

Building digital twins for granular materials using EDEM and Omniverse



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Astec Digital

Agenda



Introduction and motivation

The path to Physical AI

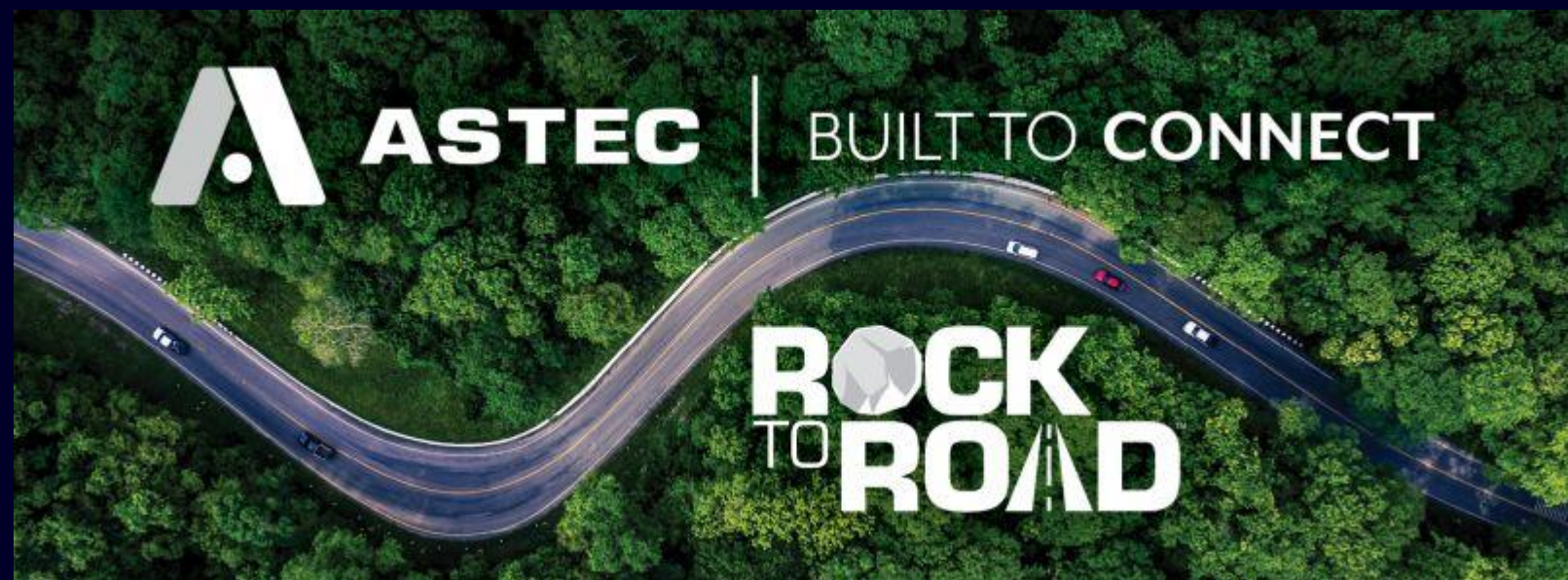
Simulation: CAE vs Robotics

Challenges of granular material simulation

Omniverse and EDEM examples

Live DEMO

Introduction to ASTEC



NASDAQ (ASTE)

www.astecindustries.com

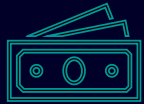
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Physical AI for infrastructure industries



Next generation asphalt plants will be autonomous and fleet connected enabling:



Reduced Labour Cost



Improved Quality



Improved Safety



Increased Sustainability



Physical AI vs Physics AI



Physical AI

Embodied artificial intelligence that is embedded into physical systems to interact directly with the material world. It relies heavily on the loop of **sensing** (cameras, LiDAR), **thinking** (processing data), and **acting** (moving motors, actuators).

Examples:

- Autonomous Vehicles
- Robots
- Smart Manufacturing

Physics AI

Machine learning and neural networks that have been trained to understand and are constrained by the laws of physics.

Examples:

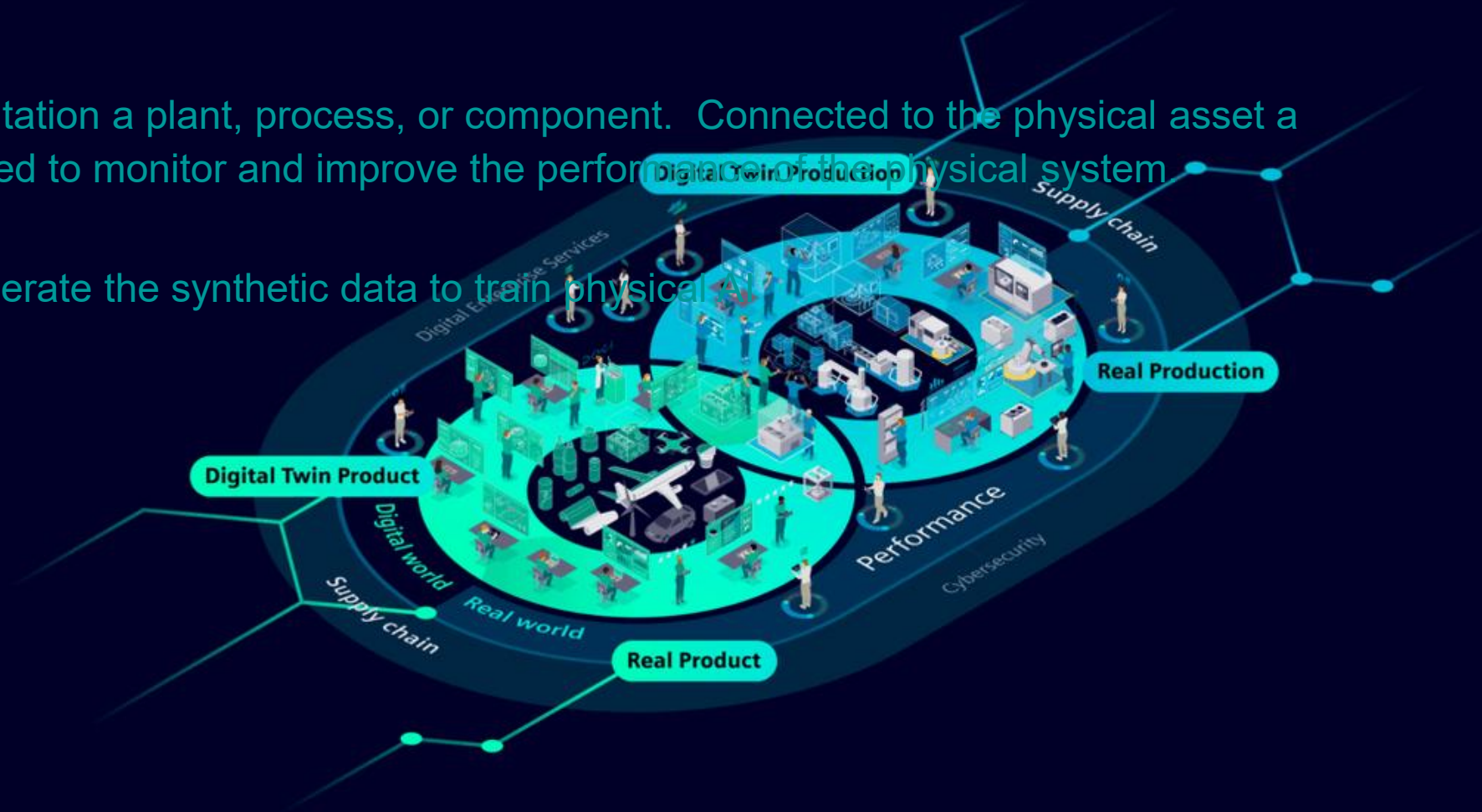
- Reduced order modeling (ROM's)
- Physics informed neural networks (PINN's)
- Emerging deep learning architectures
 - DoMino
 - Transolvers/GeoTransolvers
 - AB-UPT

Digital Twin

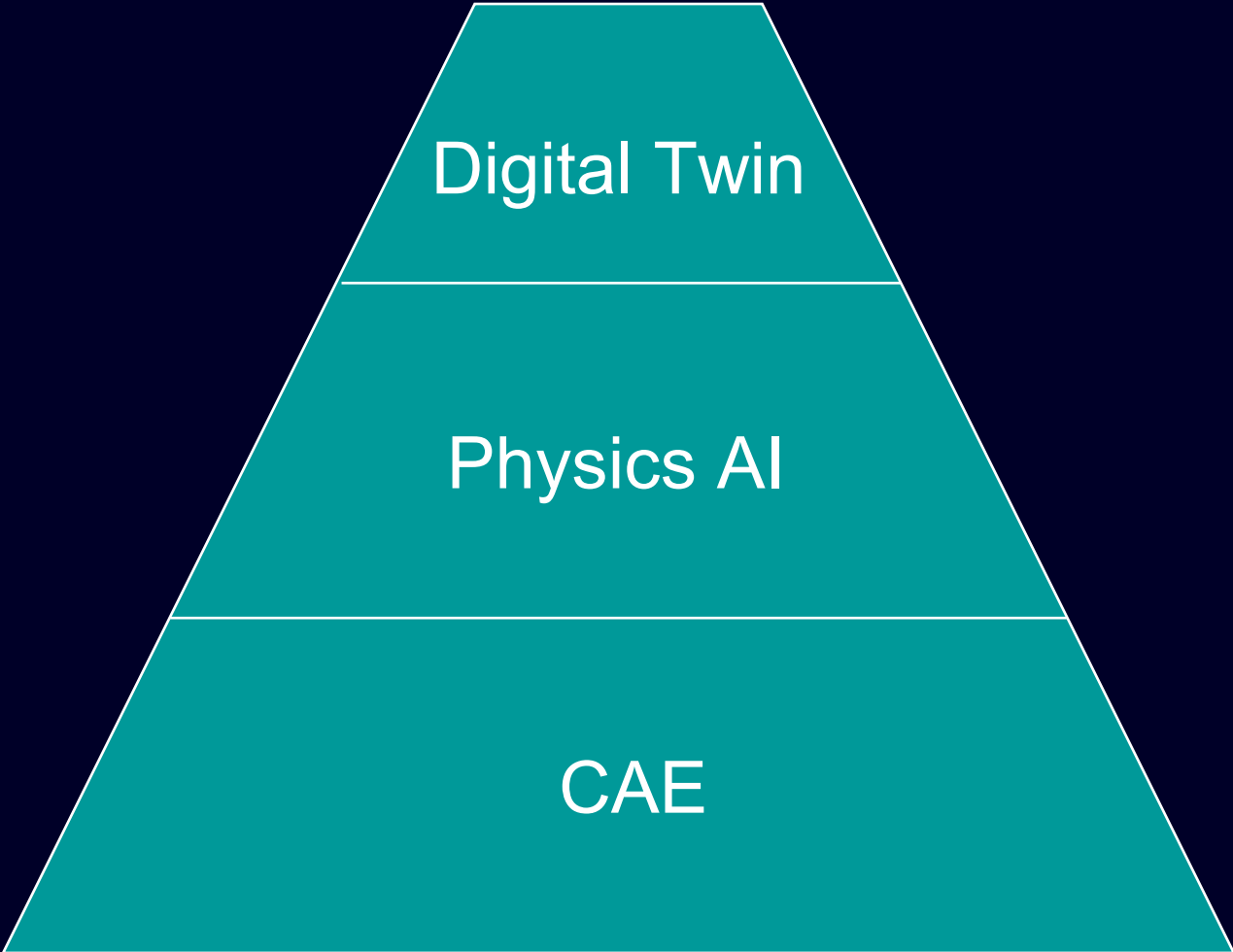


Digital Twin is a virtual representation a plant, process, or component. Connected to the physical asset a predictive digital twin can be used to monitor and improve the performance of the physical system

A digital twin is how we can generate the synthetic data to train physical AI



Road to Physical AI



A teal pyramid divided into three horizontal sections. The top section is labeled 'Digital Twin', the middle section is labeled 'Physics AI', and the bottom section is labeled 'CAE'. The pyramid is positioned on the left side of the slide, with descriptive text for each level to its right.

Digital Twin

Generate synthetic data to train physics aware DL models and connect these to the physical system

Physics AI

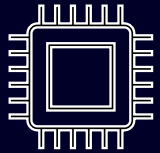
Ability to run physics models faster than real-time to inform machine controls or train policies

CAE

Models that accurately represent the physics of the system and environment

Physics Simulation: CAE and Robotics

CAE vs Robotics Simulation



CAE PARADIGM

High-Fidelity: Focused on sub-millimeter continuum accuracy to eliminate physical prototyping risks.

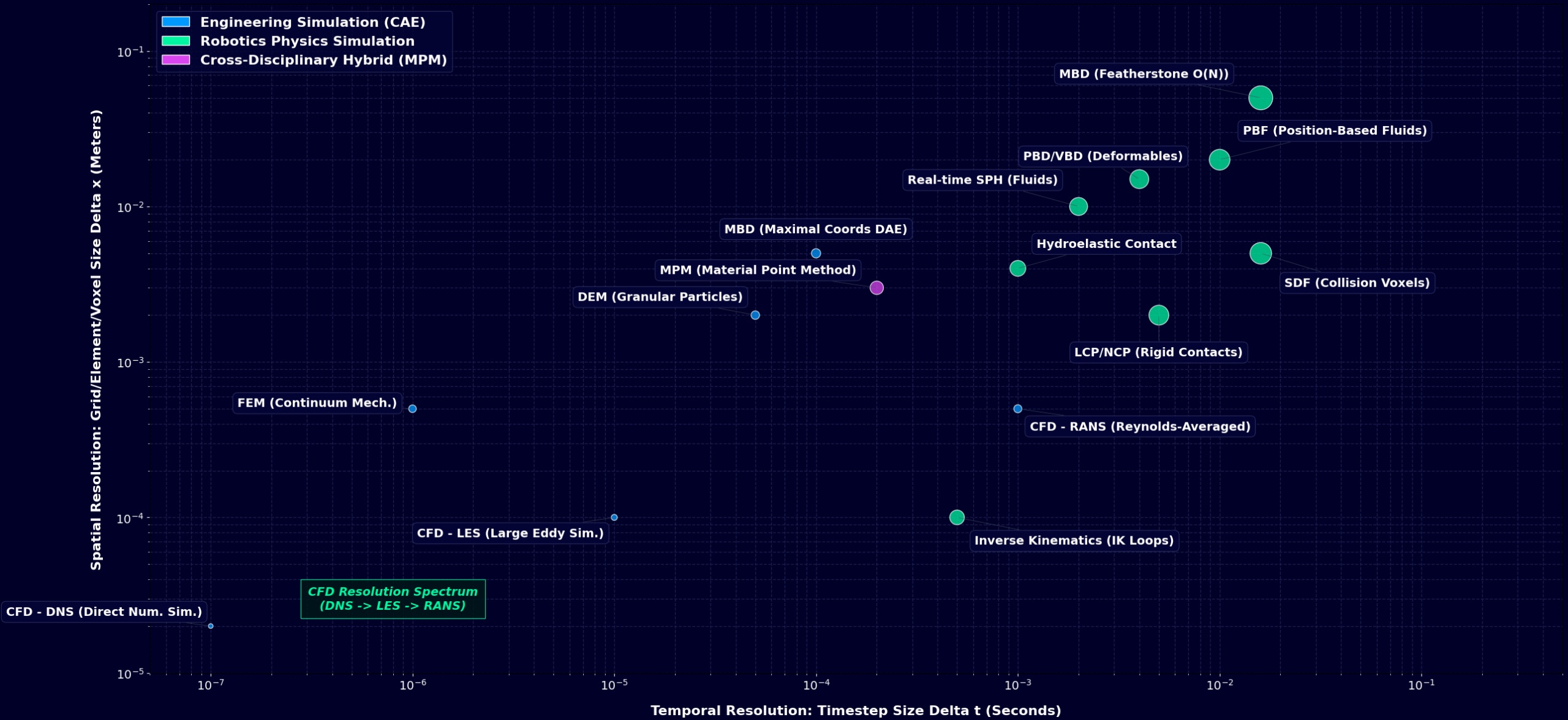


ROBOTICS PARADIGM

High-Throughput: Scaled to execute millions of quick physical interactions for continuous RL loops.

Core Task	Engineering Simulation (CAE)	Robotics Physics Simulation
Motion Mapping	Maximal coordinates tracking 6-DOF per link. Unconstrained, slow matrix processing. DAE Solver $O(N^3)$ Scale	Reduced joint coordinates via Featherstone's recursive dynamics; integrated Inverse Kinematics. Featherstone $O(N)$ Scale
Collision Detection	Direct mesh-to-mesh geometric intersection checking. Highly detailed boundary searches. FEM Topologies	Simplified primitive shapes and Signed Distance Fields (SDF) allowing fast distance calculation. Volumetric SDF $O(1)$ Check
Contact Dynamics	Microscopic penalty spring-dampers modeling physical compliance. Requires tiny timesteps. Stiff Non-Linear Springs	Rigid body non-penetration loops solved using Linear & Nonlinear Complementarity Problems. Hard LCP / NCP Solvers
Stress and Deformation	Full continuous volumetric meshes mapped to non-linear physical stress-strain equations. Continuum Mechanics (FEM)	Assumed rigid solids; dynamic meshes approximated via Position-Based Dynamics or soft-patches. Position-Based Dynamics (PBD)
Fluid Dynamics	Exact finite volume Navier-Stokes solvers Computational Fluid Dynamics (CFD)	Liquid simulation via particle positions to satisfy geometric incompressibility constraints Position-Based Fluids (PBF)
Particle Dynamics	Explicit calculation of particle-particle and particle-wall contacts and trajectories Discrete Element Method (DEM)	Tracks moving particles and solves equations of motion on a stationary background grid Material Point Method (MPM)

Simulation Fidelity



When is high fidelity physics required in robotics simulation?

Challenges – Granular Materials

Many industries process granular materials:

- Pharmaceuticals
- FMCG
- Steel manufacturing
- Chemicals
- Mining
- Construction
- Agriculture

Training autonomous systems to operate in these industries requires the ability to accurately simulate the complex behavior of granular materials



Build-up



Blockage



Spillage



Wear



Segregation



Poor compaction



Examples:

- Digital twin for visualization
- Digital twin for robotics co-simulation

What is Omniverse?



NVIDIA Omniverse™ is a collection of accelerated libraries and microservices for developing physical AI simulation applications



OMNIVERSE

- OpenUSD
- Physics and Motion
- Rendering and Simulation
- Runtime and Deployment



OpenUSD

- A file-format agnostic scene description
- Non-destructive composition
- Addressable data model
- Scalable
- Ecosystem and standardization



PXAR

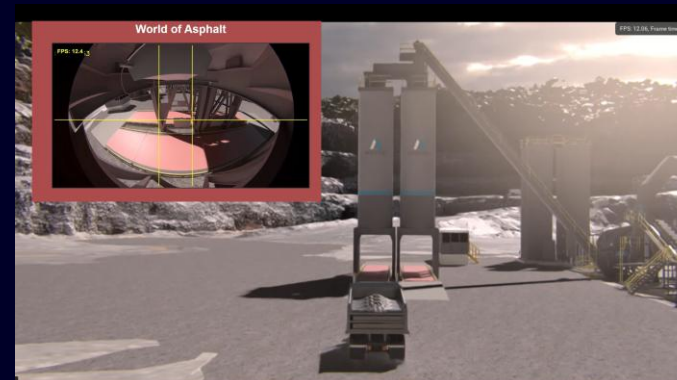
ASTEC Omniverse use cases:



CAE Visualization



XR

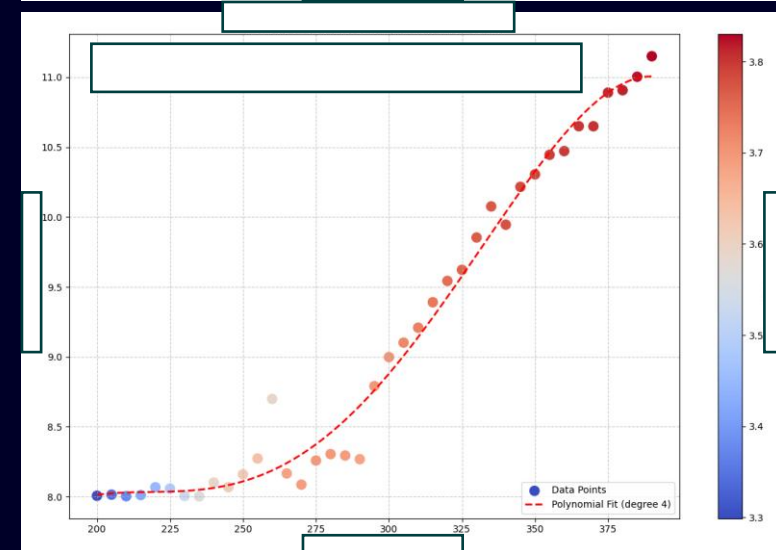
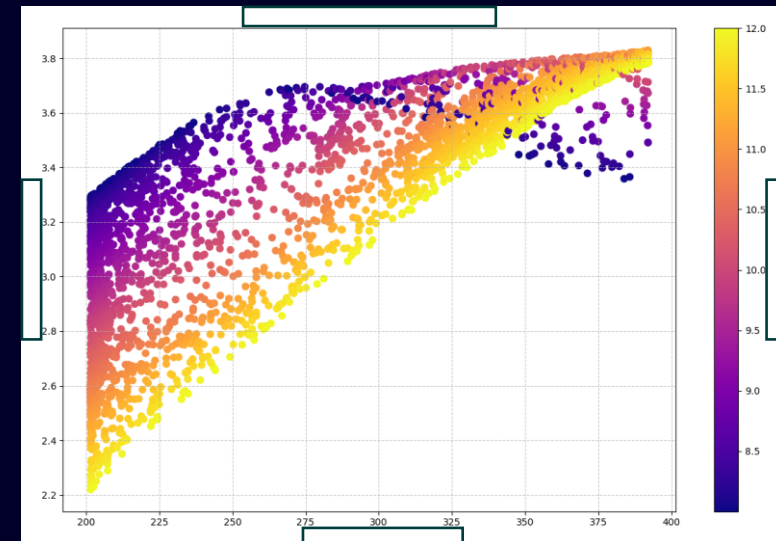
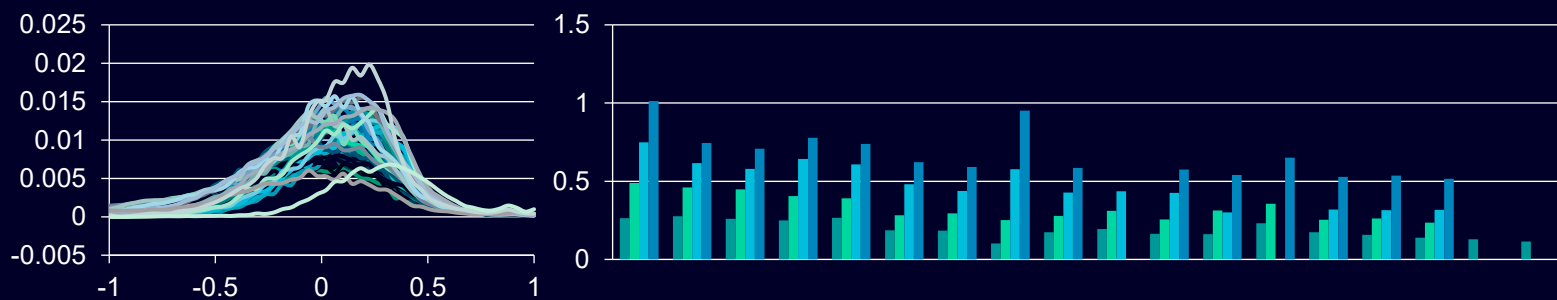
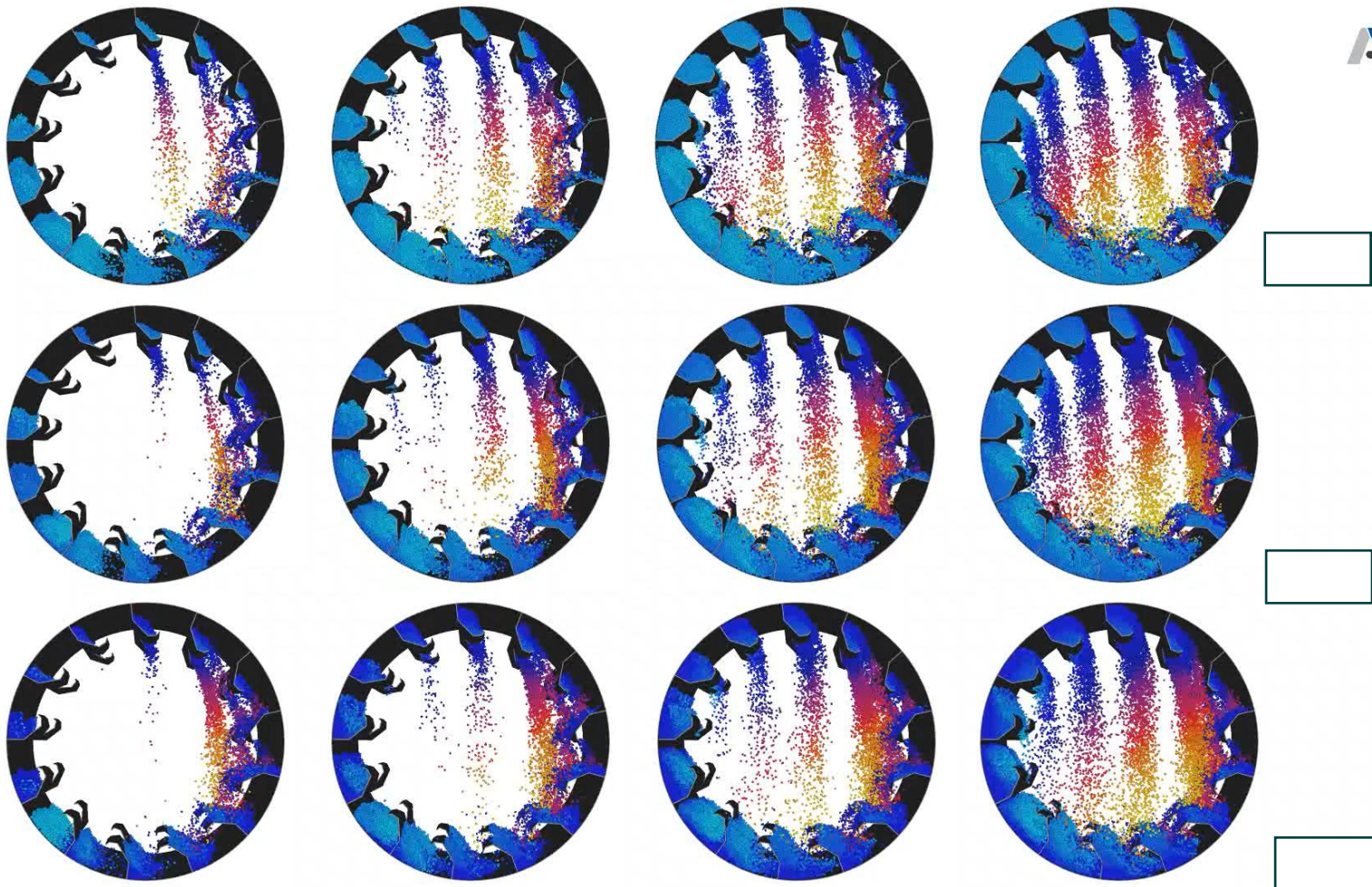


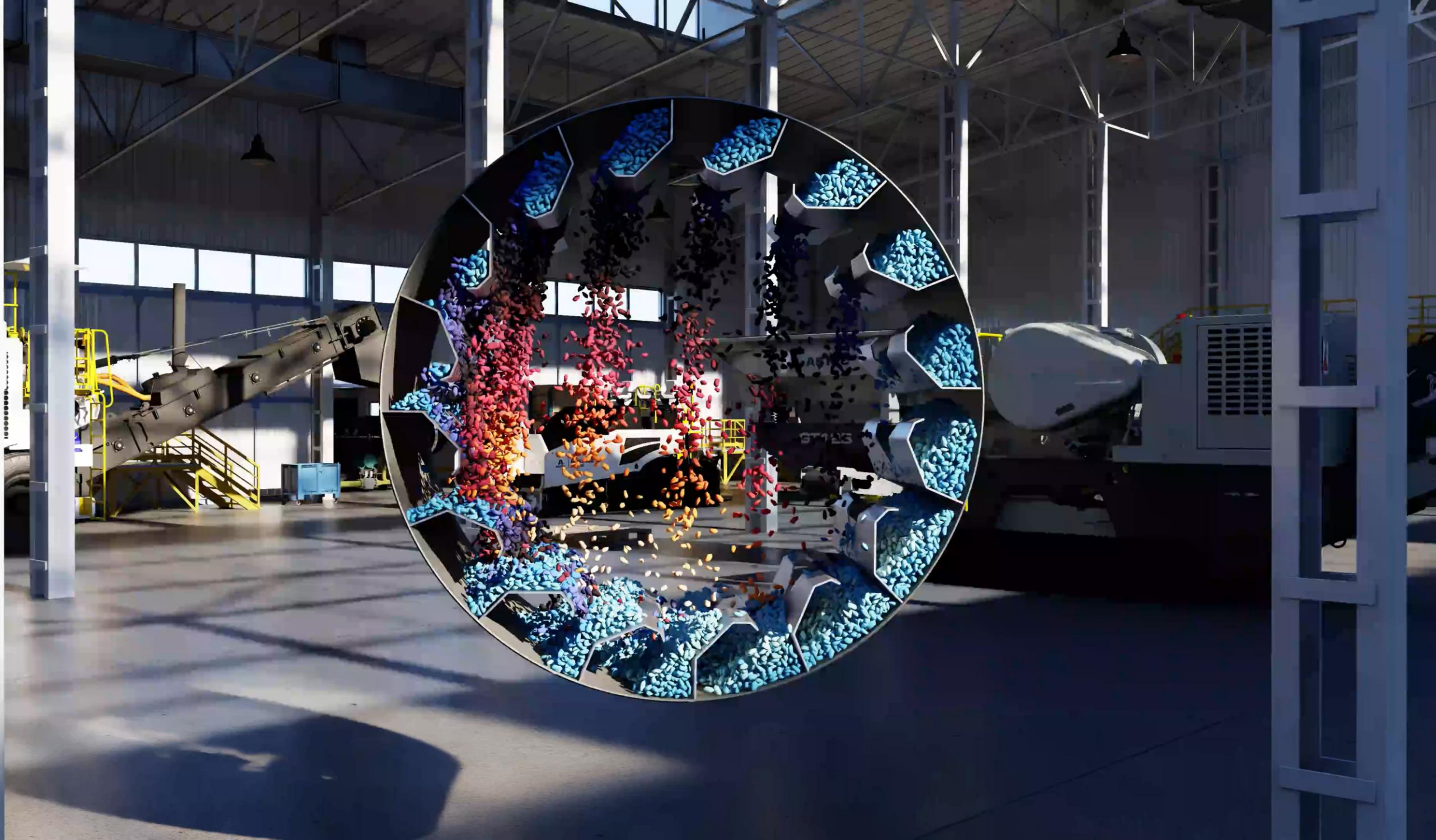
Computer Vision



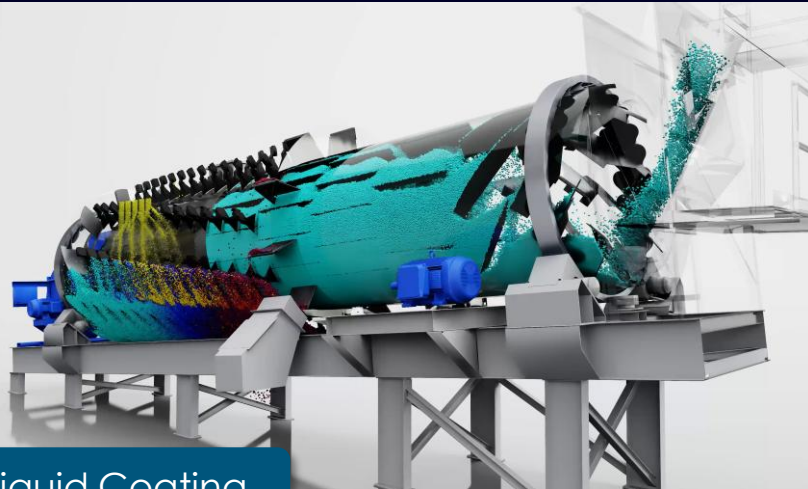
Digital Twins

Digital twin for visualization: aggregate dryer





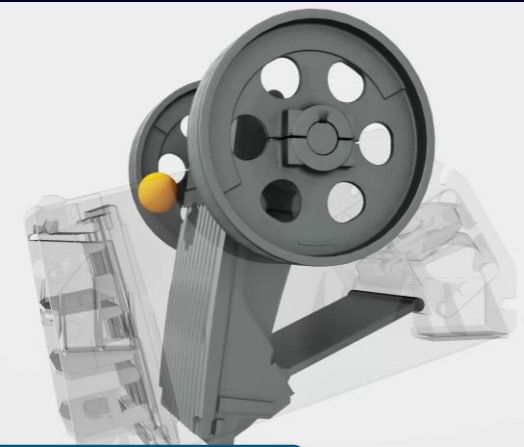
DEM Visualization Using Omniverse



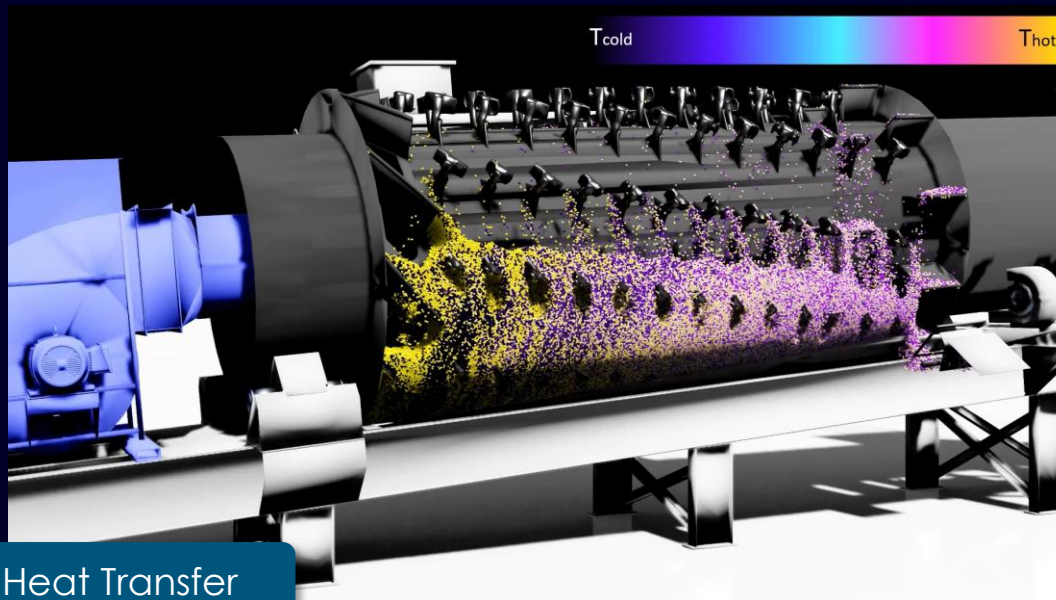
Liquid Coating



Mixing



Crushing



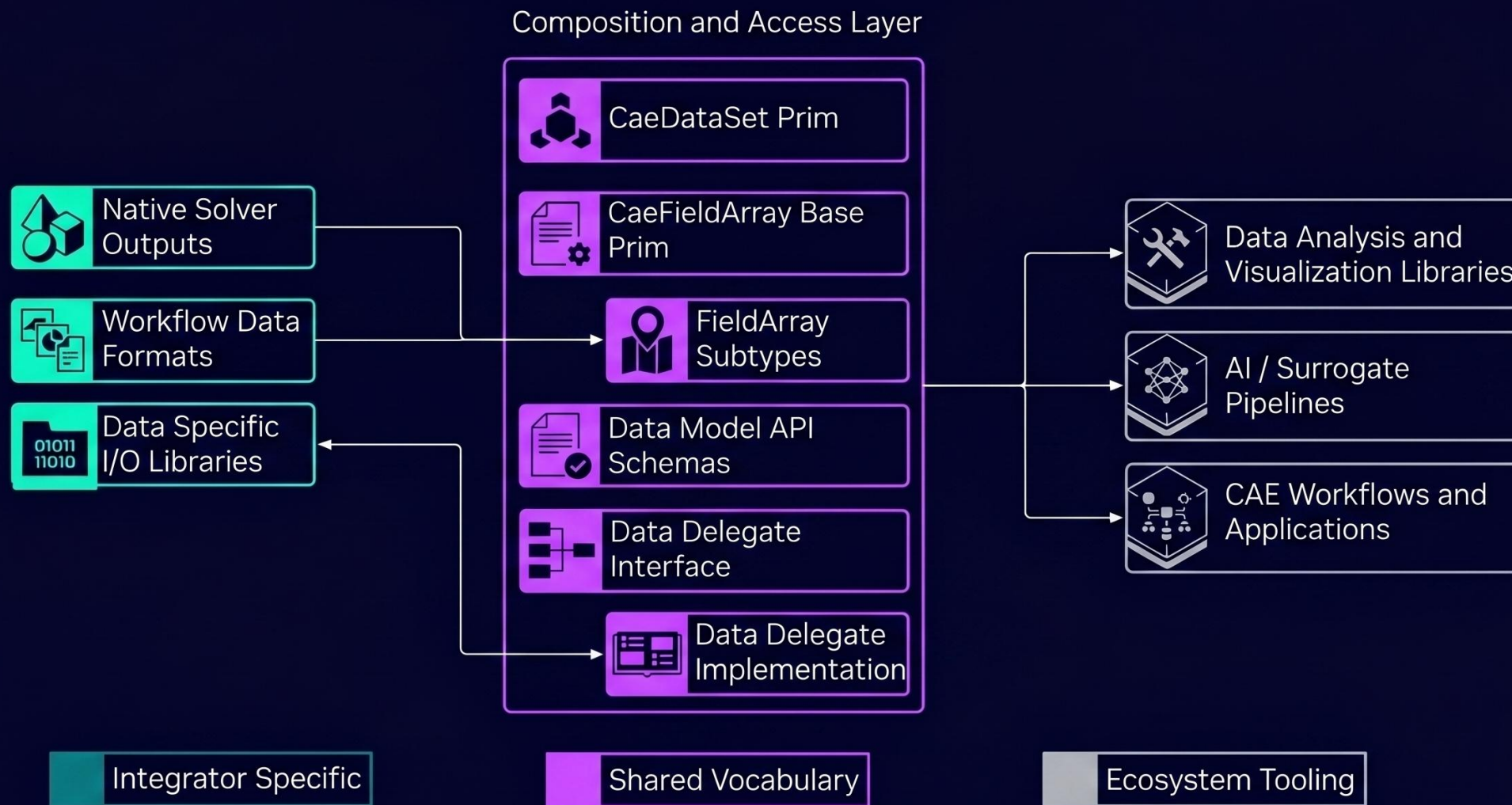
Heat Transfer



Drying

OpenUSD for CAE

The Data Composition And Access Layer



<https://github.com/NVIDIA-Omniverse/kit-cae>

Kit-CAE 3.0 File Importers



CGNS (.cgns): Structured and unstructured meshes with their field data.

EDEM (.dem): Particle simulation data.

EnSight Gold (.case, .encas): Mesh and field data, including time-varying and polyhedral results.

Eclipse reservoir (.egrid, .grdecl, .data, .init, .unrst): Reservoir grids and result properties.

NanoVDB (.nvdb): Sparse volume data.

NumPy (.npz, .npy): Point clouds and arrays from Python or AI workflows.

OpenFOAM (.foam): OpenFOAM cases without an intermediate conversion.

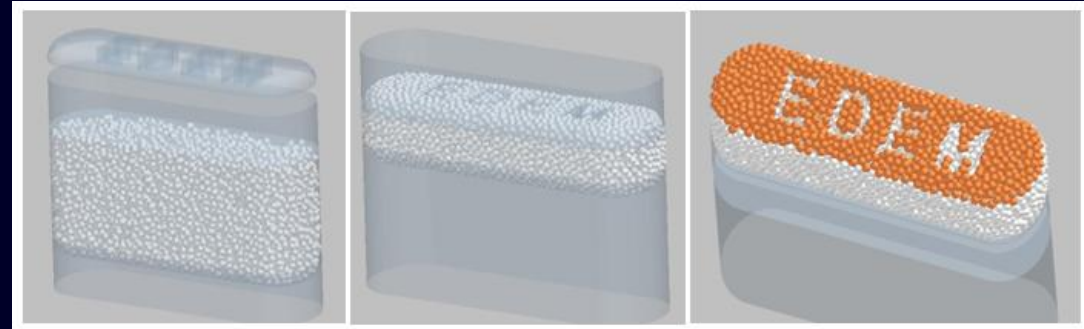
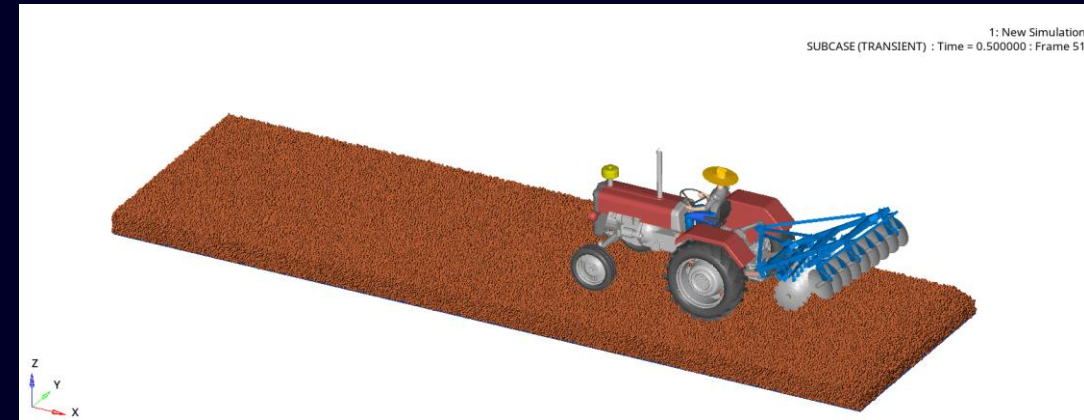
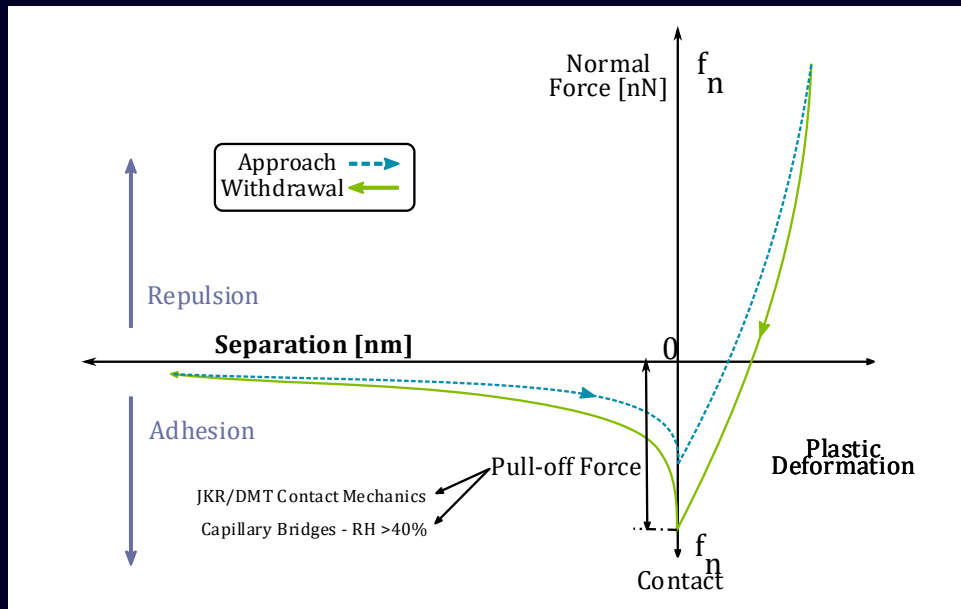
VTK (.vtk, .vti, .vtr, .vts, .vtp, .vtu): Structured and unstructured VTK datasets.

Digital twin for robotics co-simulation

Complex Particle Dynamics: Cohesive and Compressible Materials

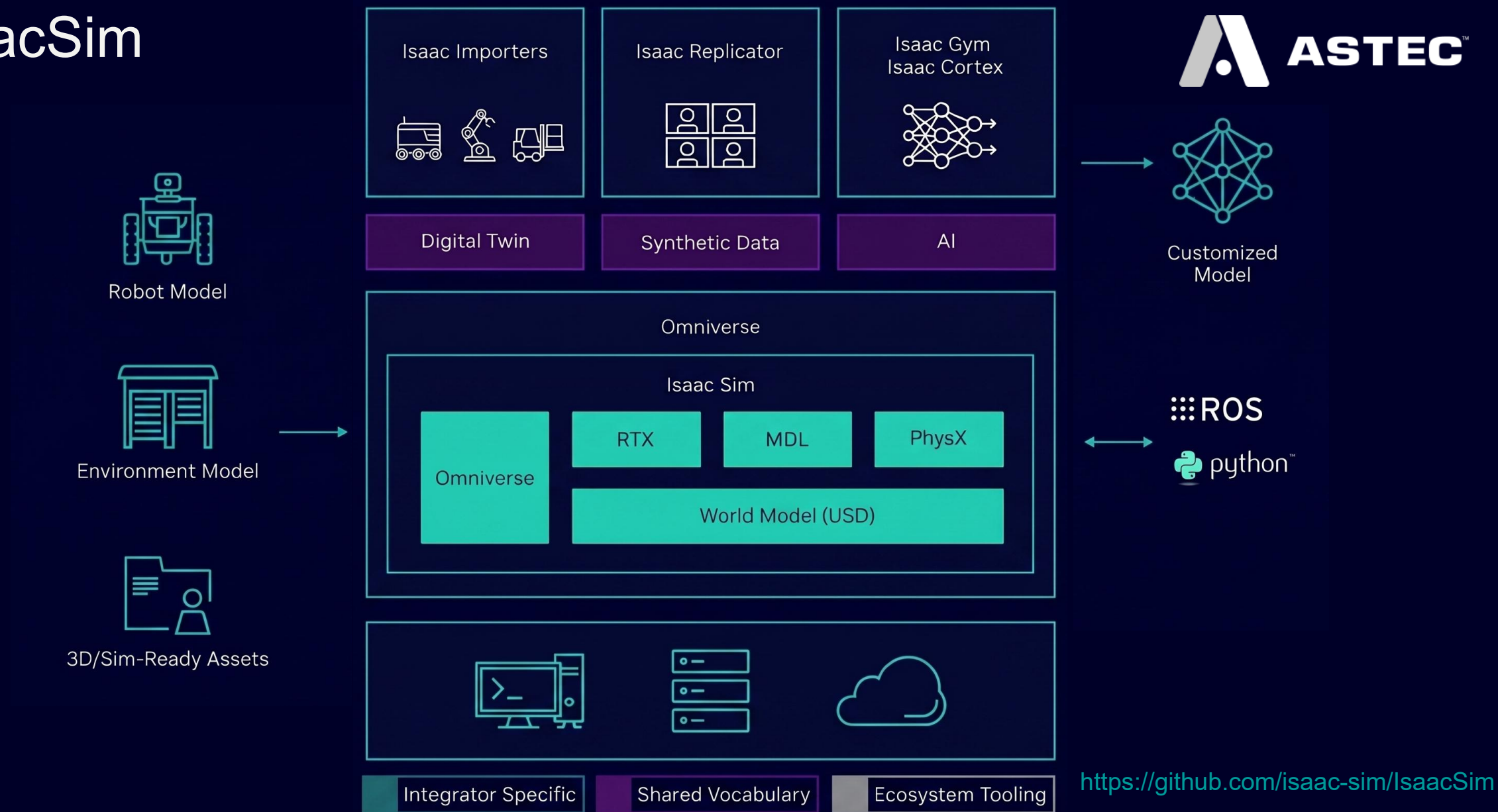


The Edinburgh Elasto-Plastic Adhesion (EEPA) model comprises a nonlinear hysteretic spring model to account for the elastic-plastic contact deformation and an adhesive force component that is a function of the plastic contact deformation. When two particles or agglomerates are pressed together, they undergo elastic and plastic deformations. It is assumed that the pull-off strength (adhesion) increases with an increase of the plastic contact area.



1. Morrissey, J.P., *Discrete Element Modelling of Iron Ore Fines to Include the Effects of Moisture and Fines*, University of Edinburgh, 2013. <https://www.era.lib.ed.ac.uk/handle/1842/8270>. ↵
2. Thakur, S.C., Morrissey, J.P., Sun, J., Chen, J.F., & Ooi, J.Y. (2014). *Micromechanical analysis of cohesive granular materials using discrete element method with an adhesive elasto-plastic contact model*. *Granular Matter*, 16(3), 383–400. doi:10.1007/s10035-014-0506-4 ↵
3. Morrissey, J.P., Thakur, S., & Ooi, J.Y. (2014a). *EDEM Contact Model: Adhesive Elasto-Plastic Model*. University of Edinburgh. doi:10.13140/RG.2.2.10139.18724 ↵
4. Morrissey, J.P., Thakur, S., & Ooi, J.Y. (2014b). *Using the Elasto-plastic Adhesion Model: Example Problem*. University of Edinburgh. doi:10.13140/RG.2.2.10978.04809 ↵
5. S. C. Thakur, John P. Morrissey, Jin Sun, Jian-Fei Chen, Jin Y. Ooi, 2014. Micromechanical analysis of cohesive granular materials using discrete element method with an adhesive elasto-plastic contact model. *Granular Matter*, vol.16, pp.383-400.

IsaacSim



Omniverse-EDEM Coupling



Omniverse/IsaacSim

Simulate rigid bodies
& robotic kinematics
Large timesteps



GEOMETRY MOTION

Rigid body position – orientation – velocity

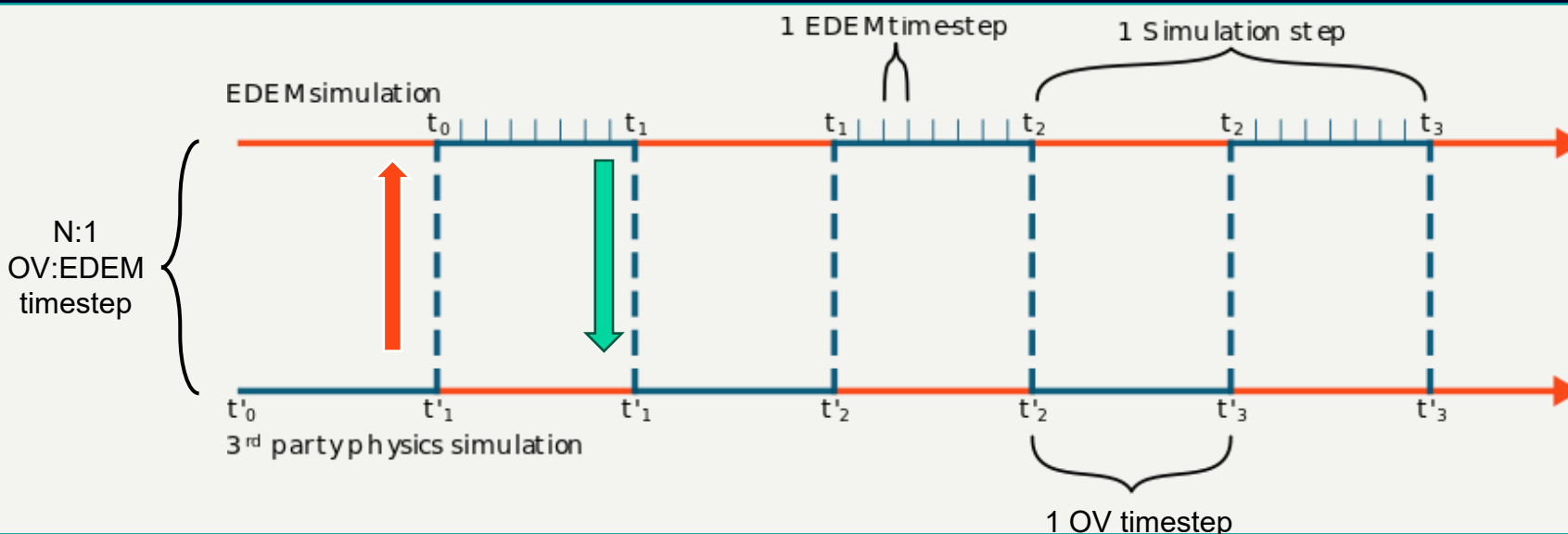
PARTICLE BEHAVIOUR

Resulting forces - torques on rigid bodies

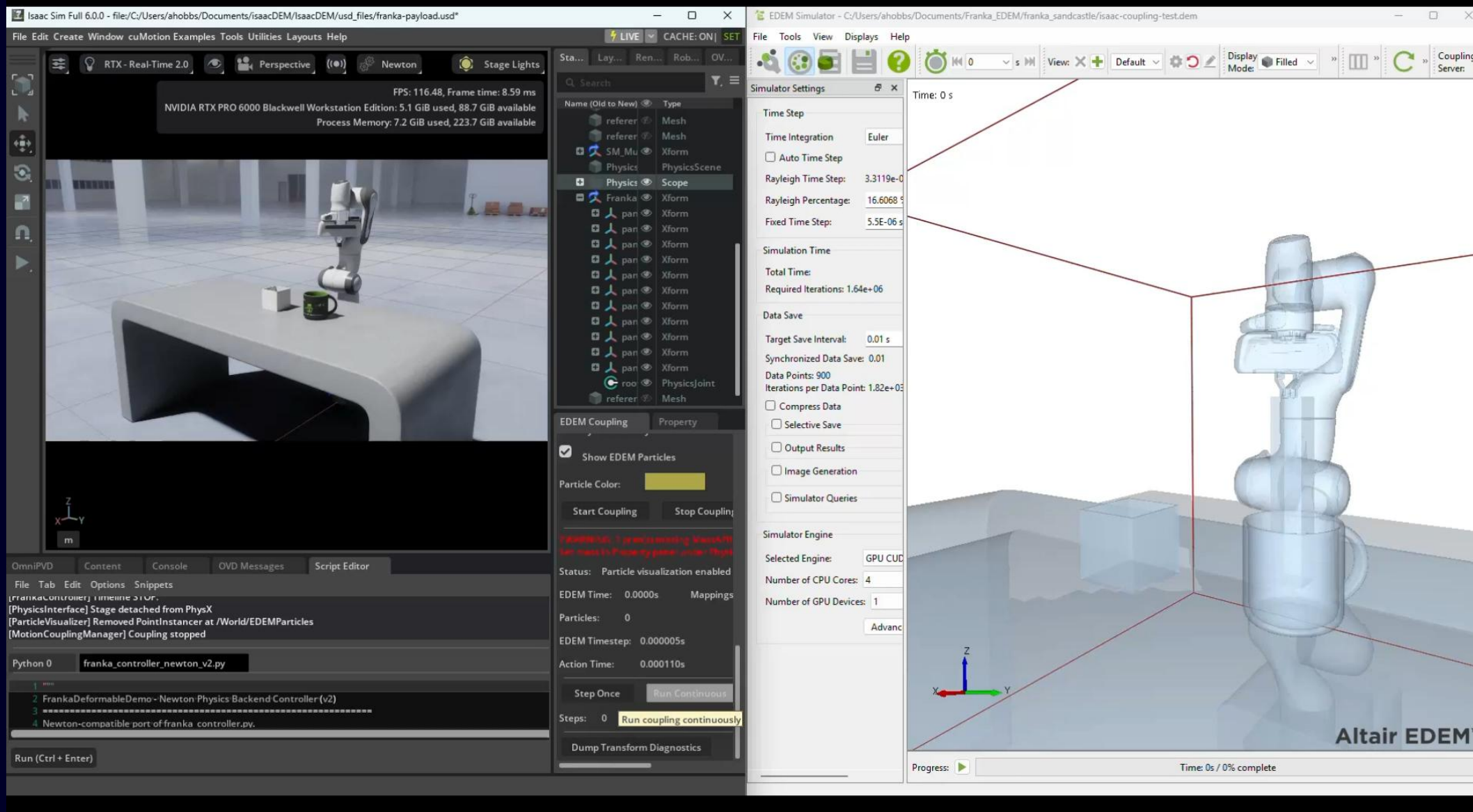


EDEM

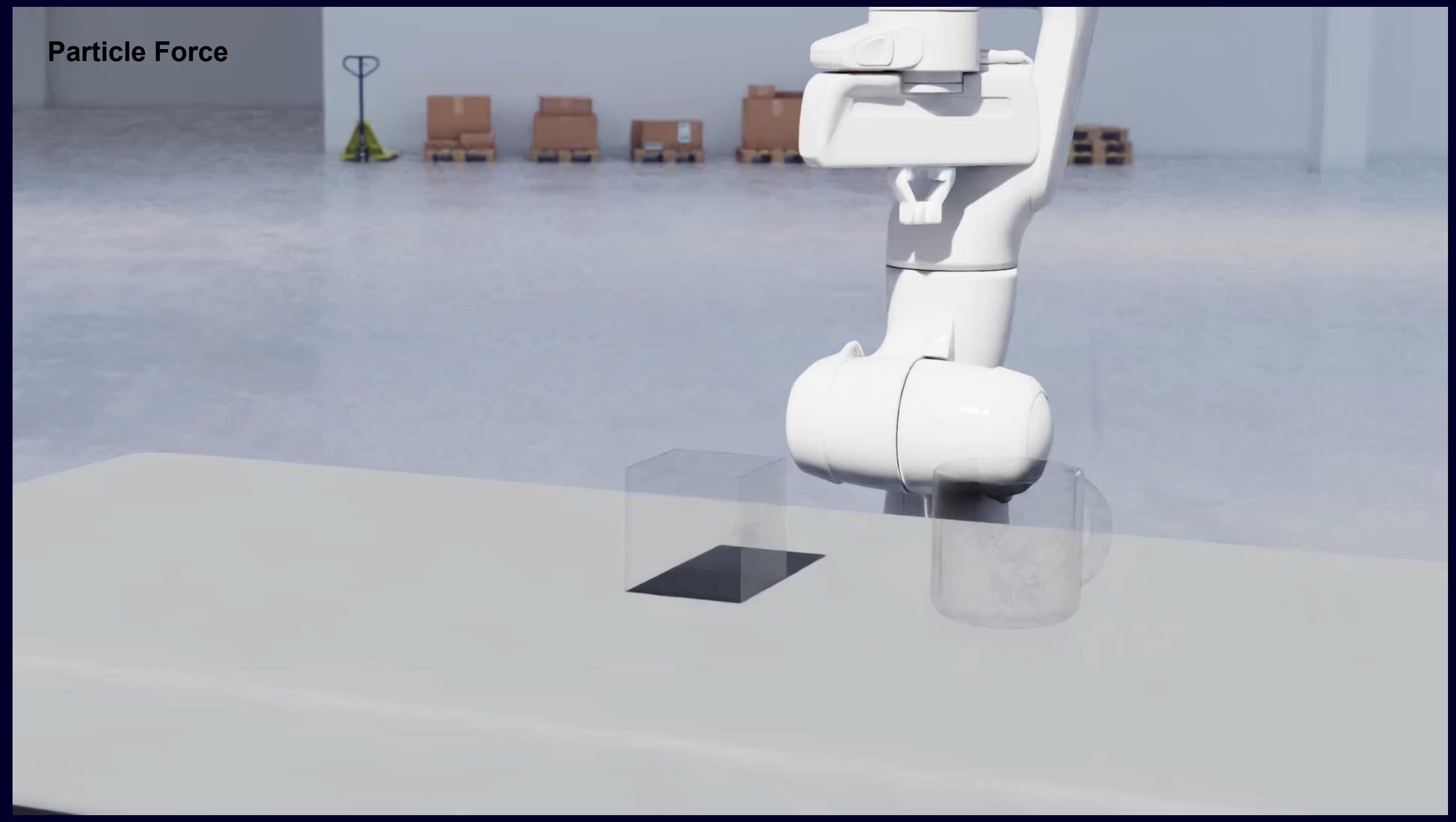
Interpolate
geometry kinematic
Multiple small
timesteps



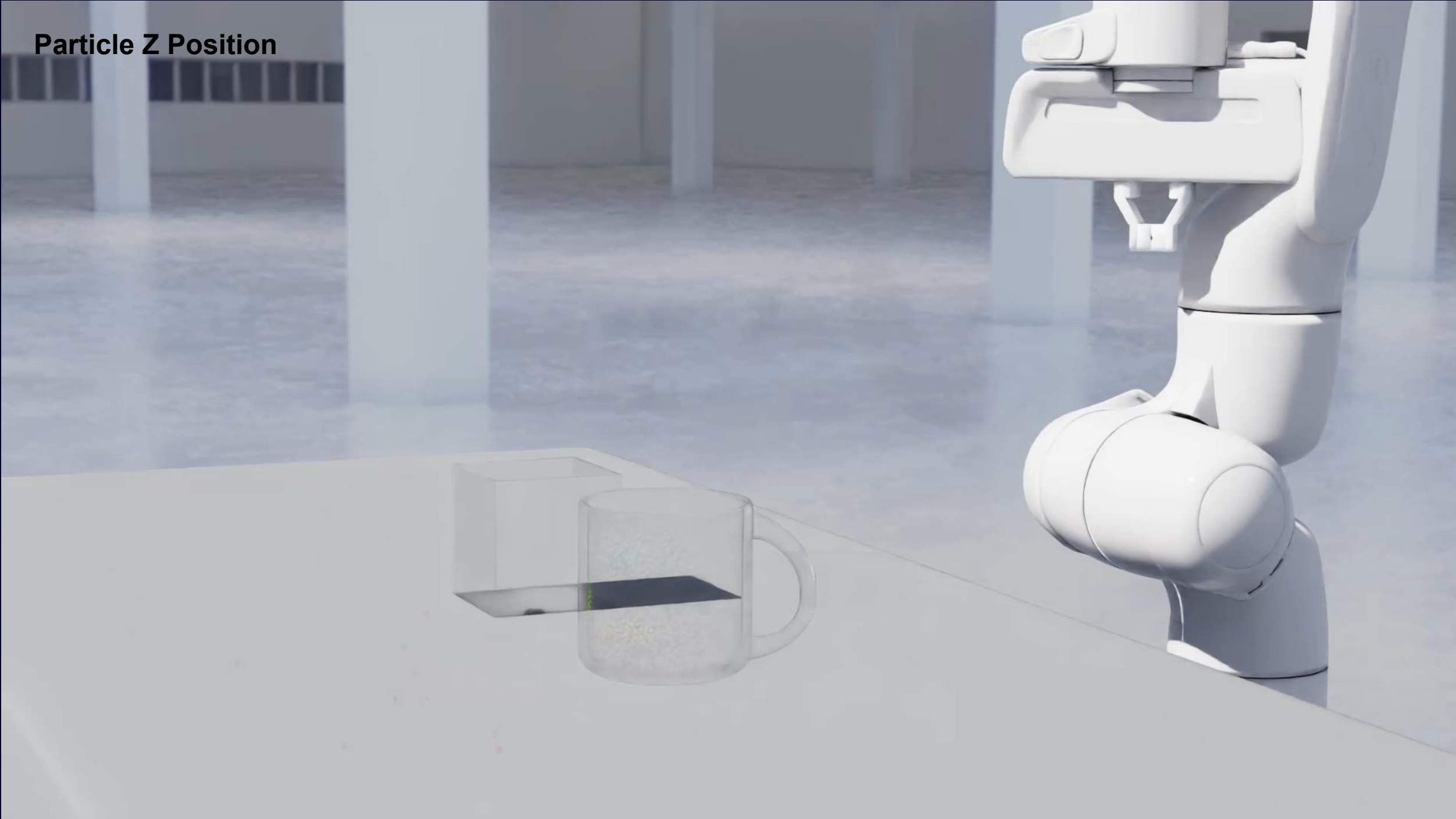
EDEM – Omniverse Coupling



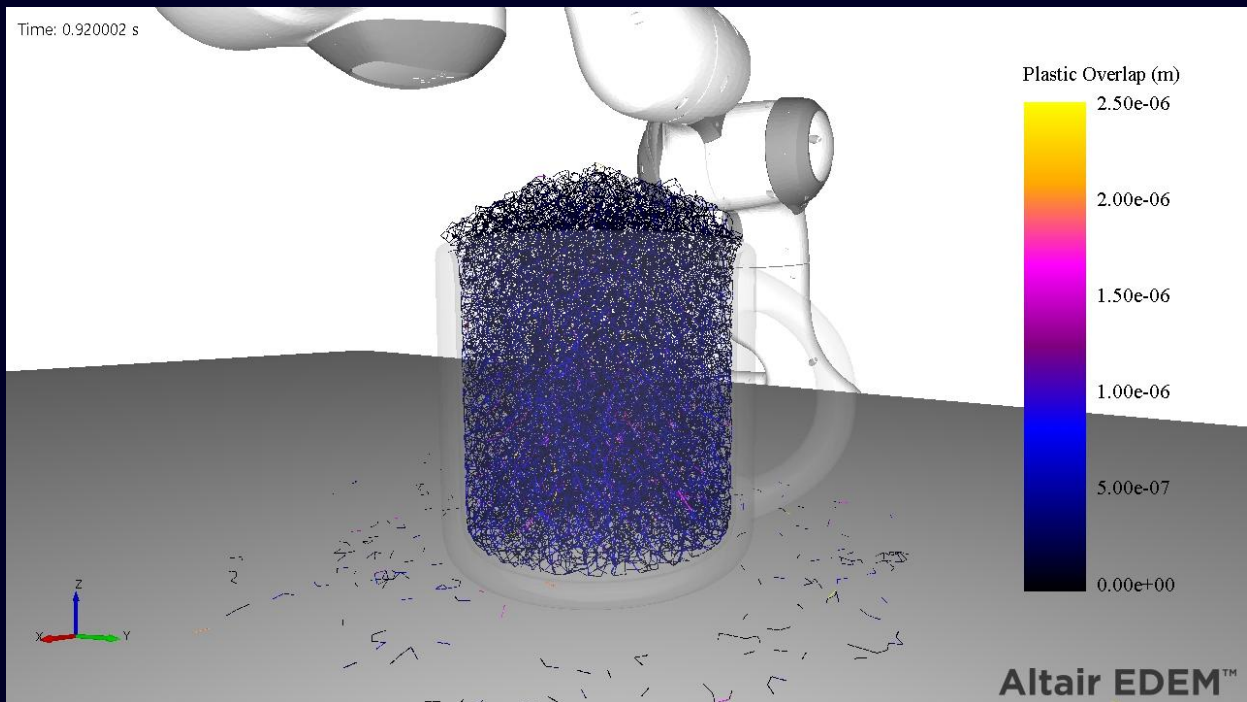
Particle Force



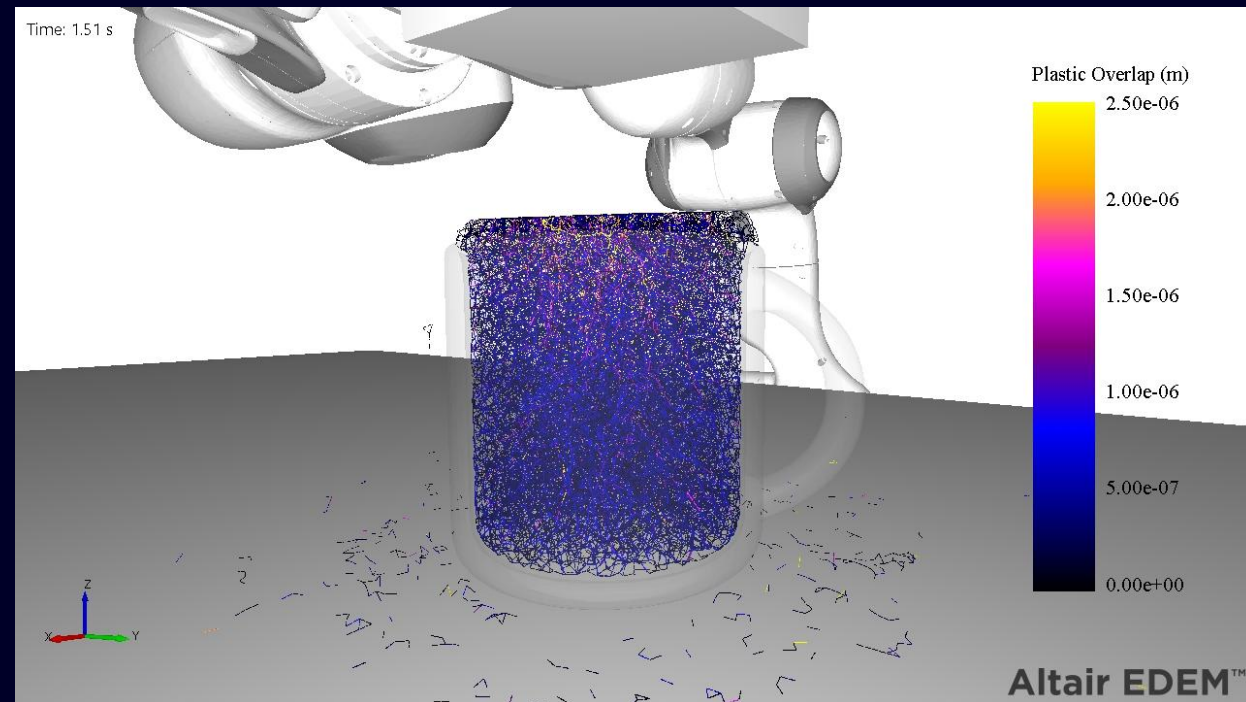
Particle Z Position



Compaction



Pre-compaction minimal plastic overlap



Post-compaction plastic overlap and permanent deformation

Conclusions



Present a path towards physical AI by demonstrating a coupling between Nvidia Omniverse/IsaacSim and Siemens Simcenter EDEM

- Demand for autonomous systems to improve safety, quality, efficiency, and performance
- Granular materials are common across many industries
- Complex behaviors like cohesion and compaction require advanced contact models
- Bridging the gap between real-time robotics simulation and high-fidelity CAE remains a challenge
- Need for robust physics AI approaches to accelerate simulation of granular systems



Physical AI starts with [fast] accurate physics

Kit-CAE DEMO



Main Repo: <https://github.com/NVIDIA-Omniverse/kit-cae>

EDEM glyph coloring fix: <https://github.com/andrewinsheffield/kit-cae-3.0>

#Clone repo:

```
gh repo clone andrewinsheffield/kit-cae-3.0
```

```
# change directory to install folder
```

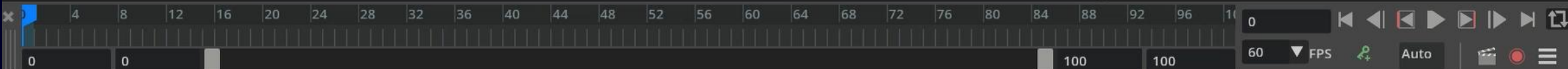
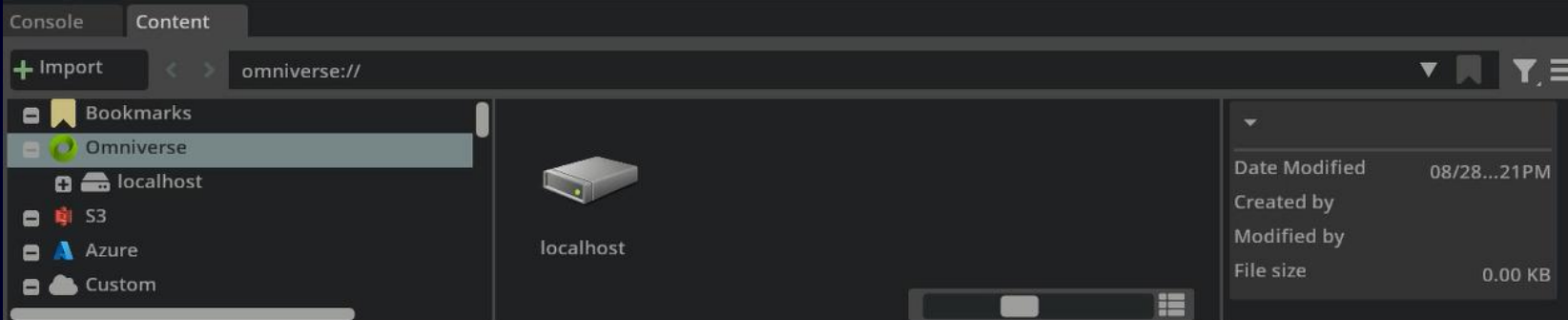
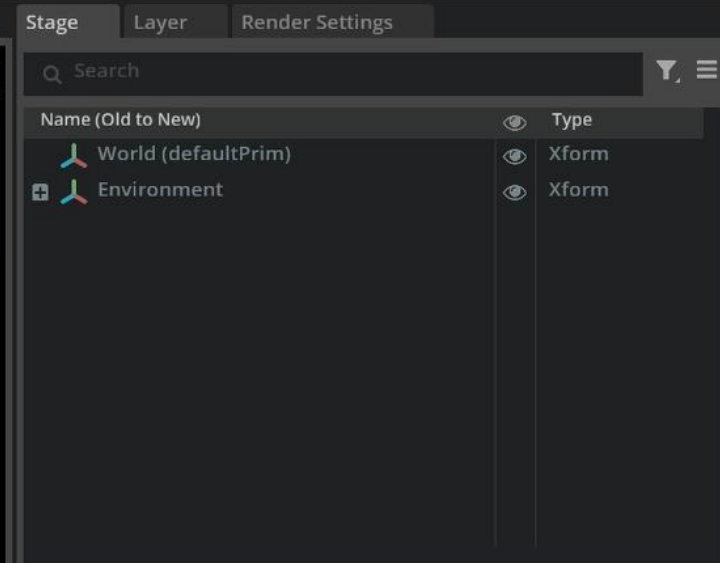
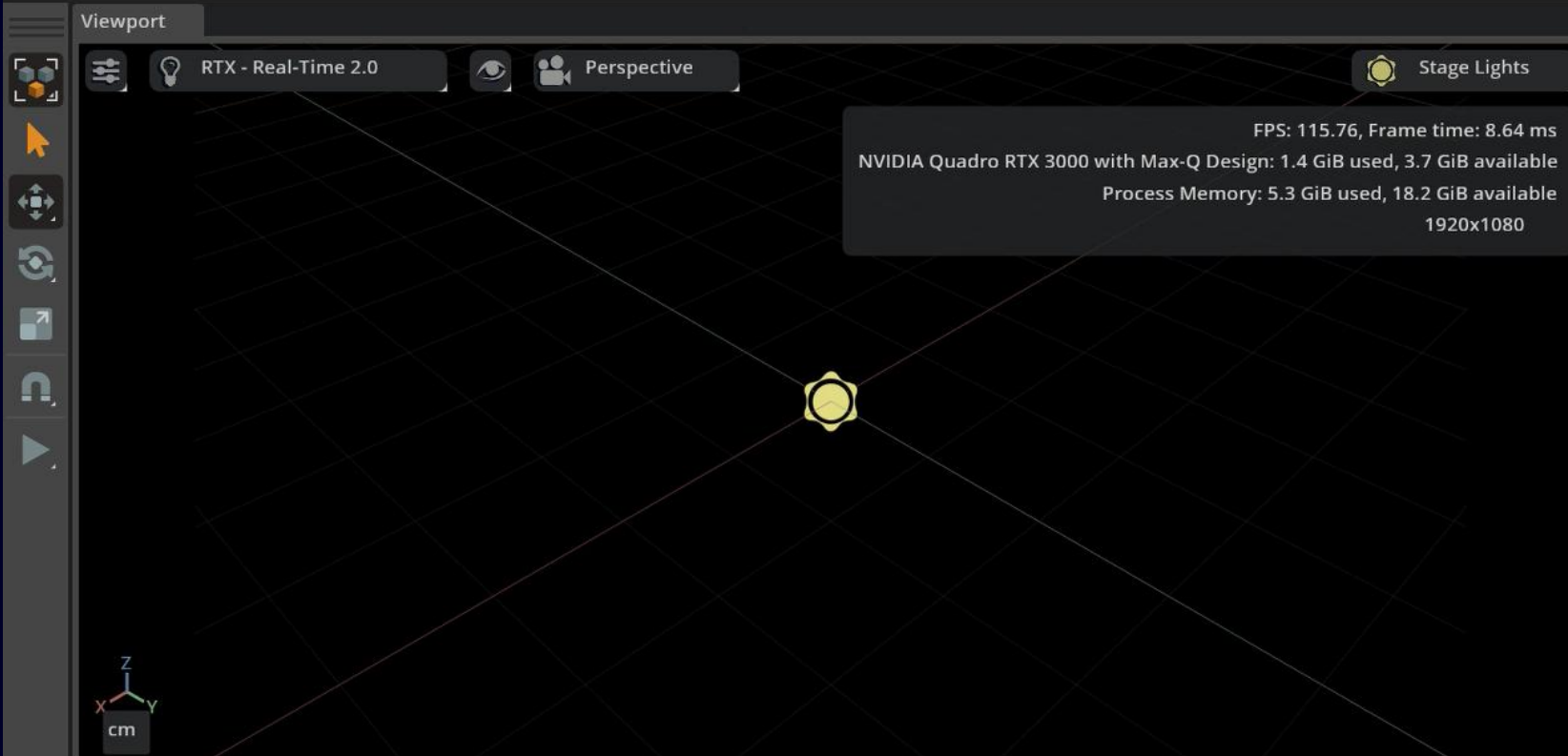
```
cd kit-cae-3.0
```

```
# Build application
```

```
./repo.bat build -r
```

```
# Launch application with legacy VTK
```

```
./repo.bat launch -n omni.cae_vtk.kit
```



Omni CAE Editor (with VTK) (Kit: 110.1.3) 3.0.0

File Edit Create Window Developer Help

Viewports

RTX - Real-Time 2.0

Perspective

Stage Lights

Process Memory: 6.6 GiB used, 32.5 GiB available

Import

C:/Users/ahobbs/Documents/OV-Composer/kit-cae-3.0/DEM-poly-test/JP_VTKHDF/

khunter

offline

Public

Warehouse-Assets

Windows

D:

E:

Name

particle_templates

simulation_t_0.usda

simulation_t_0.vtkhdf

simulation_t_1.vtkhdf

simulation_t_2.vtkhdf

simulation_t_3.vtkhdf

JP_VTKHDF

Date Modified 09/02/20...06:36AM

Created by

Modified by

File size 48.00 KB

Stage Layer Render Settings

Search

Name (Old to New)	Type
World (defaultPrim)	Xform
simulation_t_0	Scope
ParticleTypes	Scope
Particles	Scope
GeometryGroups	Scope
CAE	Xform
Glyphs_Paired	PointInstancer
Prototypes	
Materials	Scope
ScalarColor	Material
Shader	Shader
Looks	Scope
Environment	Xform

Property

Search

+ Add Shader

Prim Path /World/CAE/Glyphs_Paired/Materials/ScalarColor/Shade

Instanceable

Shader

Inputs

Coloring

Scalar Domain X 0.0 Y 2.0

Color LUT

Enable Coloring

Use Prototype Display Color

0 4 8 12 16 20 24 28 32 36 40 44 48 52 54 56 60 64 68 72 76 80 84 88 92 96 100

54

60 FPS Auto