning & collaboration

CERN | ATS Accelerators and Technology Sector

Developing digital twins for accelerator magnets

Melvin Liebsch

- on behalf of the TE-MSC-TM section -

CERN

ATS
Accelerators and
Technology Sector

Agenda

Learning from data through the lens of models is a way to exploit structure in an otherwise intractable problem [6]

- The digital twin of the accelerator magnet
- Hybrid modeling (Newton to Kepler)
  - Calibrated models
  - Delta models
  - Machine learning models
- Integration in the life cycle management
- Conclusion

[6] K. Wilcox, Predictive data science for physical systems – from model reduction to scientific machine learning. ICIAM – international congress on industrial and applied mathematics. 2019

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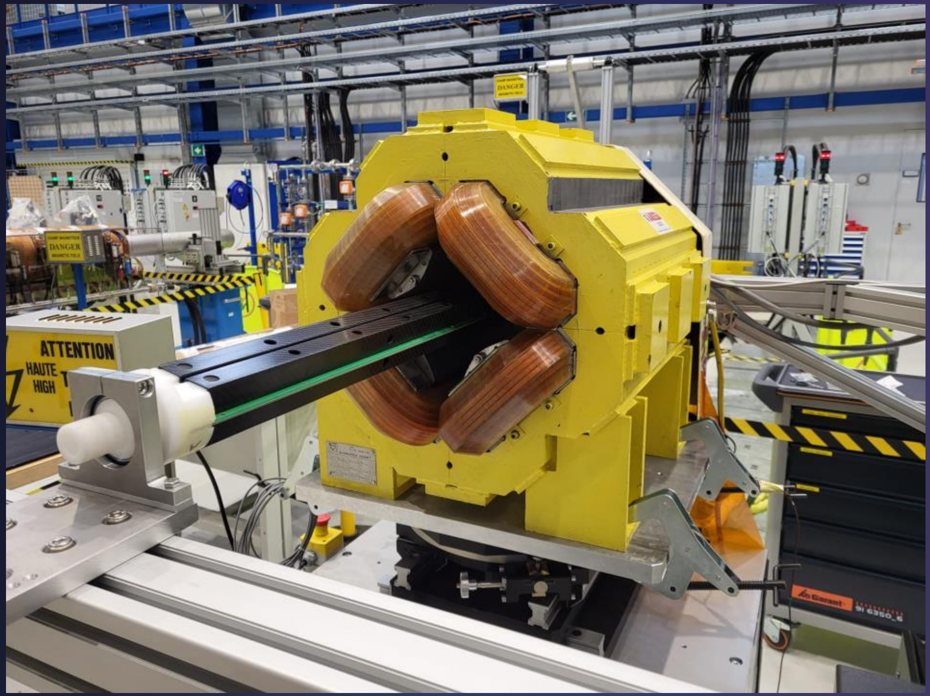
The digital twin [1]

Physical object
a.k.a. accelerator magnet

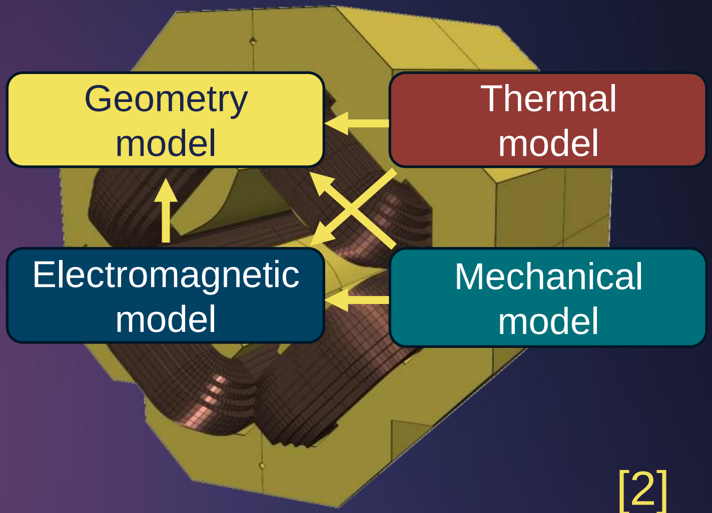


Communication channel

Data
Measurements and inputs
→

Information
Prediction
State observation
←

Digital representation
a.k.a. numerical system model



[2]

→ ***The digital twin requires multi-physical modeling***

[1] Piascik, R., et al., *Technology Area 12: Materials, Structures, Mechanical Systems, and Manufacturing Road Map*. 2010, NASA Office of Chief Technologist.

[2] M. Maciejewski, B. Auchmann, D. M. Araujo, G. Vallone, J. Leuthold and J. Smajic, "Model-Based System Engineering Framework for Superconducting Accelerator Magnet Design," in IEEE Transactions on Applied Superconductivity, vol. 33, no. 5, pp. 1-5, Aug. 2023, Art no. 4003105, doi: 10.1109/TASC.2023.3249647

Accelerator magnet life cycle

Magnet operation

- State observation
- Prediction
- Maintenance

Virtual prototyping

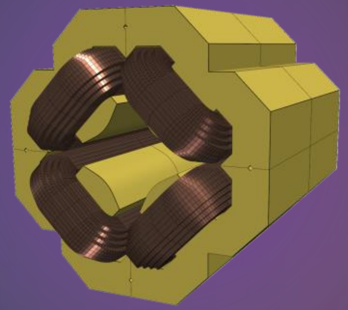
- Coupled multi-physics optimization
- Integration in virtual environment
- Multiscale modeling

Specs

Meas. Report

Simulation Data

Digital twin



Test Report

Design Report

CAD Drawings

Prototype manufacturing/testing

- Calibrating material properties
- Optimal experimental design

Series manufacturing

- Quality assurance
- Fault detection and correction procedures
- Magnet alignment data
- Magnet transfer function

Communication frequency

slow fast

Product life cycle

Principles of systems modeling

→ Document based information exchange

→ Communication frequency depends on the application

→ Data management services: EDMS, EAM, INFORMA, PLM

Document based → Model based

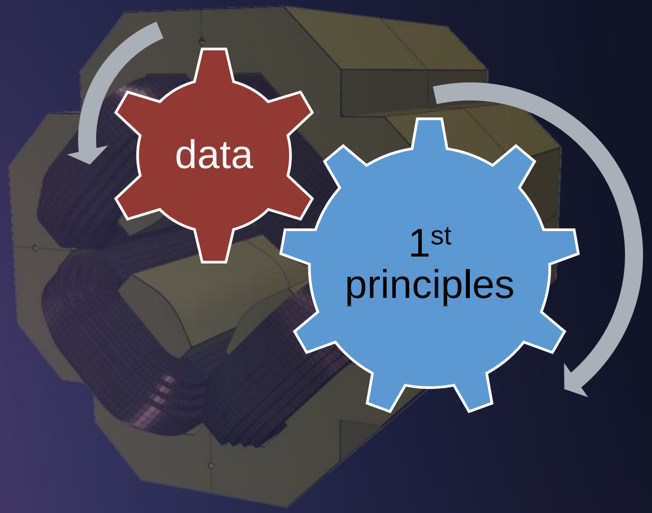
Development life cycle

Challenges

- *Complex (non-linear) dynamic system*
  - Interplay of iron saturation, hysteresis and eddy currents
  - Superconductor magnetization
  - Temperature effects
- *Computational costs*
  - A complete 3D magnet simulation does not allow for fast predictions
- *Tough requirements for machine operation*
  - Field stability at 1 unit in 100 000
  - Field quality at 1 unit in 10 000

Digital representation

a.k.a. numerical system model



→ ***Measurement data** needs to be **integrated** in the numerical model to enable **accurate predictions***

→ ***The digital twin is a hybrid model***

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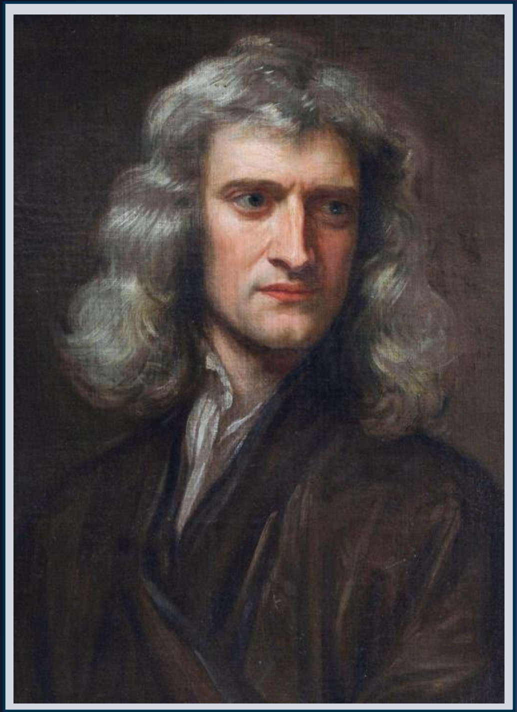
Hybrid modeling [2]

1st principle modeling

"Newtonian paradigm"

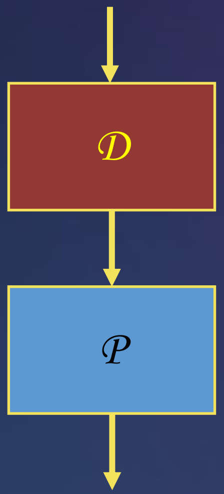
Machine learning (ML)

"Keplerian paradigm"

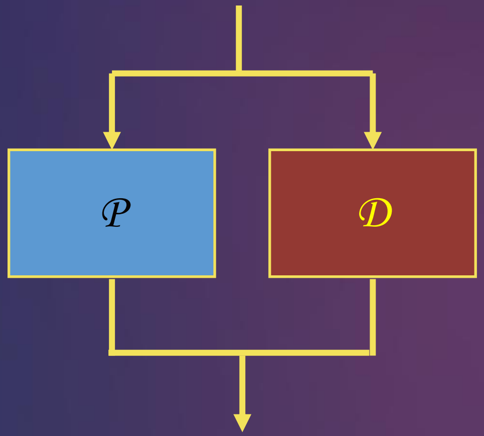


https://commons.wikimedia.org/wiki/File:Sir_Isaac_Newton_by_Sir_Godfrey_Kneller,_Bt.jpg

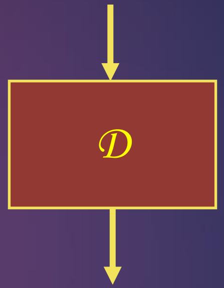
Model calibration



Delta models



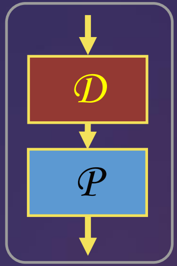
ML models



https://de.wikipedia.org/wiki/Johannes_Kepler#/media/Datei:Portrait_Confused_With_Johannes_Kepler_1610.jpg

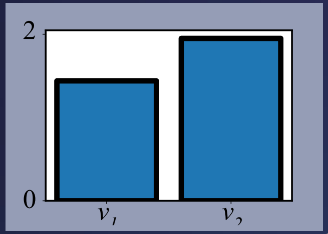
[2] Rudolph M., Kurz S. and Rakitsch B. *Hybrid modeling design patterns,* Journal of Mathematics in Industry, 2024, DOI: 10.1186/s13362-024-00141-0

Model calibration [3]

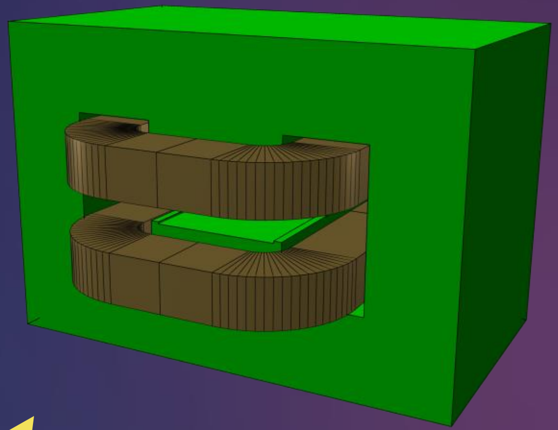


Calibration test campaign

Resulting parameters

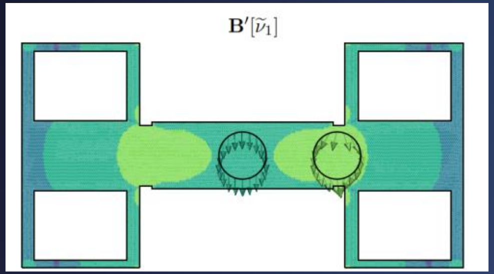


Digital twin



Design of a test campaign

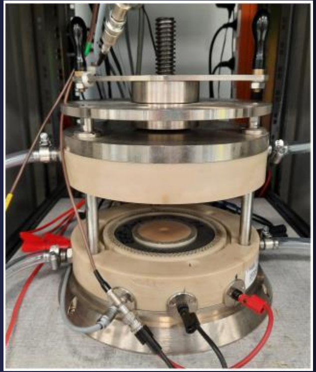
Design of experiment (DOE)
optimal sensor placement



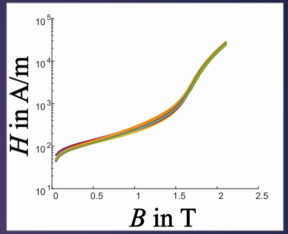
Magnet manufacturing QA

Material Specimen

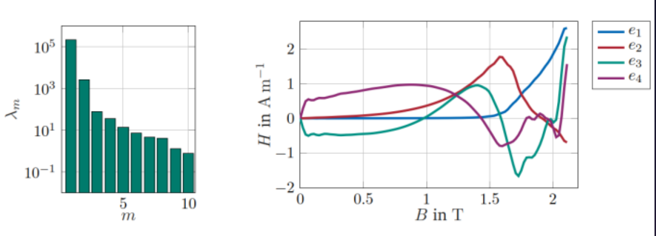
Permeameter



H(B)-curve

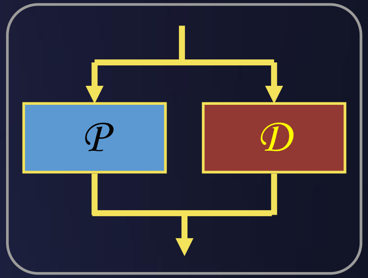


Pattern recognition (KLE)



[3] L. Fleig, M. Liebsch, S. Russenschuck and S. Schöps, "Identification of B(H) Curves Using the Karhunen Loève Expansion," in IEEE Access, vol. 12, pp. 59441-59449, 2024, doi: 10.1109/ACCESS.2024.3393348

Delta models – Example



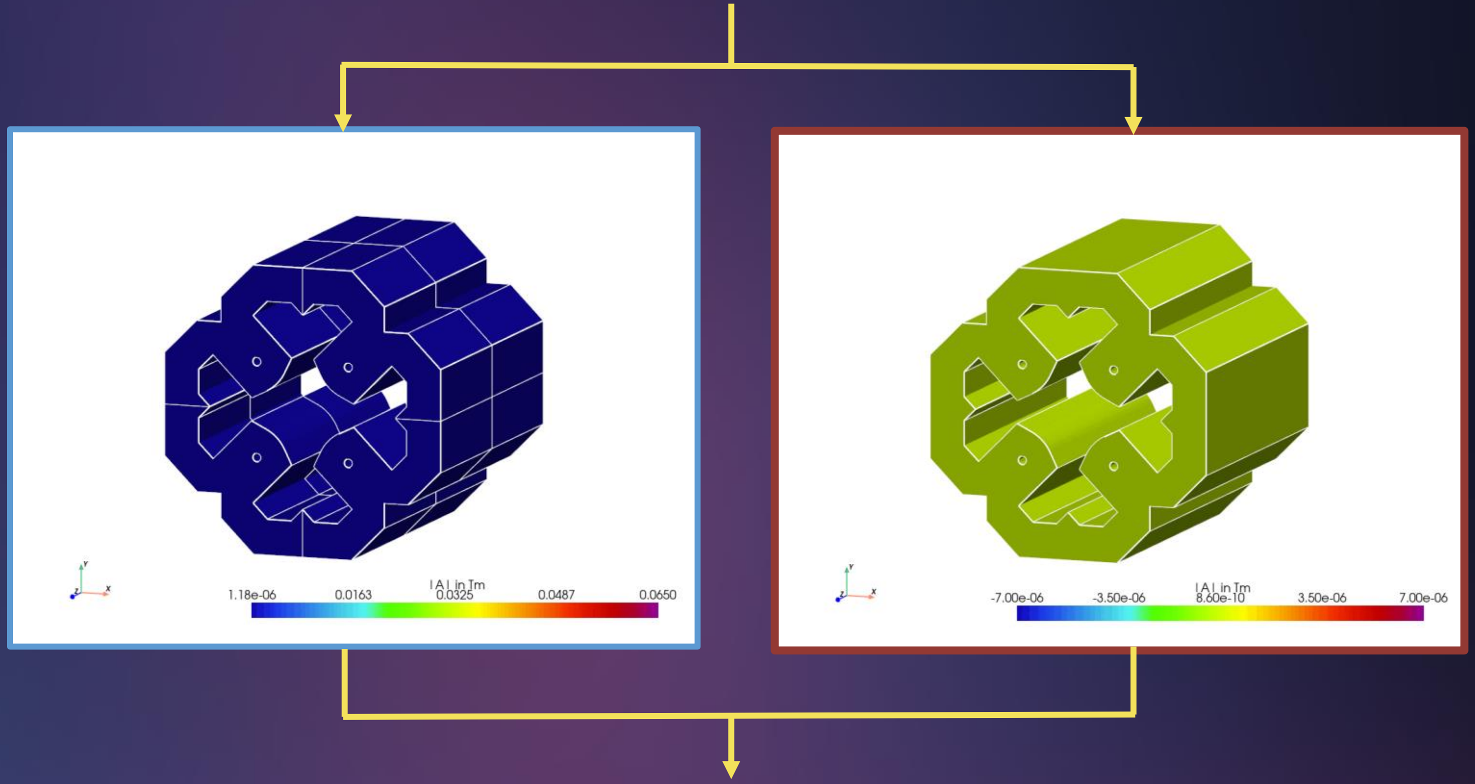
- ***Discrepancy*** between measurement and simulation ***may not vanish*** after magnet calibration

- ***Discrepancy*** drives the delta model

- Delta model ***implies Maxwell's equations*** in the vacuum domain

$$\operatorname{curl} \boldsymbol{H} = \boldsymbol{0}, \operatorname{div} \boldsymbol{B} = 0$$

- ***Boundary element*** or ***volume integral methods*** have been used



Magnet operation

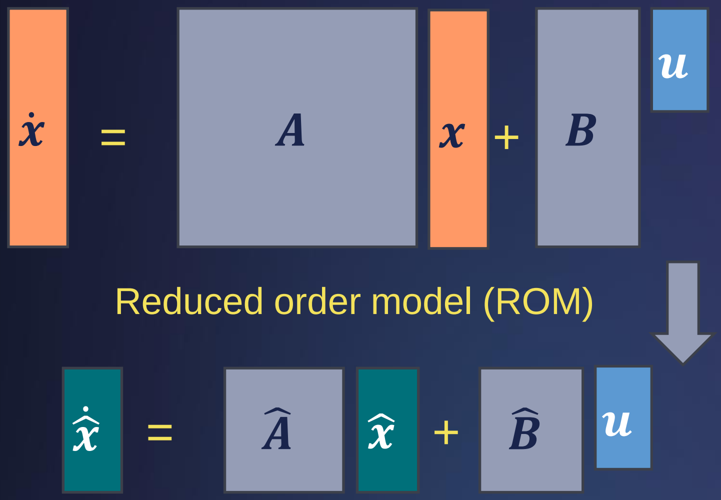
Requires rapid information exchange

Model order reduction (MOR) [5]

$$\dot{x}(t) = A\,x(t) + B\,u(t)$$

High fidelity model

Reduced order model (ROM)

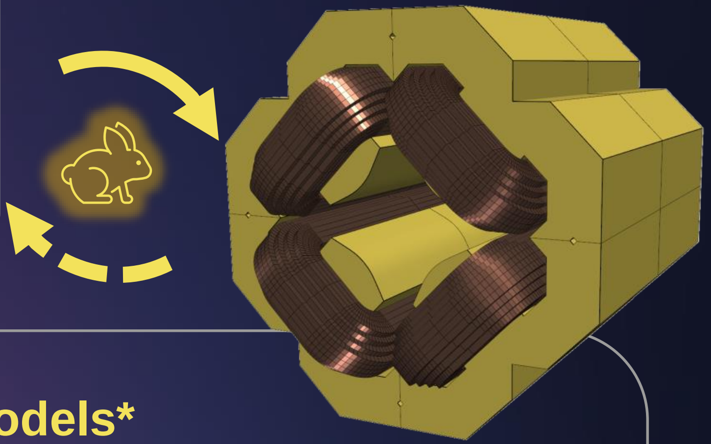


x: State vector, u: Input vector, $\hat{x}$: ROM state

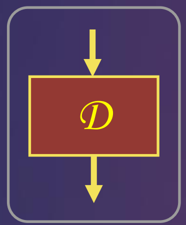
Magnet operation

- ***State observation***
- ***Prediction***
- Maintenance

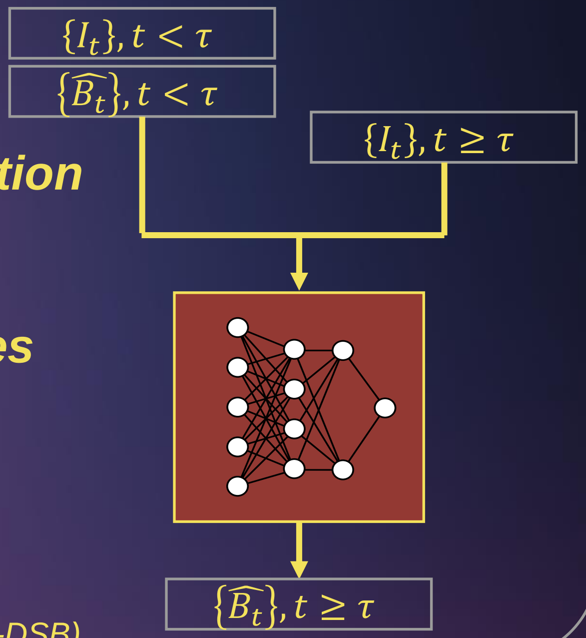
Digital twin



Machine learning models*



- Example: ***Hysteresis compensation*** for SPS dipole magnets
- Neural network-based ***Time series modeling B(I)***
- Currently developed with ***SPS training data***



*Example provided by A. Lu (PhD, BE-CSS-DSB)

[5] Sorti, S., Petrone, C., Russenschuck, S., Braghin, F.: Data-driven simulation of transient fields in air-coil magnets for accelerators, Nuclear Instruments and Methods in Physics Research Section A, 2021

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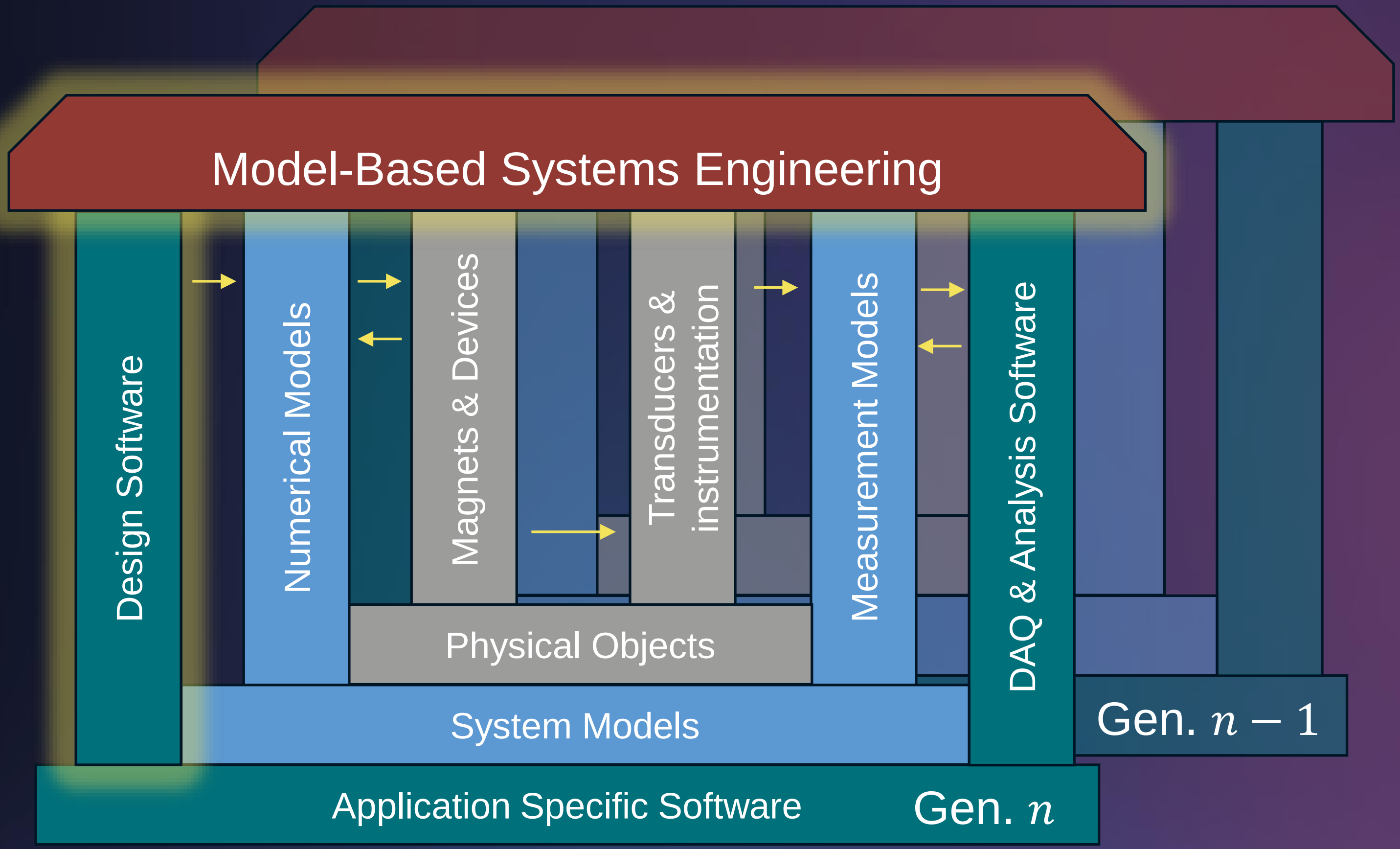
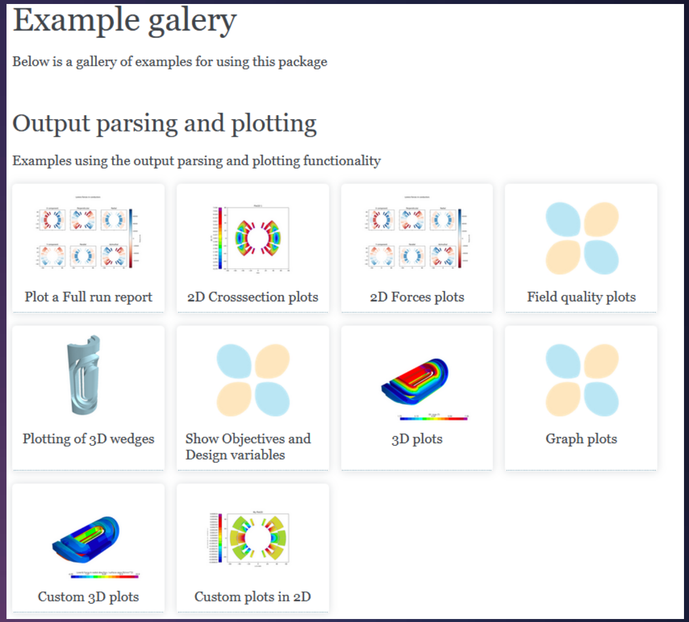
Integration

*Integrating the digital twin with **M**odel **B**ased **S**ystems **E**ngineering*

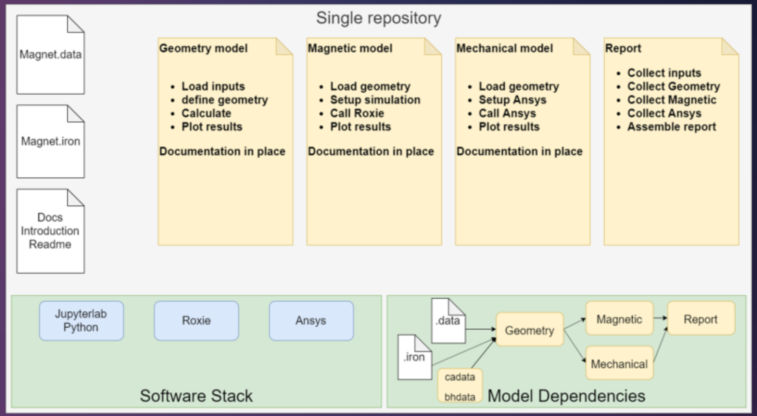


ROXIE API* for integration in MBSE



code available

pyMBSE: self-contained multi model execution



**Developed by M. Bonora (TE-MSC-TM)*

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Conclusion

Developing the digital twin of the accelerator magnet

- Numerical models and simulation data ***need to be integrated*** in the ***development and product lifecycle management***
- We must use ***hybrid modeling*** to design application specific digital twins
- In doing so we must follow the principles of ***model-based systems engineering***
  → ***Traceability, platform independence, versioning***
- A ***collaborative effort in the ATS sector*** will make the digital twin possible!

We need to build up the bridges together