



Contribution ID: 612 Contribution code: WEPA102

Type: **Poster Presentation**

Physics-constrained 3D convolutional neural networks for electrodynamics of relativistic charged particle beams

Wednesday, 10 May 2023 16:30 (2 hours)

We present a physics-constrained neural network (PCNN) approach to calculating the electromagnetic fields of intense relativistic charged particle beams via 3D convolutional neural networks. Unlike the popular physics-informed neural networks (PINNs) approach, in which soft physics constraints are added as part of the network training cost function, our PCNNs respect hard physics constraints, such as $\nabla \cdot \mathbf{B} = 0$, by construction. Our 3D convolutional PCNNs map entire large ($256 \times 256 \times 256$ pixel) 3D volumes of time-varying current and charge densities to their associated electromagnetic fields. We demonstrate the method on space charge dominated, relativistic (5 MeV), short (hundreds of fs), high charge (2 nC) electron beams, such as those in the injector sections of modern free electron laser and plasma wakefield accelerators. We show that the method is accurate, respects physics constraints, and that the trained 3D convolutional PCNNs perform electromagnetic calculations orders of magnitude faster than traditional solvers which require a $O(N^2)$ process for calculating the space charge fields of intense charged particle beams.

Funding Agency

U.S. Department of Energy (DOE), Office of Science, Office of High Energy Physics contract number 89233218CNA000001 and the Los Alamos National Laboratory LDRD Program project 20220074DR.

Footnotes

I have read and accept the Privacy Policy Statement

Yes

Primary author: Dr SCHEINKER, Alexander (Los Alamos National Laboratory)

Presenter: Dr SCHEINKER, Alexander (Los Alamos National Laboratory)

Session Classification: Wednesday Poster Session

Track Classification: MC5: Beam Dynamics and EM Fields: MC5.D13: Machine Learning