

# PICon GPU ||| + X

## *Building blocks for successful Exascale accelerator simulations*

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## PICongGPU is ready for Exascale

- Scales up to the largest HPC systems
- Performance-portable via (NVIDIA, AMD, Intel, ARM,...)
- Implements **exascale workflows** for scalable I/O capabilities, such as streaming, data reduction and visualization workflows.

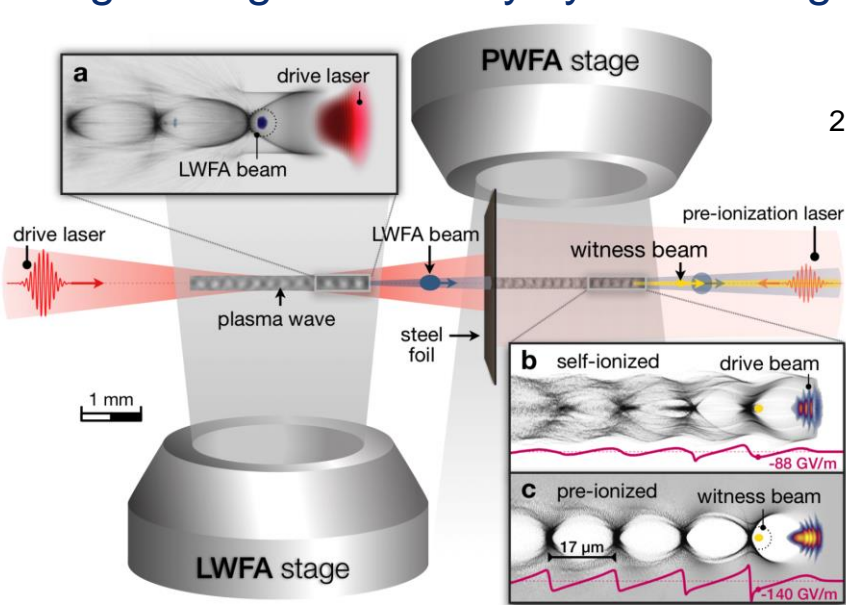


Experiments usually do not have an exascale system nearby...  
 How do we leverage the power of exascale computing?



# From laser-plasma accelerator experiments to digital twins

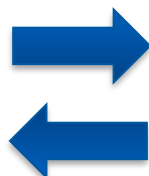
tightening the links by synthetic diagnostics and experimental data reconstruction



**Hybrid LWFA+PWFA accelerator experiment**

$\sim 50\text{-}100$  simulations  
2-3 sims per week on 200 GPUs

**Experimental data  
reconstruction**



**Synthetic diagnostics  
& surrogate models**

**PICon GPU**



**Large-scale start-to-end simulations**

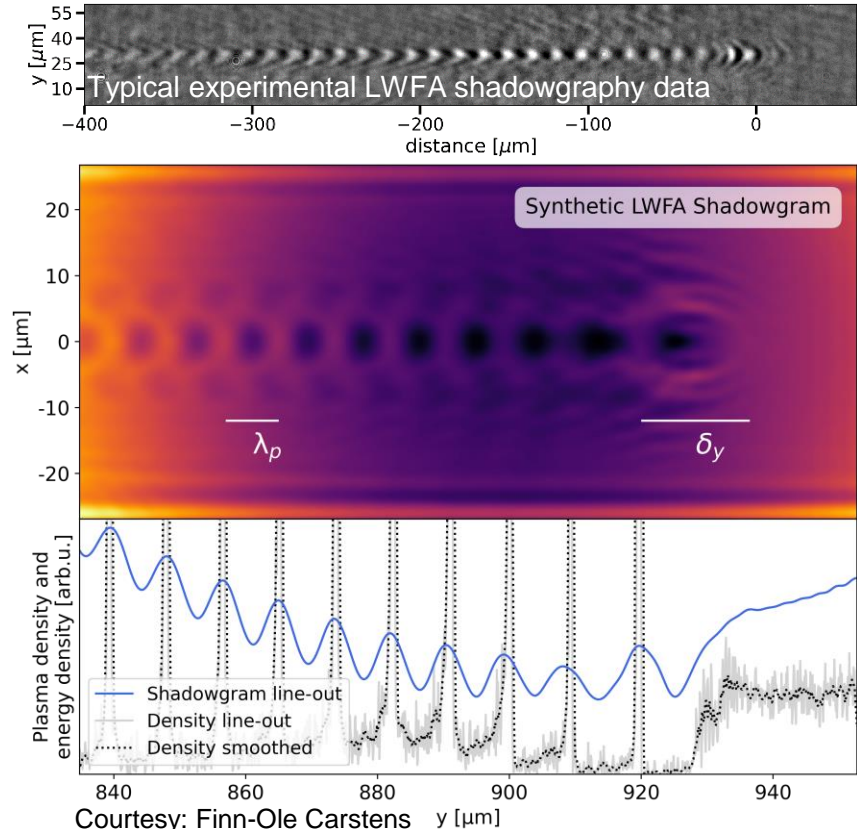
“Matching” the large parameter space of complex experimental designs to simulations needs to be understood and further constrained.

# In-situ synthetic diagnostics give insights into the plasma dynamics

## Acquire shadowgraphy images for comparison to experiment

PICon GPU

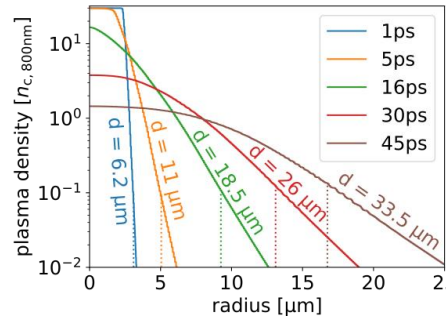
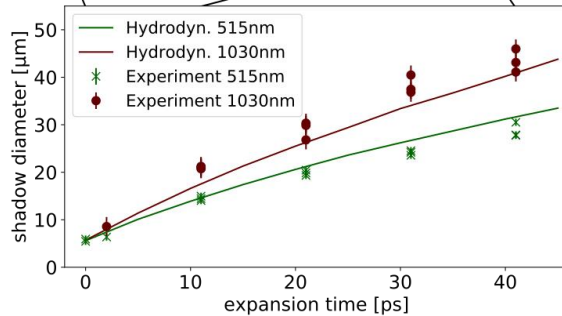
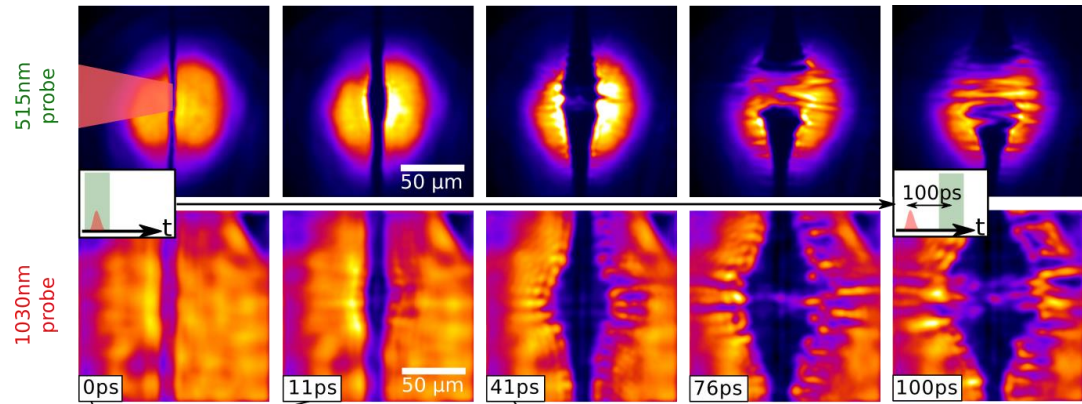
- Probe laser is simulated directly within PIC simulation
- Propagation to camera modeled in-situ by a plugin, i.e. no post-processing needed.
- First results already show characteristics not observed in static simulations (e.g. from strong fields or relativistic particles)





# In-situ synthetic diagnostics give insights into the plasma dynamics

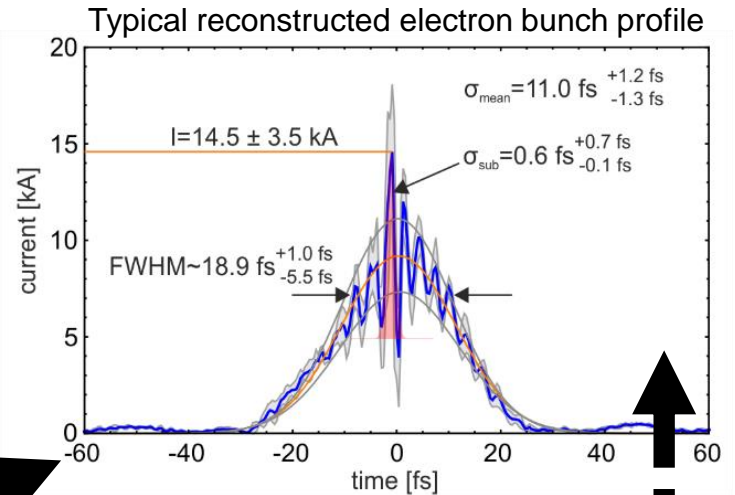
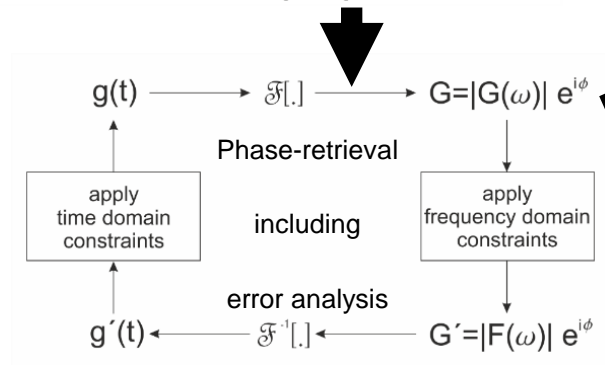
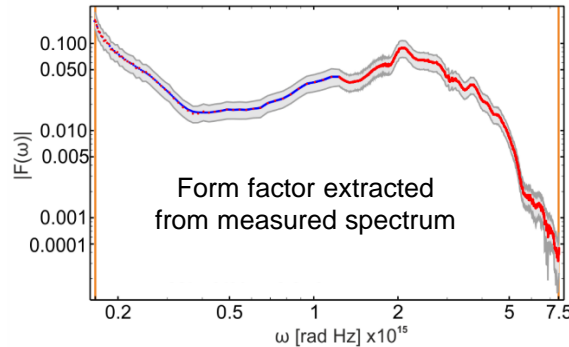
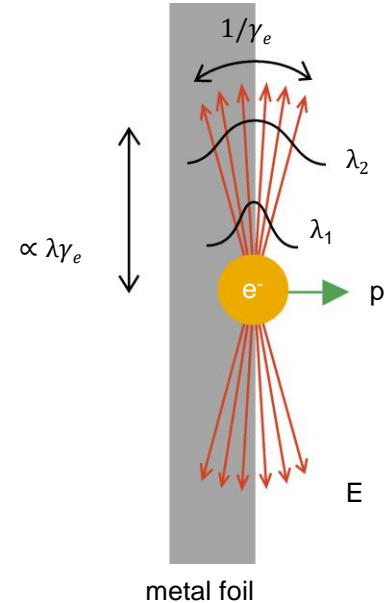
## Acquire shadowgraphy images for comparison to experiment



- Synthetic, in-situ shadowgraphy diagnostics enable analyzing expansion measurements of time-resolved shadowgraphy at near-critical hydrogen jet targets.
- PIC simulations need to ensure correct initial conditions for ion acceleration. In-situ synthetic shadowgram diagnostics now can check for consistency with experiment.

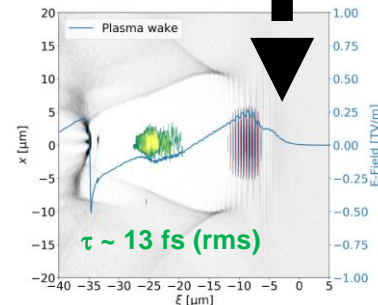
Courtesy: Constantin Bernert

# Experimental data reconstruction to infer the longitudinal electron bunch profile from measured CTR spectra



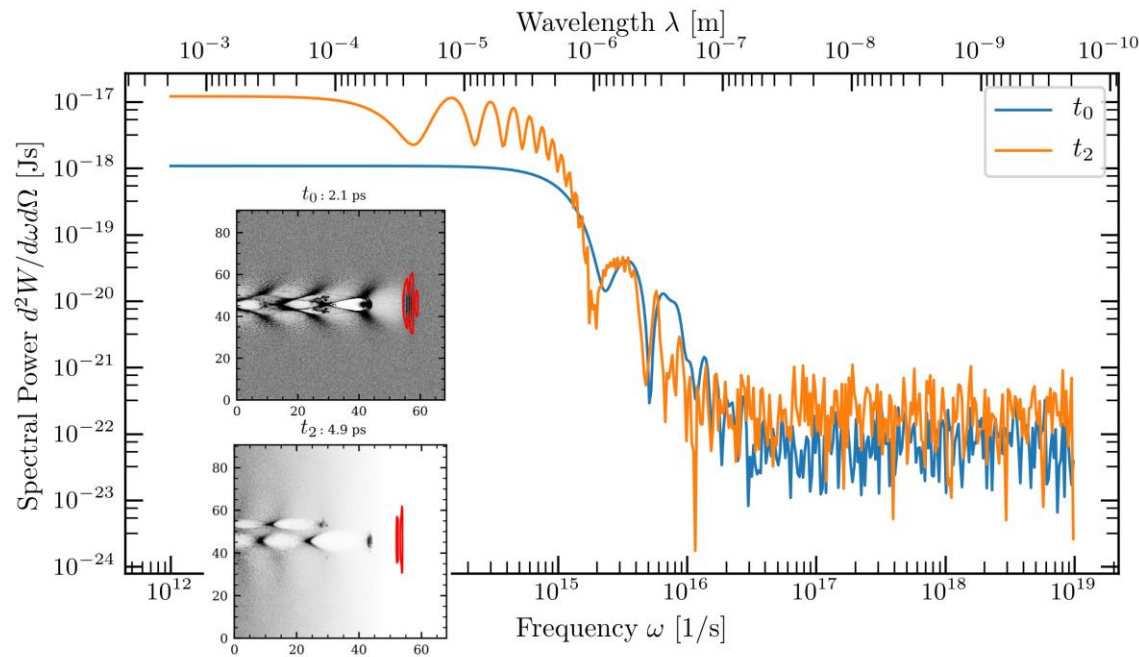
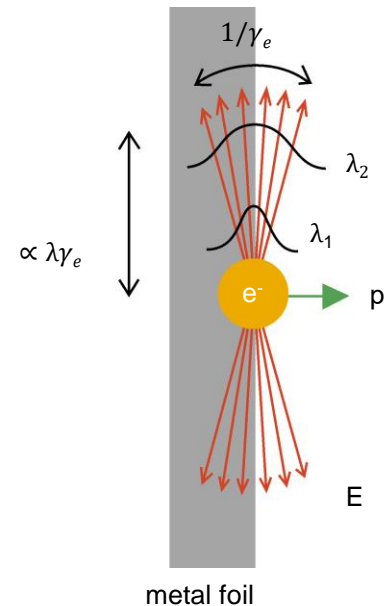
Electron bunch length agrees with 3D-PIC simulations

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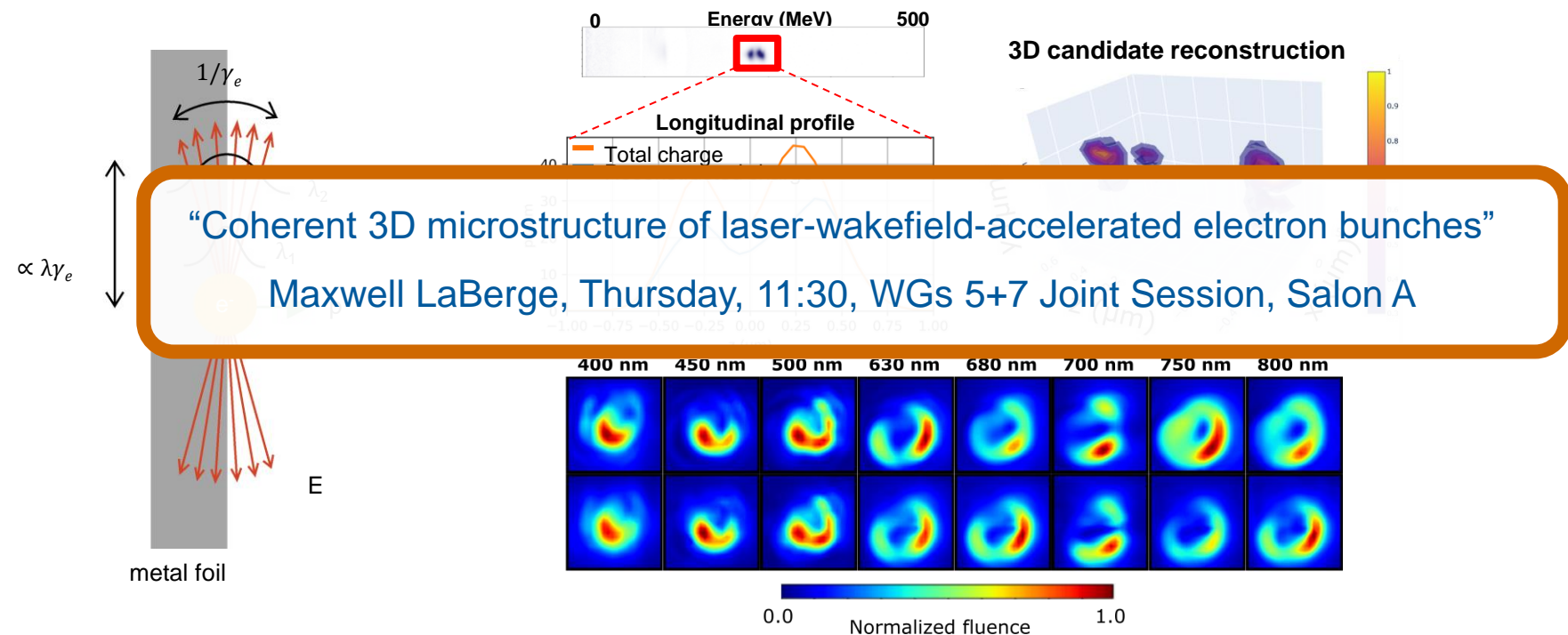
# Synthetic in-situ diagnostics in PIConGPU

## to simulate the measured CTR spectrum



Allows for direct comparison with non-reconstructed measured CTR data.  
Provides high-quality data for testing & refining reconstruction routines.

**Good longitudinal and transverse coherent transition data**  
is key to revealing 3D electron beam structures on a  $\mu\text{m}$  and fs scale

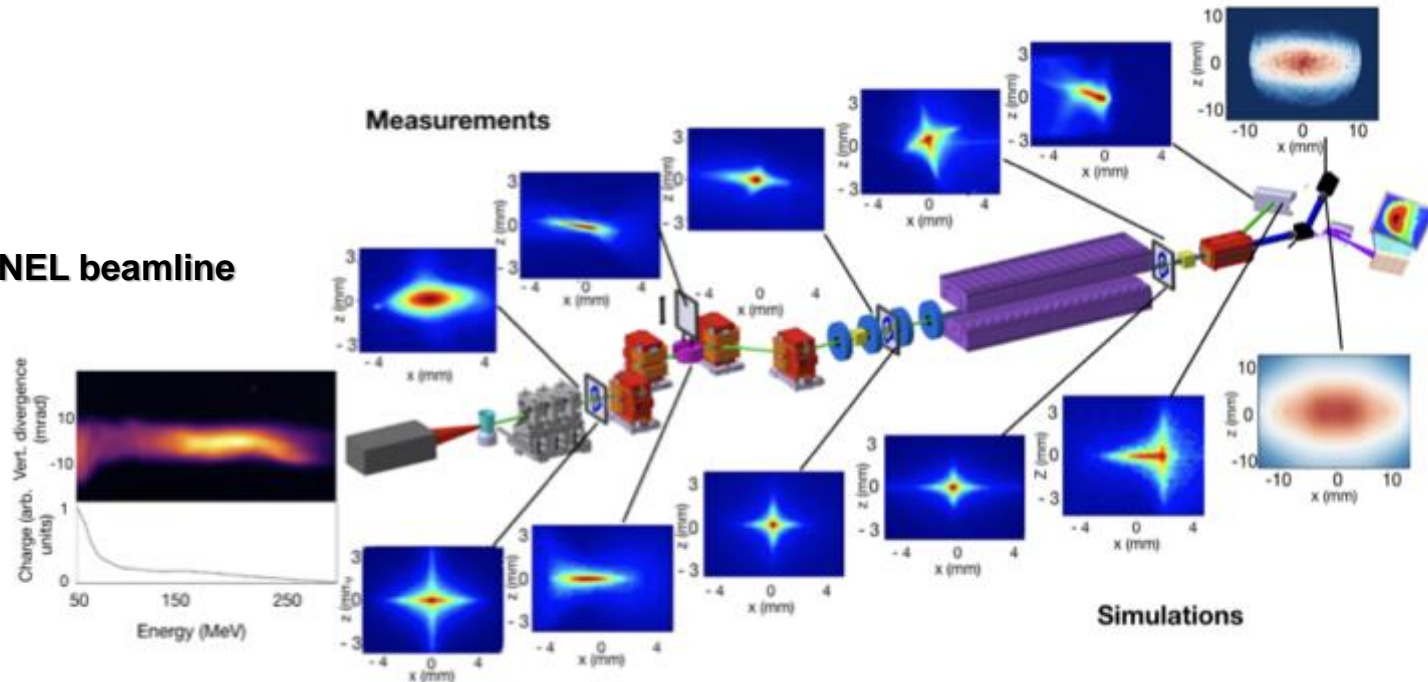




# Data-driven Digital Twin of Free Electron Laser

Surrogate models can provide close to real-time feedback during experiment

## COXINEL beamline

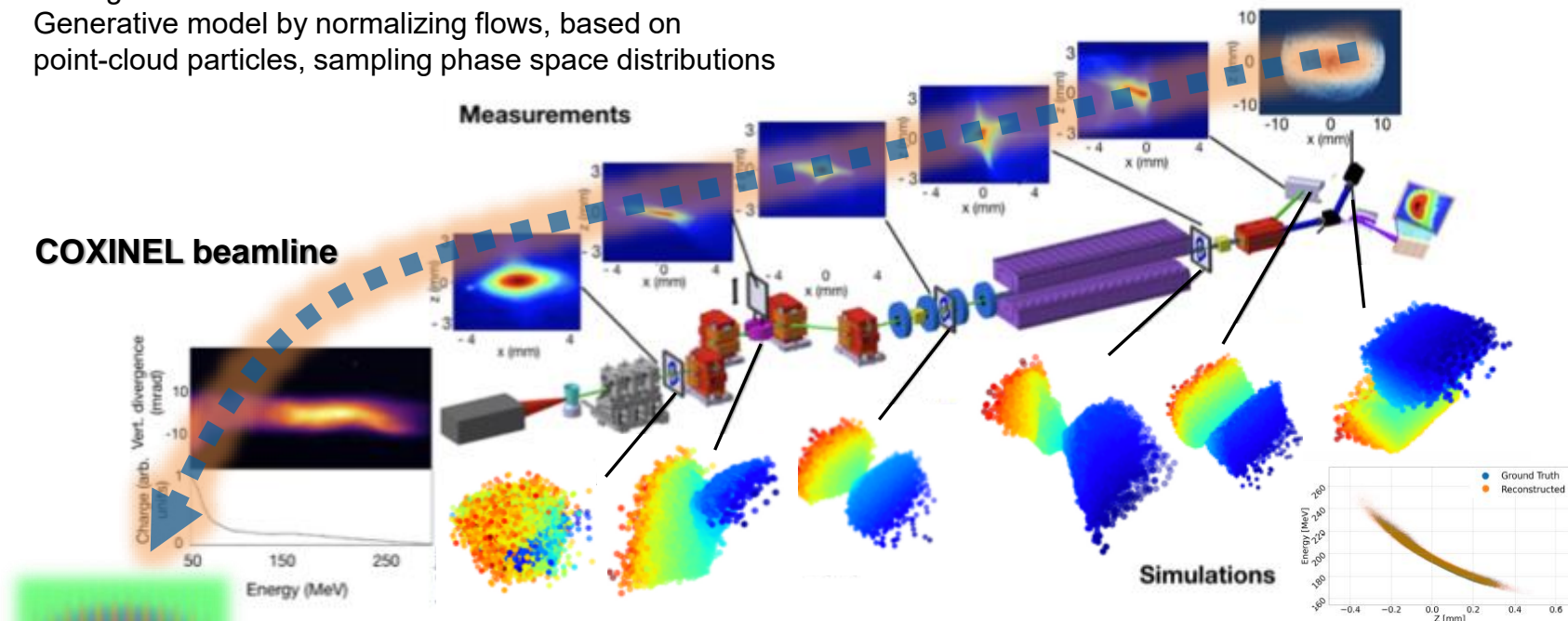


SOLEIL @ HZDR, COXINEL collaboration

# Data-driven Digital Twin of Free Electron Laser

**Surrogate models** can provide close to real-time feedback during experiment

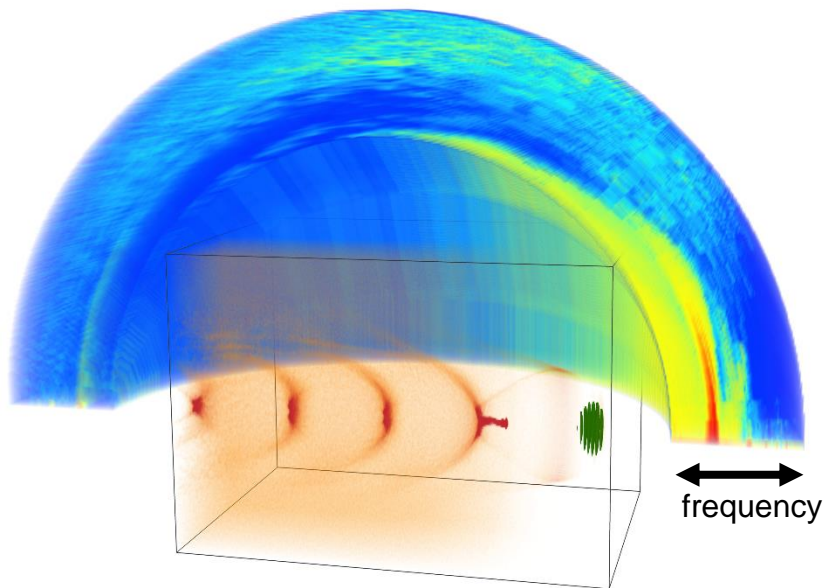
- Surrogate model of COXINEL FEL beamline at HZDR
- Generative model by normalizing flows, based on point-cloud particles, sampling phase space distributions



Willmann et al., NeurIPS 2022 (accepted)  
"Learning Electron Bunch Distribution along  
a FEL Beamline by Normalising Flows"

# An exascale knowledge extraction challenge:

What is the physics origin of emitted plasma radiation by LWFA?

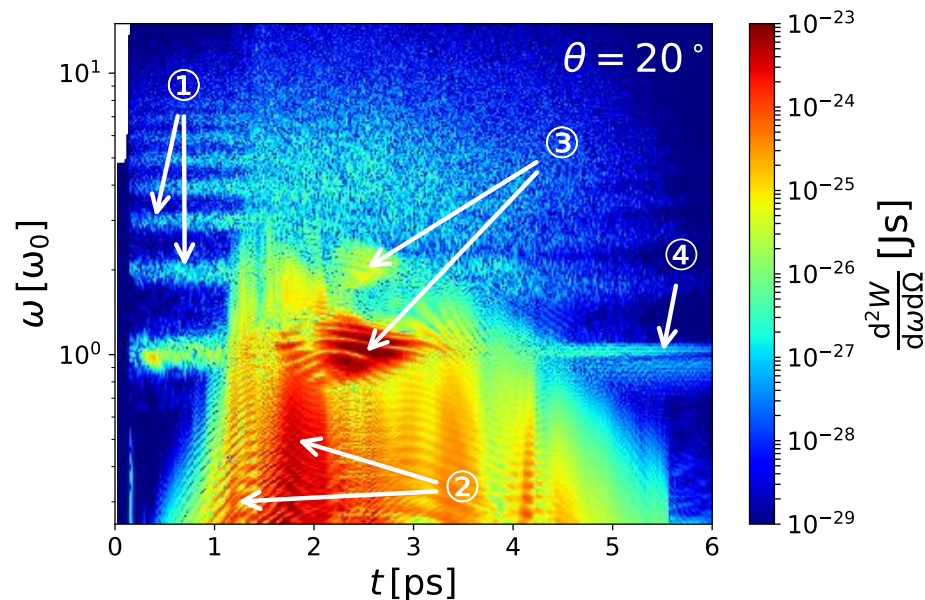
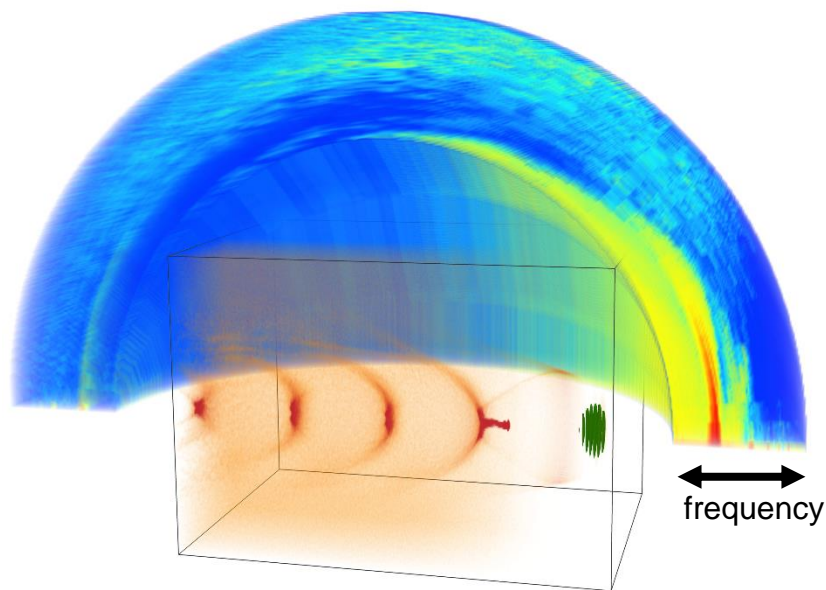


Radiation simulations are computationally very expensive!

- 512 frequencies IR-VIS-UV
- 250 virtual observers on half-dome
- Computed for  $\sim 10^9$  particles at each time step
- **Over 20x more compute intensive than simulation without radiation**

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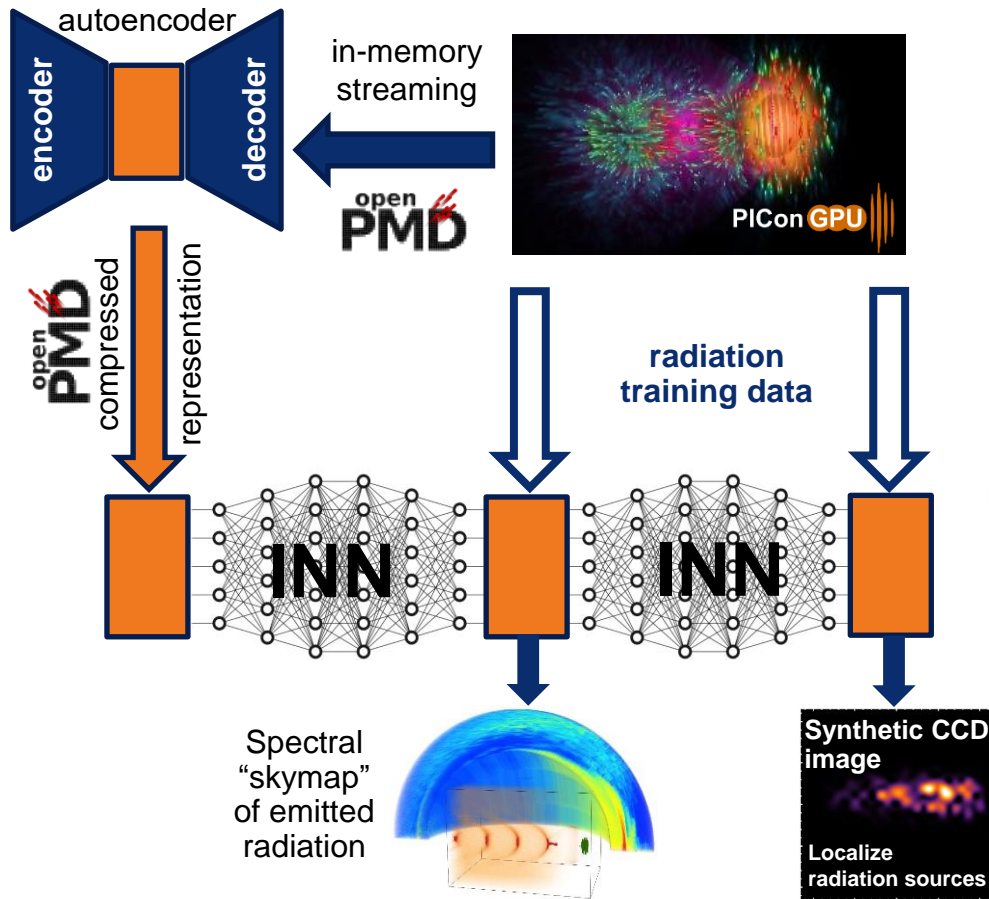
**Current state-of-the art of knowledge extraction:**

PhD student brain power and running lots of reduced-model simulations.

Courtesy: Richard Pausch

# Outlook & vision

– Generate surrogate models on exascale machines to infer the physics origin of plasma radiation emissions

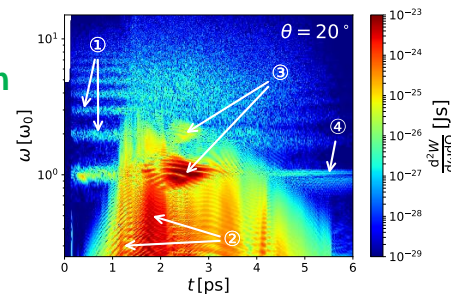


**Data source from PIConGPU**  
9D unstructured point-cloud data  
For each macroparticle every time step:  
Position, Momentum, Acceleration

**Knowledge extraction**



by exploiting INN  
inverse mapping





# Conclusions

- **PIConGPU is performance portable and scales up to exascale machines (Summit and Frontier).**  
“Matching” the large parameter space of complex experimental designs to simulations needs to be understood and further constrained.
- **In-situ synthetic diagnostics** in particle-in-cell simulations (PIConGPU), such as shadowgraphy, facilitate direct and quantitative comparison to experimental data.
- **Fast inverse-problem solving** including uncertainty quantification requires access to suitable compute resources. Such experiments often start with a lack of high-quality ground truth-data, PIC simulations can provide large quantities of training & test data.
- **Surrogate models** trained by extensive simulations can provide close to real-time feedback during experiment.