

Beam-Dust Studies of SuperKEKB Relevant to FCC-ee

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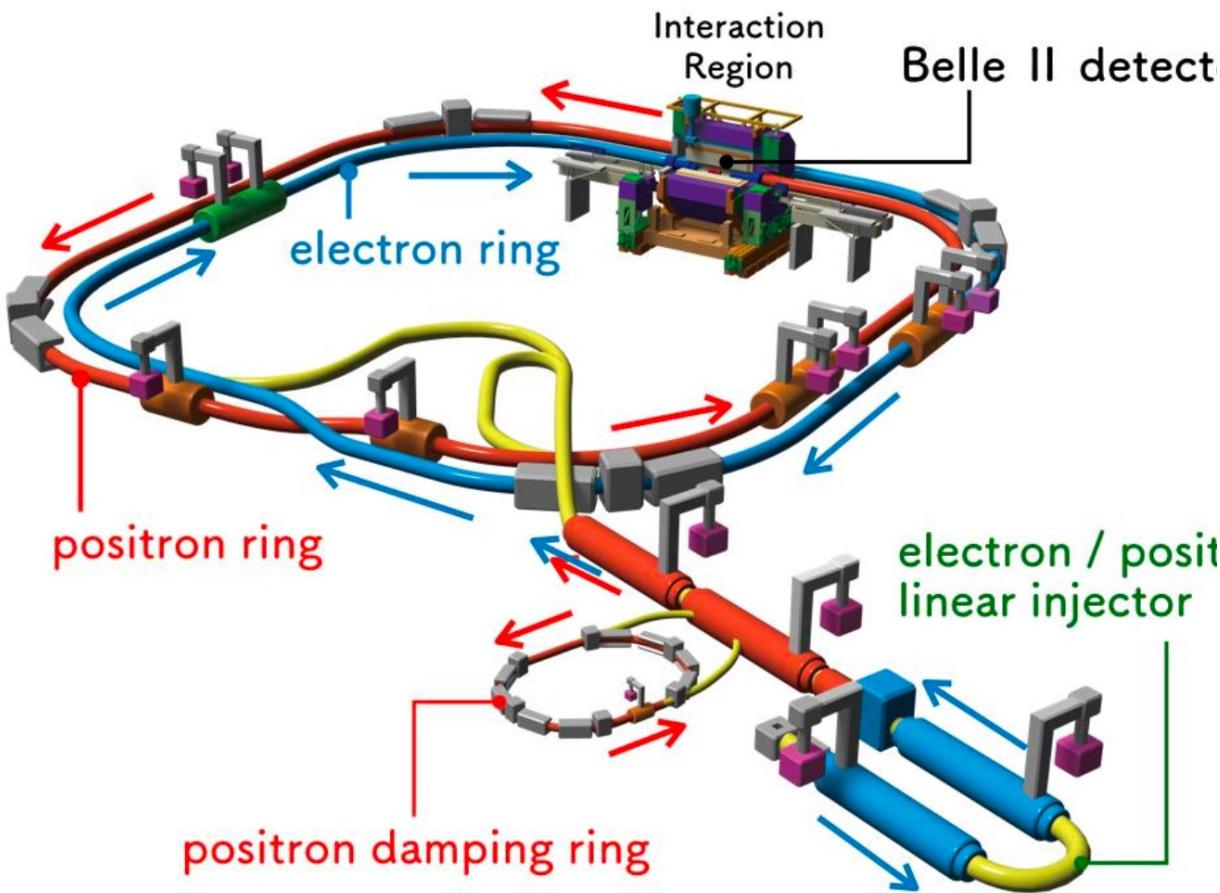


Fourth US FCC meeting
SLAC, Sep. 8, 2026

Overview

- Background: beam-dust models of SBL at SuperKEKB
- Challenges and goals of this study
- Simulation Tools: HIPACE++ and ImpactX
- Multi-turn modeling of beam interaction with dust plasma
 - single-bunch studies
 - multi-bunch studies
- Future Work

SuperKEKb Layout



		e^+	e^-
		LER	HER
Beam energy	E [GeV]	4.0	7.0
Circumference	C [m]	3016	
Harmonic number	h	5120	
RF frequency	f_{RF} [MHz]	509	
Revolution frequency	f_0 [kHz]	100	
Minimum bunch spacing	[ns]	4	
Beam current	I [A]	1.632	1.259
Number of bunches	n_b	2346	
Charge per 1 mA bunch	[nC]	10	
Horizontal beta function at IP	β_x^* [mm]	60	60
Vertical beta function at IP	β_y^* [mm]	1.0	1.0
Horizontal tune	ν_x	44.525	45.531
Vertical tune	ν_y	46.589	43.599
Vertical beam size at IP	σ_y^* [nm]	~ 300	
Bunch length	σ_z [mm]	~ 6	
Instantaneous Luminosity	L [cm $^{-2}$ s $^{-1}$]	5.1×10^{34}	5.319×10^{34}

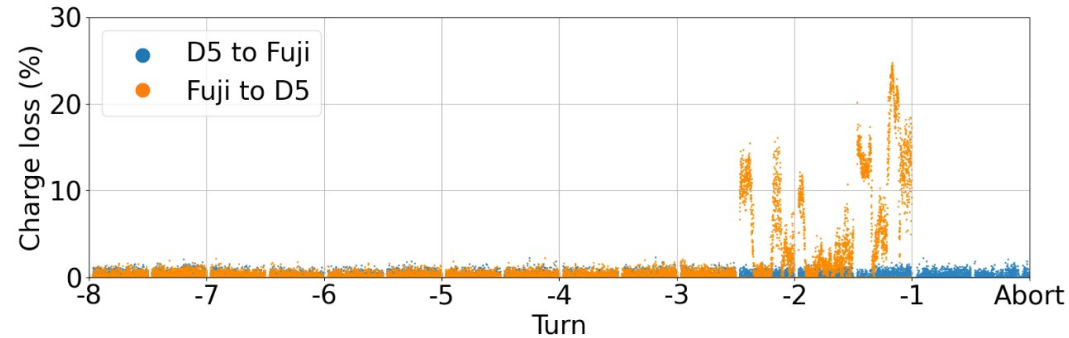
Table 1: Main parameters of the SuperKEKB main ring. Parameters below the beam current correspond to the conditions at the time of the highest recorded luminosity on December 27, 2024 [9].

- B-factory for CP violation studies and beyond Standard Model searches
- World record luminosity $5.319 \times 10^{34} \text{ cm}^{-2}\text{s}^{-1}$ (June 2026) – still well below (1/15) design value

Beam-dust models of sudden beam loss

Rapid loss ~ 1 turn ($10\mu\text{s}$)

Vertical beam size
blowup ~ 10 turns



(*SBL event Oct. 28, 2024 in LER)
Most 2024 events in LER

Current understanding and mitigation:

K. Ohmi, H. Fukuma, and S. Terui, PRAB **29**, 021002 (2026)

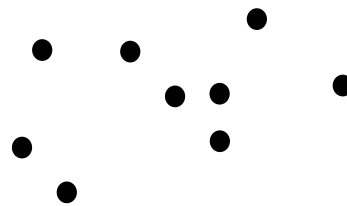
*R. Nomaru, G. Mitsuka, and L Ruckman, JINST 21 03, T03008 (2026)

S. Terui *et al*, Phys. Rev. Accel. Beams (accepted 28 Aug, 2026)

positron bunches



dust (graphite or silicon)



- ionization and heating
- fission
- evaporation/sublimation

dust plasma

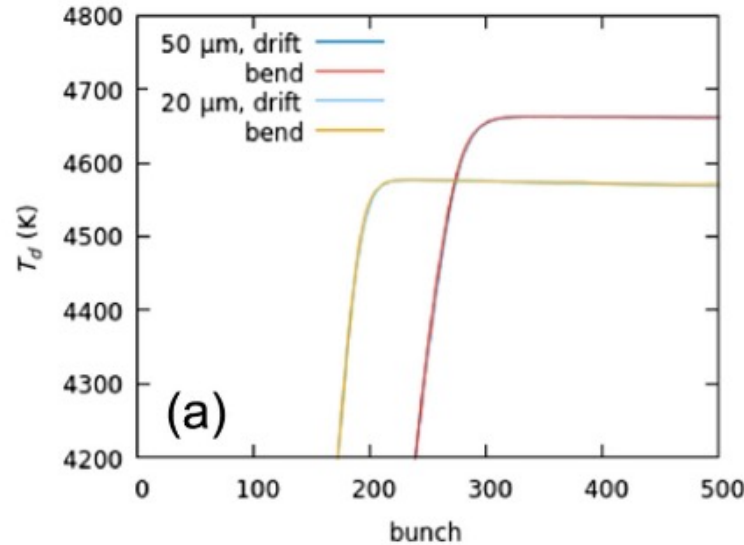


- Cleaning of VacSeal in MO-type vacuum flanges mitigated most SBL events in the LER.
- In 2026, SBLs are seen primarily in the HER when operating at high current.

Beam-dust models of sudden beam loss

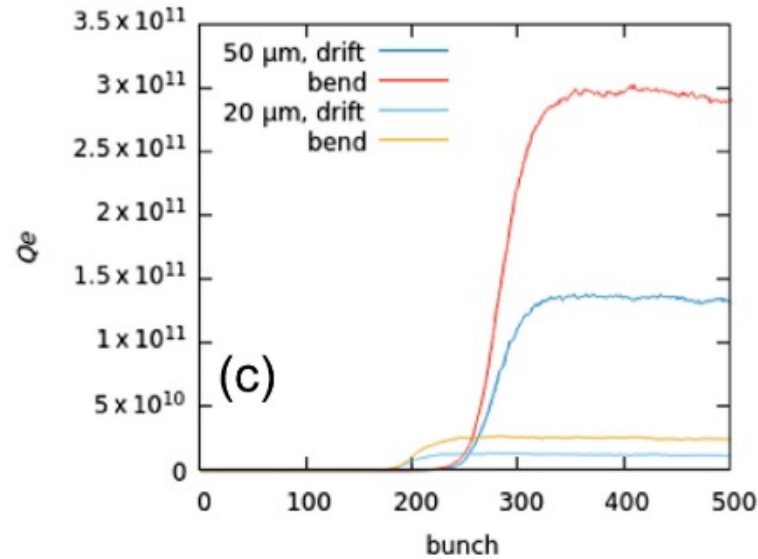
K. Ohmi *et al* (2026):

Fig. 4a



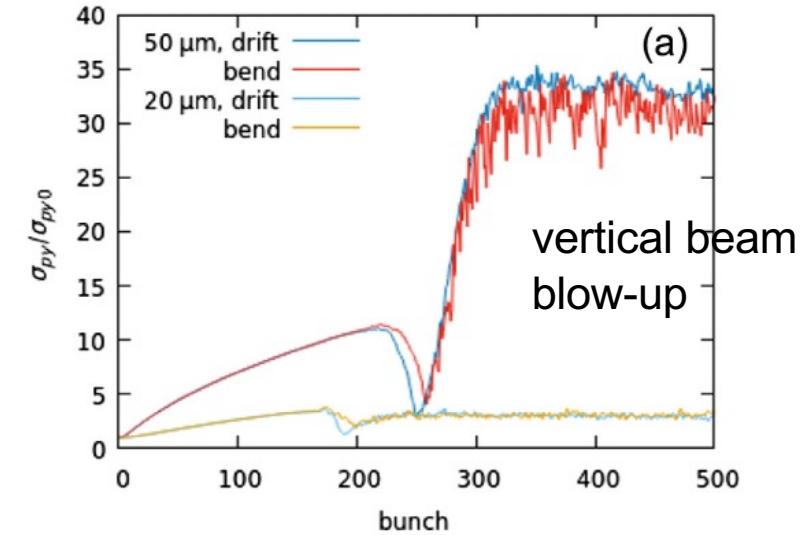
Temperature of dust/plasma

Fig. 5c



Electron charge in dust plasma

Fig. 6a



Bunch vertical momentum spread induced by plasma passage

- modeling of bunch interaction with graphite dust particle of size 20 or 50 μm (in either a drift or a bend)
- dust sublimation point is reached after 130 (20 μm) - 190 (50 μm) bunch passages in the train
- dust plasma reaches $\sim$ steady state *100 bunch passages after sublimation*

Challenges and goals of this study

Challenges:

Complexity of physical processes, including dust ionization/sublimation and plasma processes.

Modeling the interaction of an entire bunch train (2,346 bunches) with dust x ~ 10 turns is computationally infeasible. [K. Ohmi] An approximate scheme is used, using 100 bunch passages per turn (after sublimation).

The motion of electrons and ions during the interval between bunches (bunch gap of 4 ns) is challenging to resolve (plasma frequencies as large as ~ 284 GHz). [K. Ohmi] “Interactions within the dust plasma itself, including plasma oscillations, will be addressed in future work.”

The plasma is treated using a 2D electrostatic particle-in-cell model. [K. Ohmi] “... Since the electrons and ions condensing around the beam are the primary focus, the two-dimensional approximation is considered appropriate.”

To analyze beam loss accurately, it is necessary to track the particle distribution in phase space through the accelerator using a beam tracking code and examine losses at the collimator (code coupling).

Challenges and goals of this study

Previous work:

Small-scale study of single bunch+plasma interaction at typical SuperKEKB parameters (positron LER)

Goals of this work:

- demonstrate an integrated multi-bunch, multi-turn workflow on GPU
- coupling of plasma + lattice/beam transport codes
- contribute to validation/benchmarking of previous simulation efforts
- (longer term) investigate the size of neglected effects (e.g. 3D effects, plasma oscillations, nonlinear lattice transport)

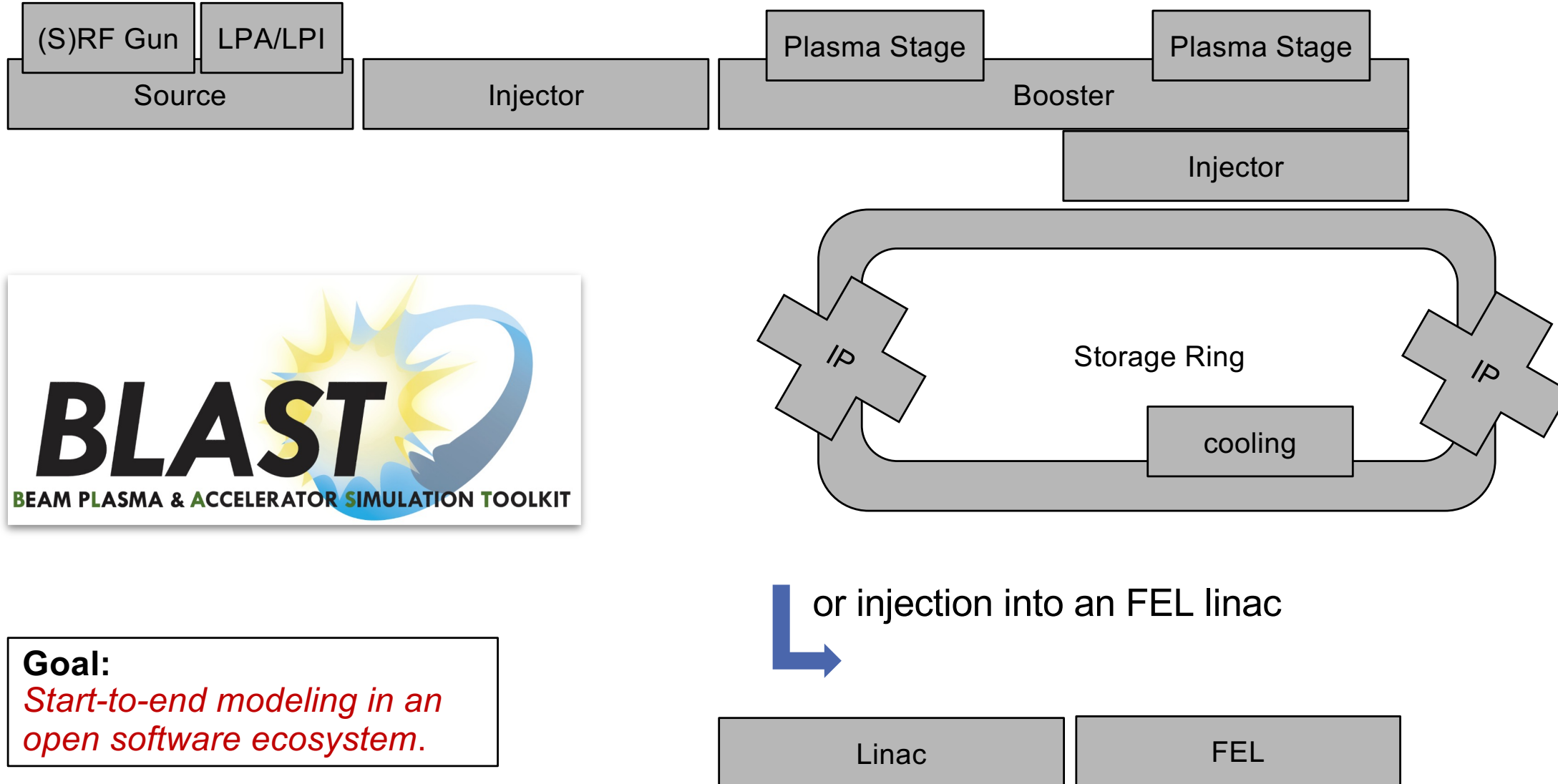
Notes:

- We track through the sublimated dust plasma, including ionization processes
- Initial heating and interaction with the dust particle are not treated

- Simulation Tools

Beam Plasma and Accelerator Simulation Toolkit: integrated beam physics across accelerator subsystems

Imagine a future, **hybrid particle accelerator**, e.g., with conventional and plasma elements.



Goal:
*Start-to-end modeling in an
open software ecosystem.*

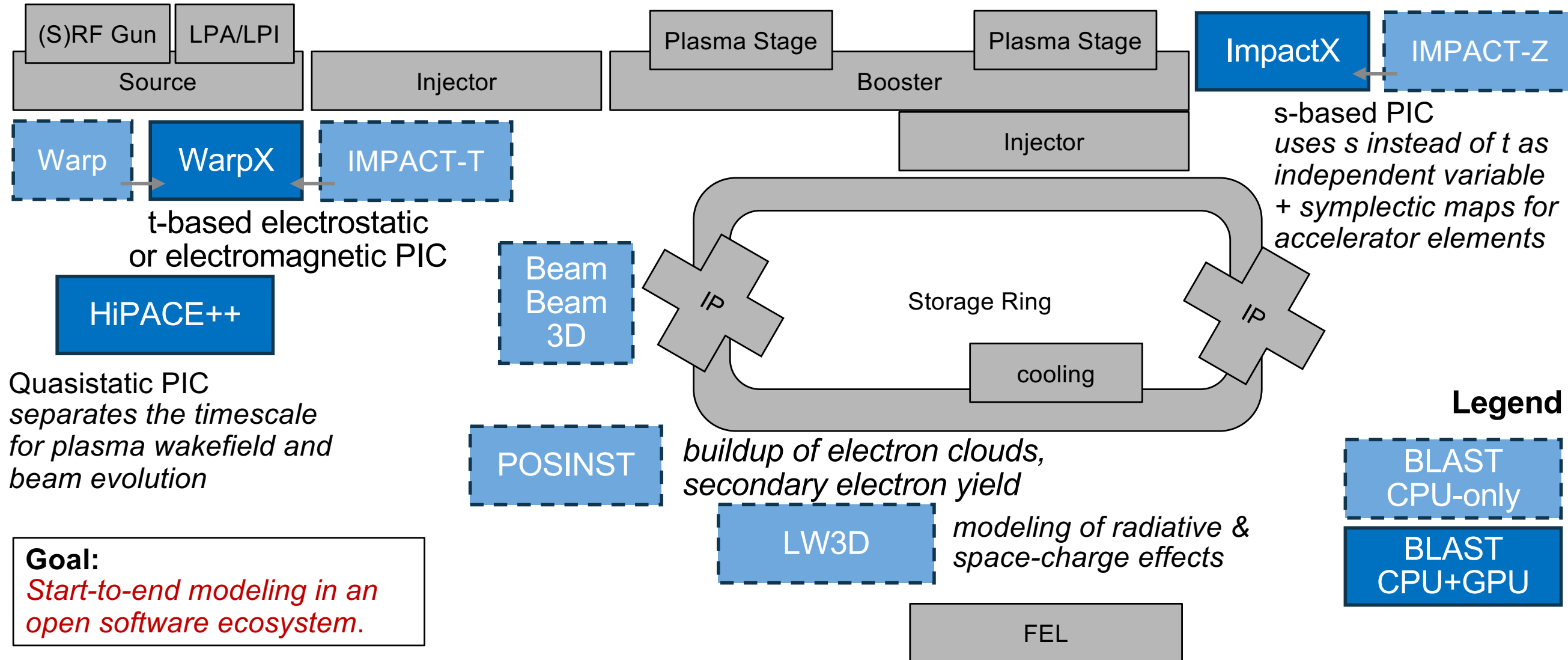
Legend

BLAST
CPU-only

BLAST
CPU+GPU

Beam Plasma and Accelerator Simulation Toolkit: integrated beam physics across accelerator subsystems

Imagine a future, **hybrid particle accelerator**, e.g., with conventional and plasma elements.



Software stack for the BLAST ecosystem

User
Interfaces

Python Bindings & Particle-In-Cell Modeling Interface (PICMI)

Apps

WarpX

HiPACE++

ImpactX

...

Libraries

pyAMReX

ABLASTR
common PIC components

PICSAR
QED modules

Computer
Science

AMReX
Containers, Communication,
Portability, Utilities

openPMD I/O,
streaming, in situ:
ADIOS/HDF5

Math
LinAlg., FFT

Vendor

MPI multi-node comm.

CUDA, OpenMP, DPC++, HIP



Desktop
to
HPC

HIPACE++: quasi-static PIC on GPU for plasma acceleration

- Supports multiple beam and plasma species
- Built on top of AMReX framework (LBNL)
Data structures and communications
Performance-portability layer: supports NVIDIA & AMD GPUs
- HPC programming standards
 - Documented, **open-source**
 - Automated testing (CI)
- openPMD I/O

$$\nabla_{\perp}^2 \psi = -\frac{1}{\epsilon_0} \left(\rho - \frac{1}{c} j_z \right)$$

$$E_x - c B_y = -\partial_x \psi$$

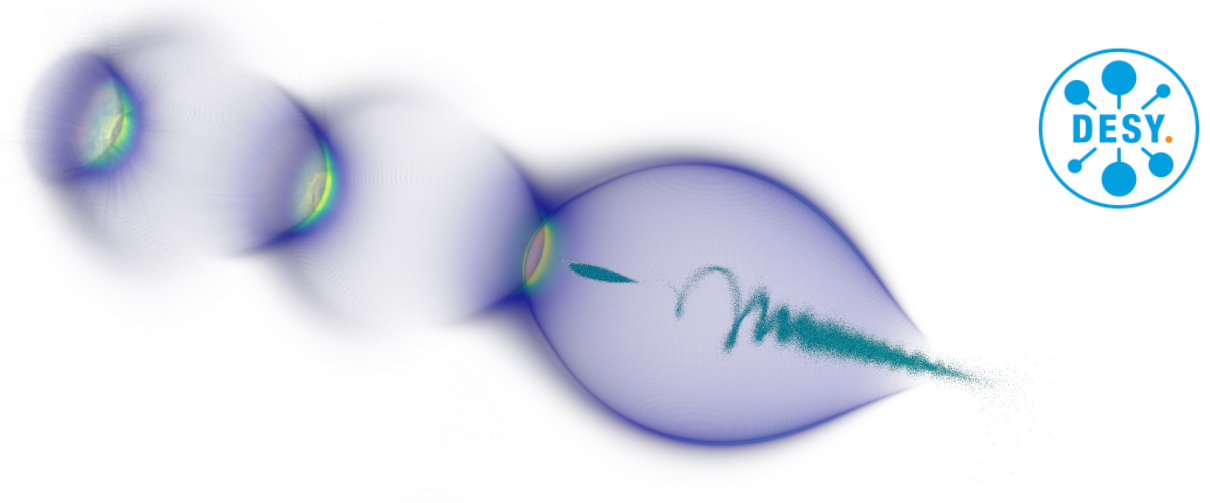
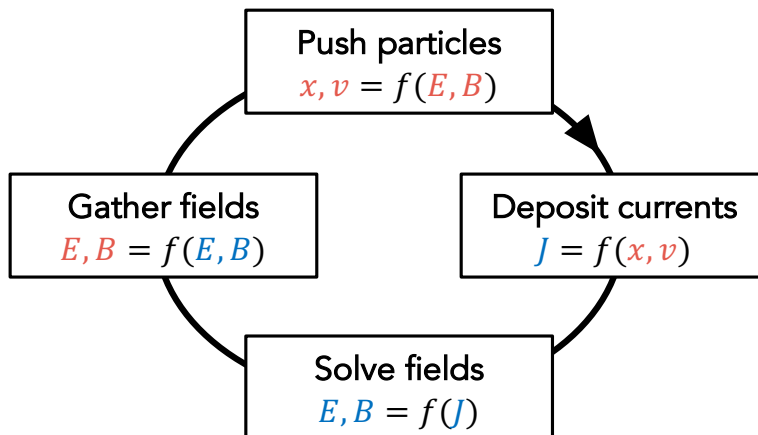
$$E_y + c B_x = -\partial_y \psi$$

$$\nabla_{\perp}^2 E_z = c \mu_0 (\partial_x j_x + \partial_y j_y)$$

$$\nabla_{\perp}^2 B_x = \mu_0 (-\partial_y j_z + \partial_z j_y)$$

$$\nabla_{\perp}^2 B_y = \mu_0 (\partial_x j_z - \partial_z j_x)$$

$$\nabla_{\perp}^2 B_z = \mu_0 (\partial_y j_x - \partial_x j_y)$$



HiPACE++

3D quasi-static PIC code

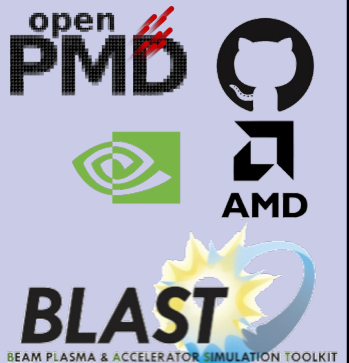
Developed by DESY & LBNL

PI: Maxence Thévenet

<https://github.com/Hi-PACE/hipace>

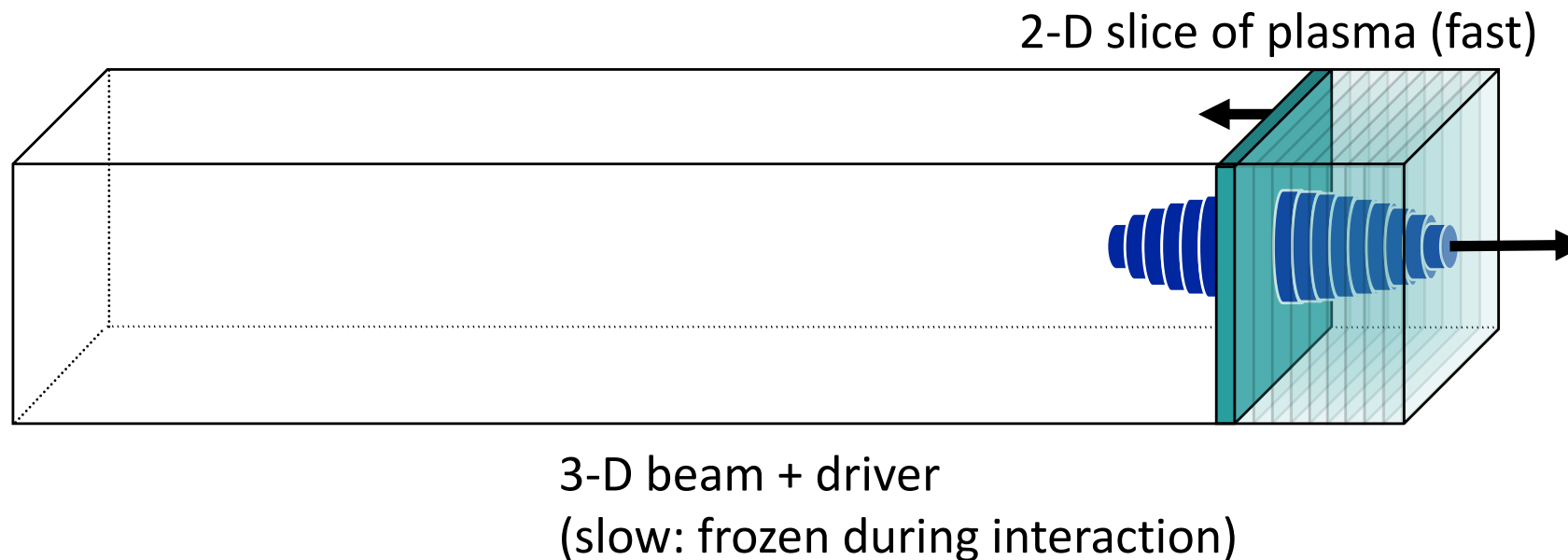
Doc: <https://hipace.readthedocs.io>

Diederichs et al., CPC 278, 108421 (2022)



Quasistatic approximation uses separation of scales

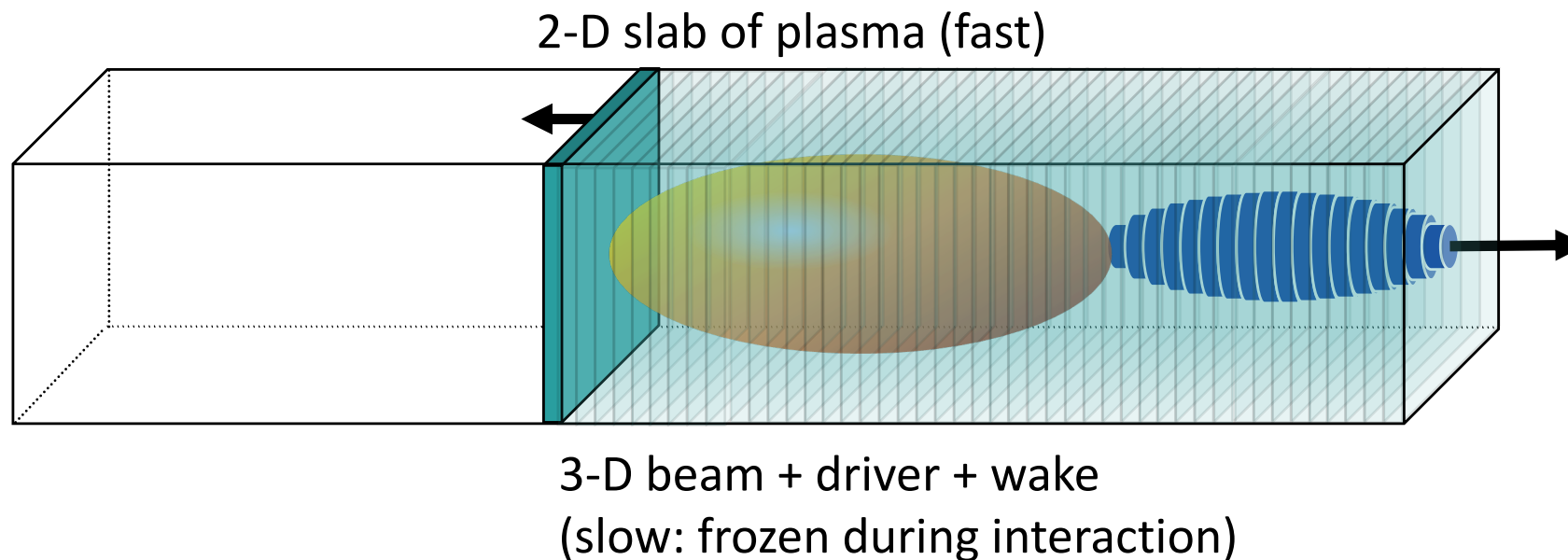
1. **Fast time scale** - 2-D slice of plasma is **stepped “backward”** (with small steps) through the beam field and its self-field (solving iteratively 2-D Poisson-like equations at each step).



- Quasi-static approximation (QSA)
Sprangle, P. et al., *PRL* 64.17 (1990): 2011

Quasistatic approximation uses separation of scales

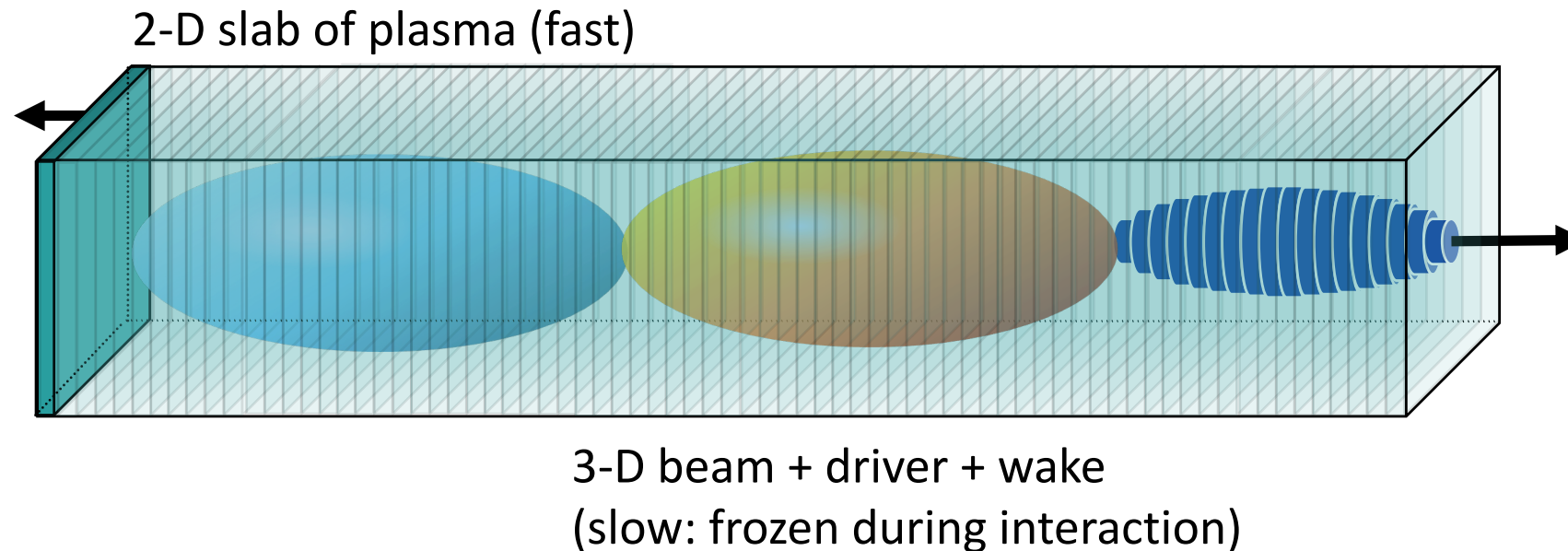
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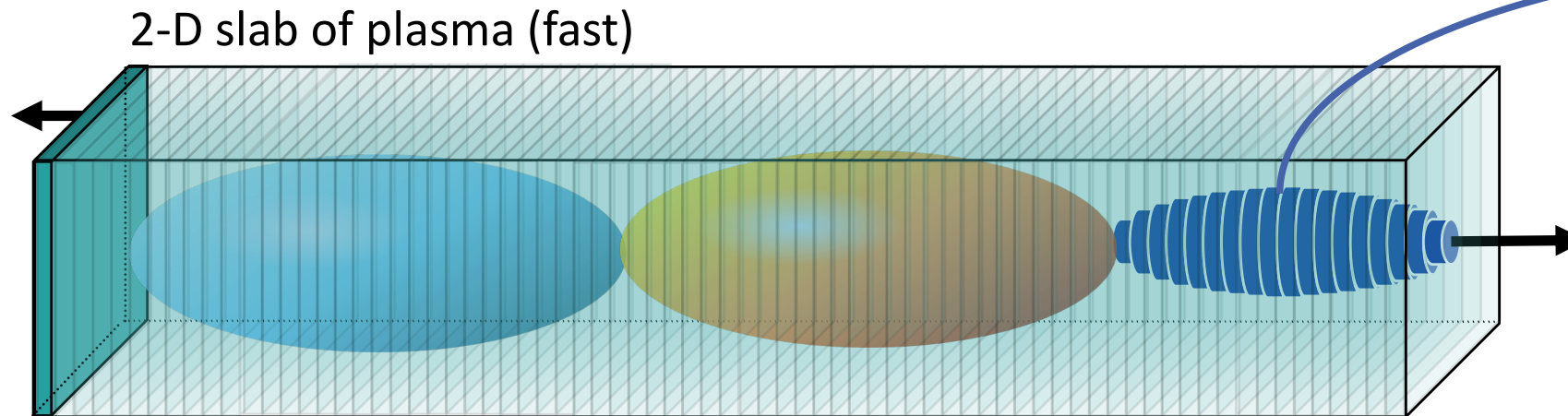
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- Quasi-static approximation (QSA)
Sprangle, P. et al., *PRL* 64.17 (1990): 2011

Quasistatic approximation uses separation of scales

- repeat
1. **Fast time scale** - 2-D slice of plasma is **stepped “backward”** (with small steps) through the beam field and its self-field (solving iteratively 2-D Poisson-like equations at each step).
 2. **Slow time scale** - 3-D beam is pushed far down the plasma with large time steps.



2-D slab of plasma (fast)

3-D beam + driver + wake
(slow: frozen during interaction)

Example

Costs: 1 hour on a laptop
(NVIDIA RTX2070)
in FP32, **1024×1024×1024** grid
points, 10^7 beam particles, **300**
time steps

- Quasi-static approximation (QSA)
Sprangle, P. et al., *PRL* 64.17 (1990): 2011

ImpactX: GPU-, AMR- & AI/ML-Accelerated Beam Dynamics

Physical Model based on IMPACT-Z

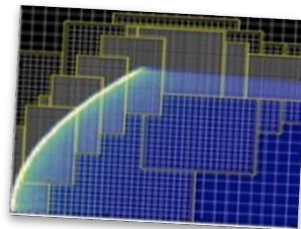
- s-based tracking with symplectic maps
- detailed RF cavity models and standard magnetic elements (w/soft-edge models)
- exact nonlinear maps for sbends, drifts
- translation/rotation errors and apertures
- radiation (CSR/ISR) and spin tracking

Space Charge Model

- space charge included using a split-operator approach
- 3D, 2D, and 2.5D models
- Multi-Level, Multi-Grid Poisson solver

Triple Acceleration Approach

- GPU support
- Mesh Refinement (AMReX)
- AI/ML & Data Driven Models



User-Friendly

- single-source C++, full Python control
- fully tested
- fully documented

💡 **Same Script**
CPU/GPU & MPI

Multi-Node parallelization

- MPI: domain decomposition
- dynamic load balancing (in dev.)

On-Node Parallelization

- GPU: CUDA, HIP and SYCL
- CPU: OpenMP

Scalable, Parallel I/O

- openPMD
- in situ analysis



github.com/BLAST-ImpactX/impactx

- Multi-pass modeling of beam interaction with dust plasma

Beam and lattice parameters (modeled in HIPACE++)

Positron bunch

beam energy: 4 GeV
positrons per bunch: 2×10^{10}
geometric rms emittances: $\varepsilon_x = 5.56$ nm, $\varepsilon_y = 29.9$ pm
RMS bunch length: 6.5 mm
RMS relative momentum spread: $\sim 0.08\%$

*Parameters targeting sudden
beam loss event March 2024
(SuperKEKb - LER)*

Setup and parameters suggested
by F. Zimmerman

Lattice focusing

Twiss functions: $\beta_x = \beta_y = 10$ m, $\alpha_x = \alpha_y = 0$
Tunes (2023): 44.540 (x) / 46.601 (y) / -0.0235 (synchrotron)

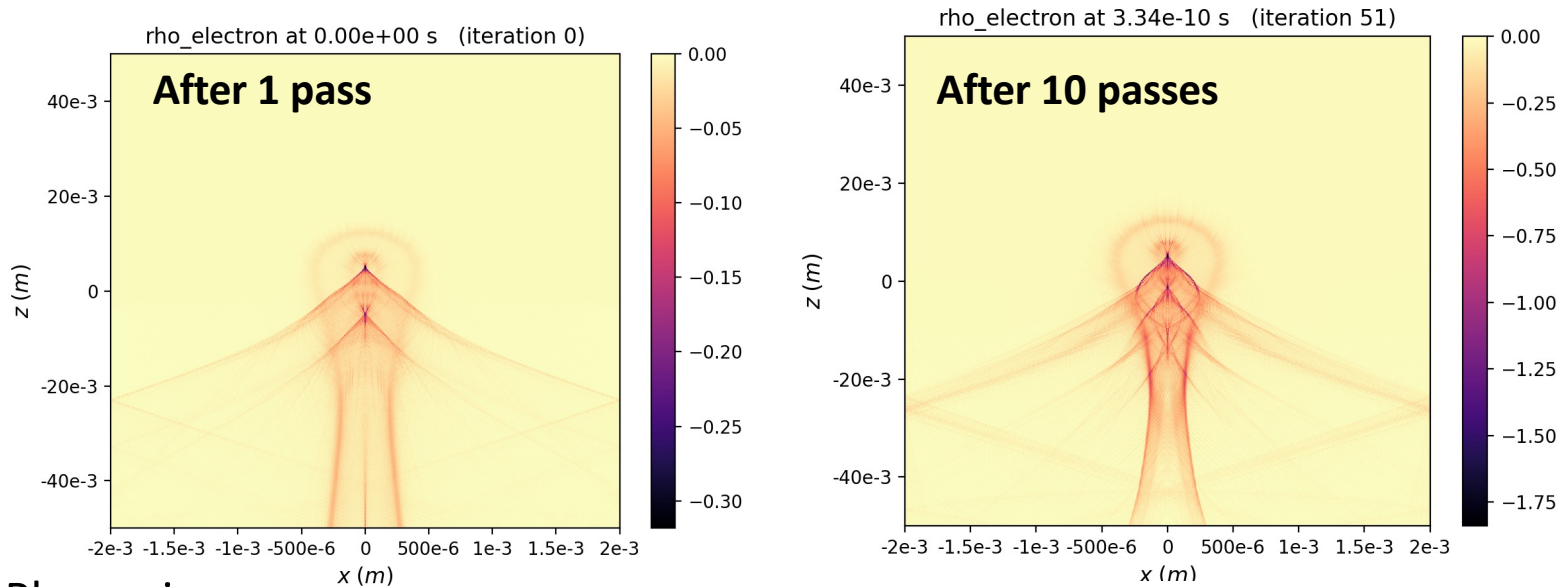
Dust plasma / e-cloud

Dust plasma over 10 cm length – for several values of density in the range $10^{11} - 10^{17}$ / m³
Dynamical effects on a single pass become visible at plasma e- densities $\sim 10^{16}$ / m³

Model multiple passes of a bunch interacting with the same plasma (via one-turn map)

Plasma evolution for a 0.1 m plasma of density $10^{17}/\text{m}^3$

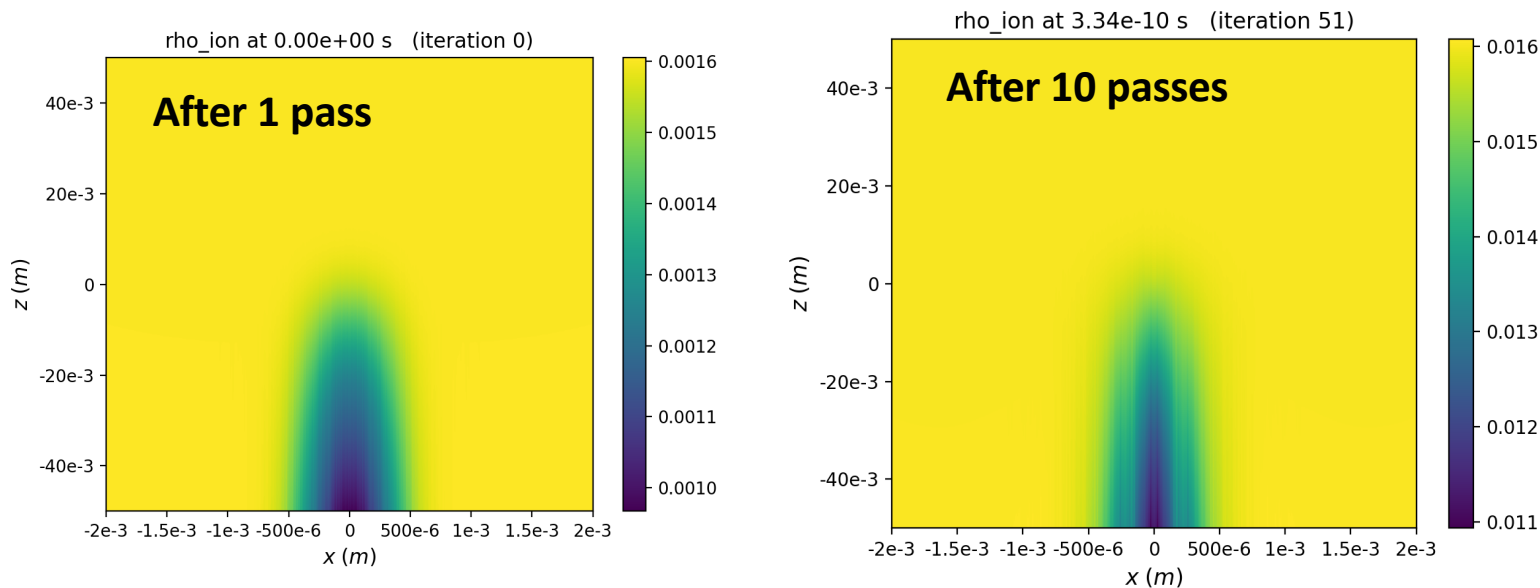
Plasma electrons



10 passes of the
positron bunch

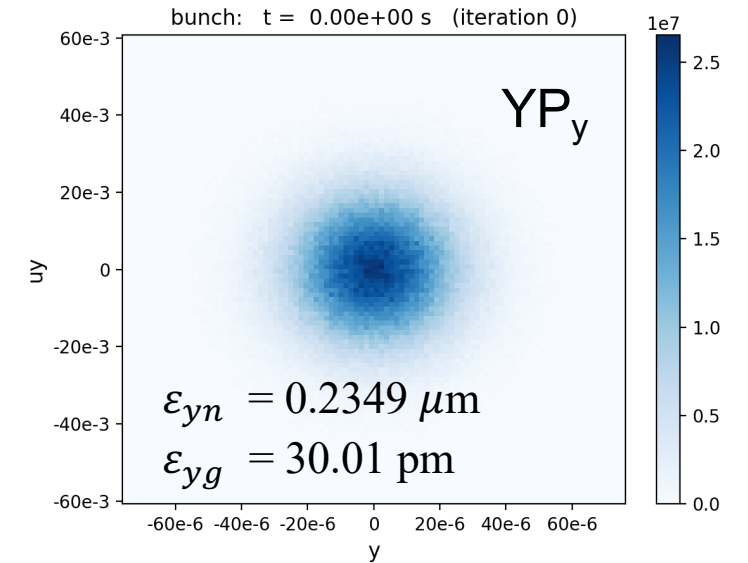
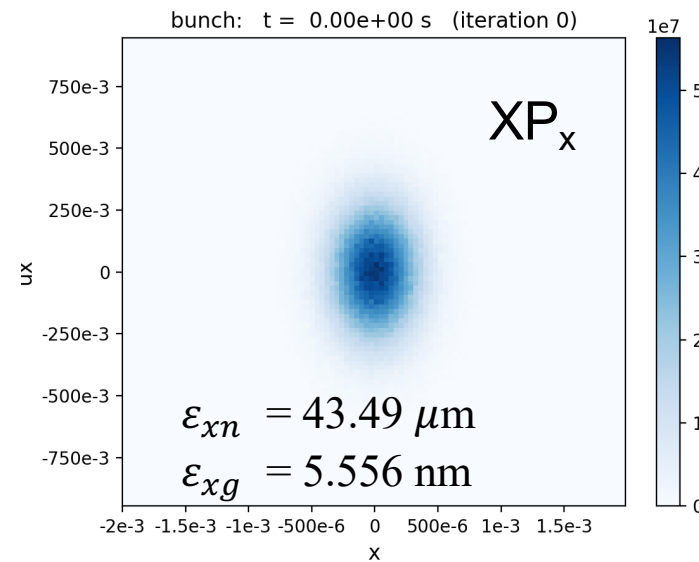
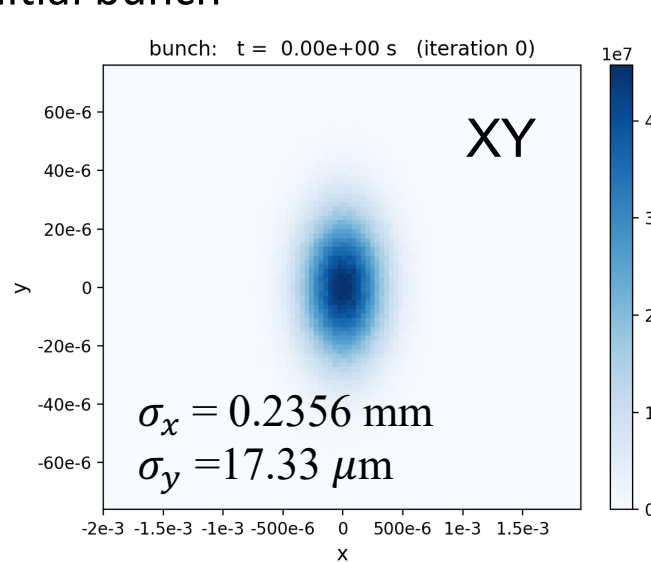
Visualization using
openPMD-viewer
(Jupyter notebook)

Plasma ions

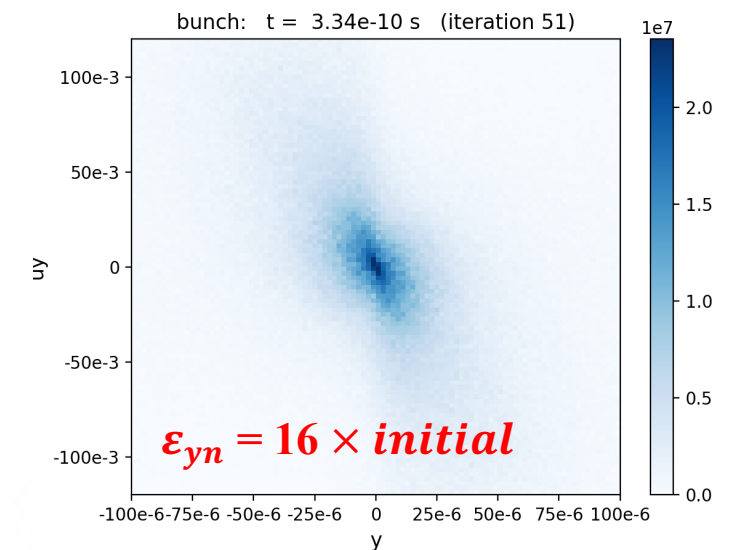
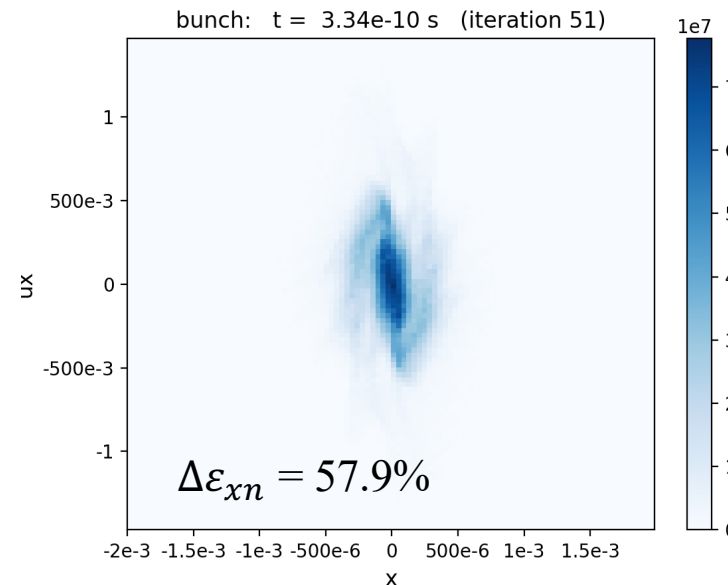
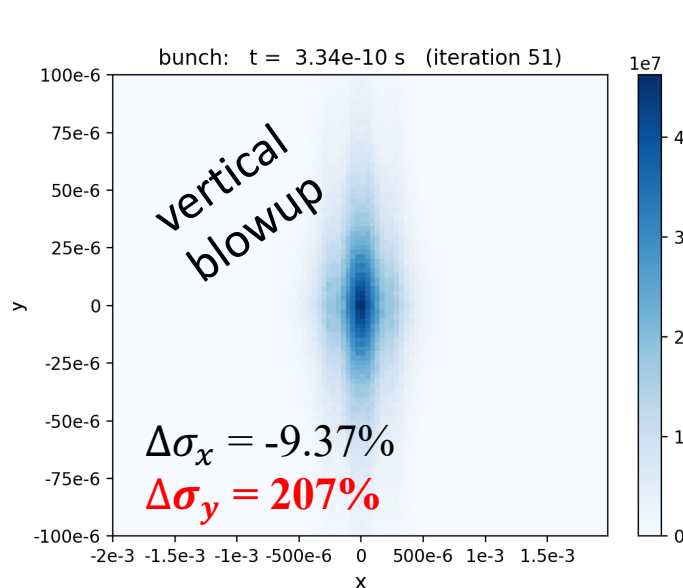


Beam evolution in a 0.1 m plasma of density $10^{17}/\text{m}^3$

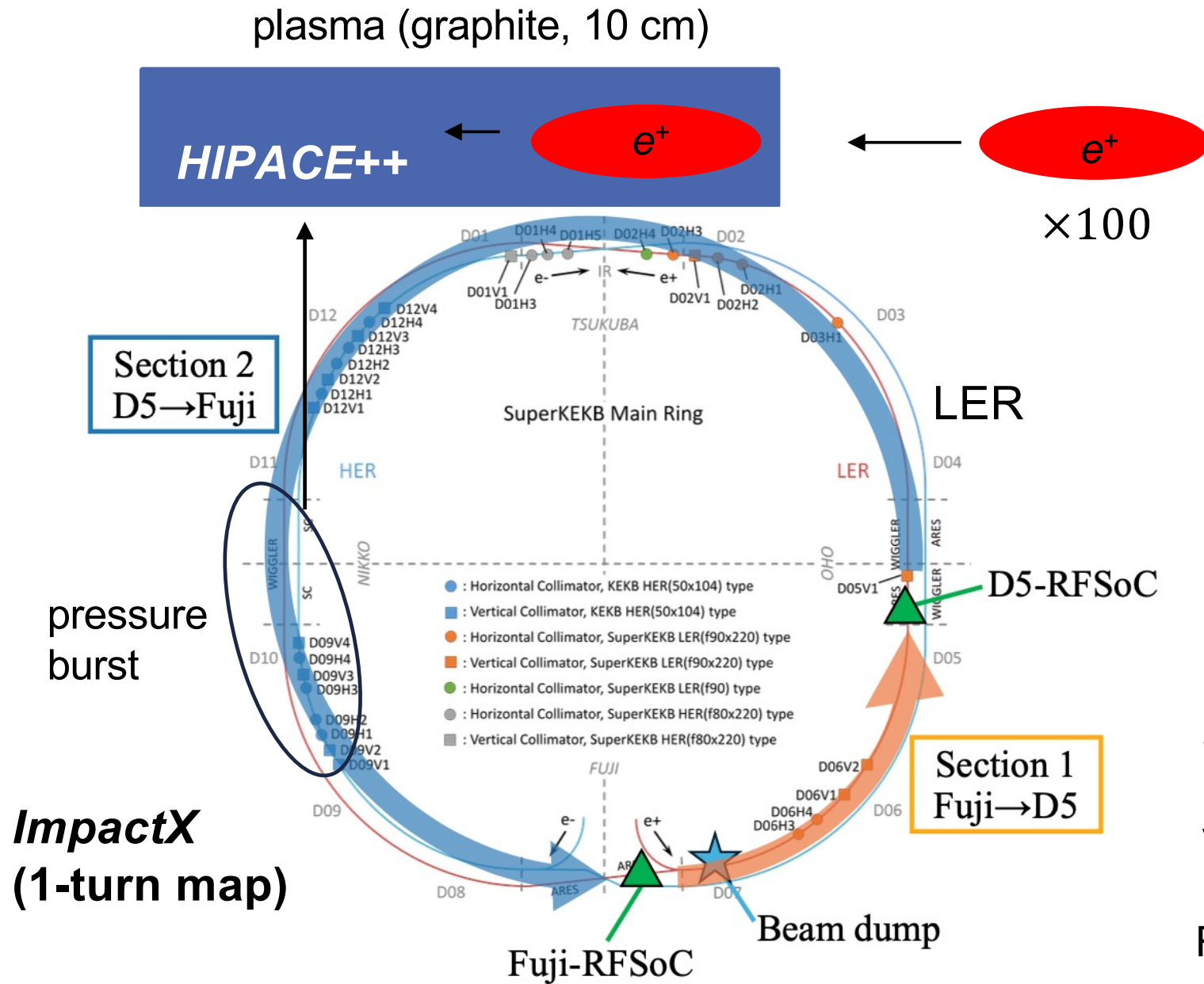
Initial bunch



After 10 passes



Workflow for integrated multi-bunch multi-turn modeling



Plasma transit time: 0.33 ns

Circulation time: 10.06 μ s

Bunch spacing: 1-4 ns

Number of bunches: 2,346
(Simulated train: 100)

Bunch population: $2-6.25 \times 10^{10}$
(Simulated particles/bunch: 10^6)

SBL event Oct. 28, 2024:
Most charge lost on
vertical collimators in Section 1

R. Nomaru *et al* (2026), Fig. 3

Workflow for integrated multi-bunch multi-turn modeling

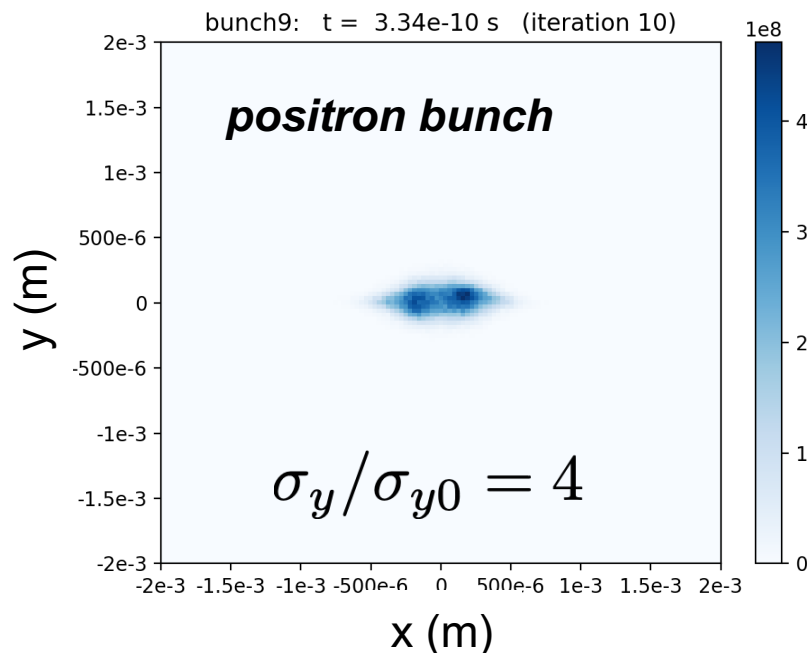
Simulation Workflow (NERSC Perlmutter, 8 x A100 GPUs)

- use HIPACE++ to model interaction of a train of 100 identical bunches with the (10 cm) plasma channel
- store the final state of the *last bunch* and the *plasma* as openPMD (hdf5) data [to approximate steady-state plasma and bunch properties associated with the train passage]
- read the bunch particle data into ImpactX and propagate for 1 turn (currently, using the 1-turn map within ImpactX)
- use the updated bunch data to populate a new train of 100 bunches together with the stored plasma in HIPACE++
- push the updated bunch train through the updated plasma using HIPACE++
- repeat the previous steps x 10 times to model 10 turns

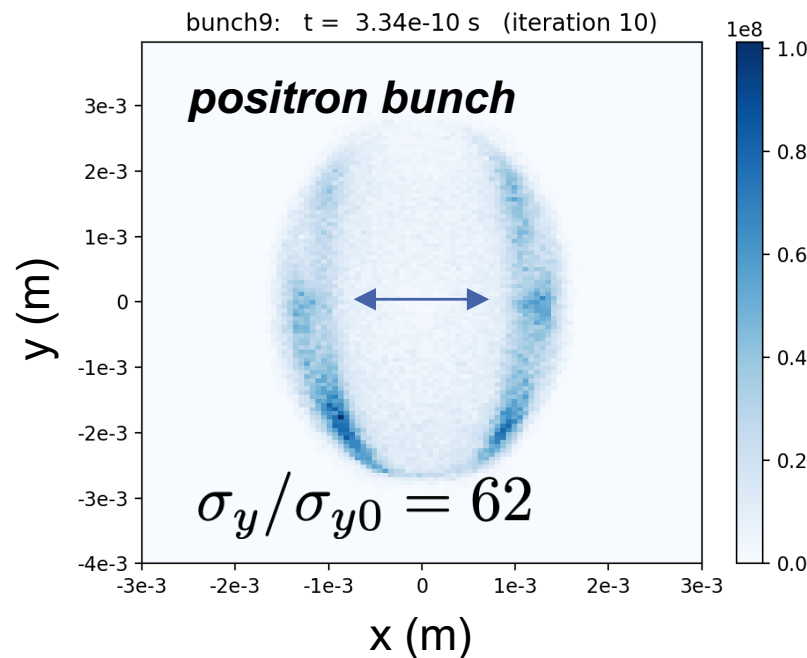
Observations (ongoing / WIP)

- Low-resolution exploratory runs reveal compute time is dominated by:
1) number of longitudinal grid cells required to resolve the train ($>10^5$), 2) openPMD data management for the bunch train I/O, 3) queue wait times (chained dependency jobs).
Note 2) + 3) could be addressed with improved code integration.
- Demonstration using a short train of 10 bunches ($\sim 10^4$ longitudinal cells):

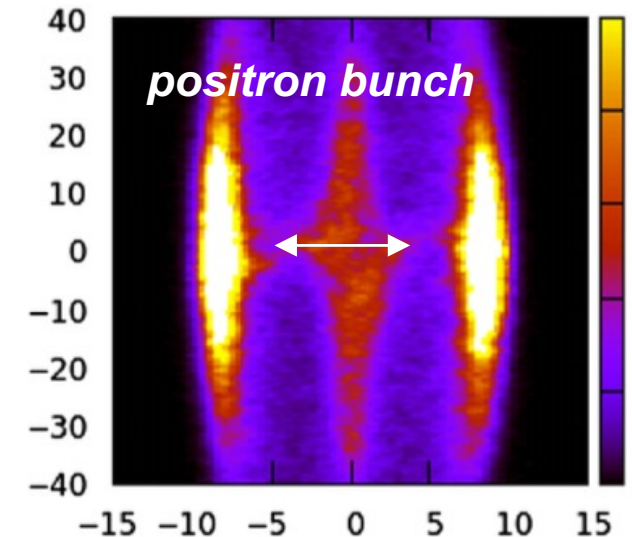
After 3 turns (10th bunch)



After 4 turns (10th bunch)



$n_e = n_i = 2 \times 10^{18} / \text{m}^3$
Qualitative comparison only



K. Ohmi *et al* (2026), Fig. 13A

Summary and Future Work

- Developed an integrated simulation workflow modeling beam blow-up triggered by a long bunch train interacting with graphite dust plasma over 10 turns in the SuperKEKB LER.
- Exploited coupling of HIPACE++ with ImpactX running on NERSC Perlmutter GPU; compute time is strongly dominated by plasma interaction.

Next steps:

- Use full element-by-element tracking in ImpactX model to support lattice transport from proposed dust location (pressure bump) to vertical collimator, and back to dust location.
- Collaborate toward benchmarking among codes and with experimental data for SuperKEKB.
- Investigate the size of neglected effects (e.g. 3D effects, plasma oscillations, nonlinear lattice transport, etc.)

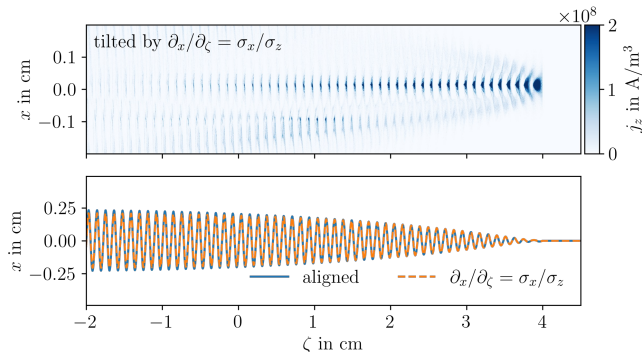
Backup material

Examples: Applications of HIPACE++

DESY

Tilted proton bunch in AWAKE

Long beam, many beam particles, long simulation box, long plasma

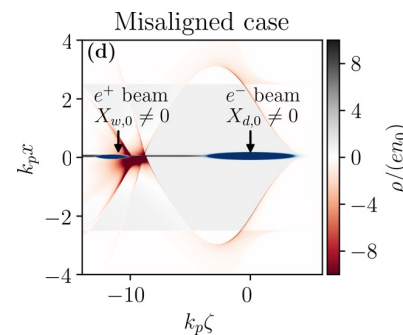


Costs: 1 node hour*

on the JUWELS Booster (16 nodes),
in FP64, **512×512×2048** grid
points,
120×10⁶ beam particles, **400** time
steps

e⁺ acceleration in a plasma column

High-resolution, sharp plasma spikes, many time steps, many beam particles



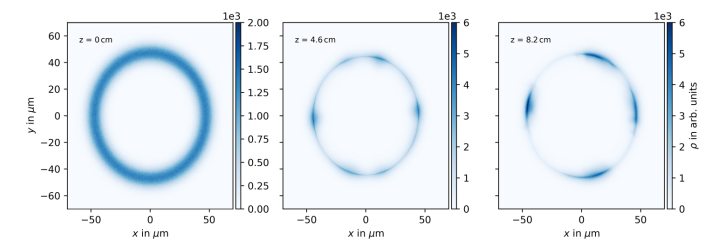
Costs: 18 node hours*

on the JUWELS Booster (32 nodes),
in FP64, **4096×4096×2048** grid
points, 225×10⁶ beam particles,
750 time steps

*Our JUWELS allocation: 26 000 node hours
"large JUWELS allocation": 156 000 node hours

e⁺ acceleration with hollow-core driver (Jain et al., PRL 2015)

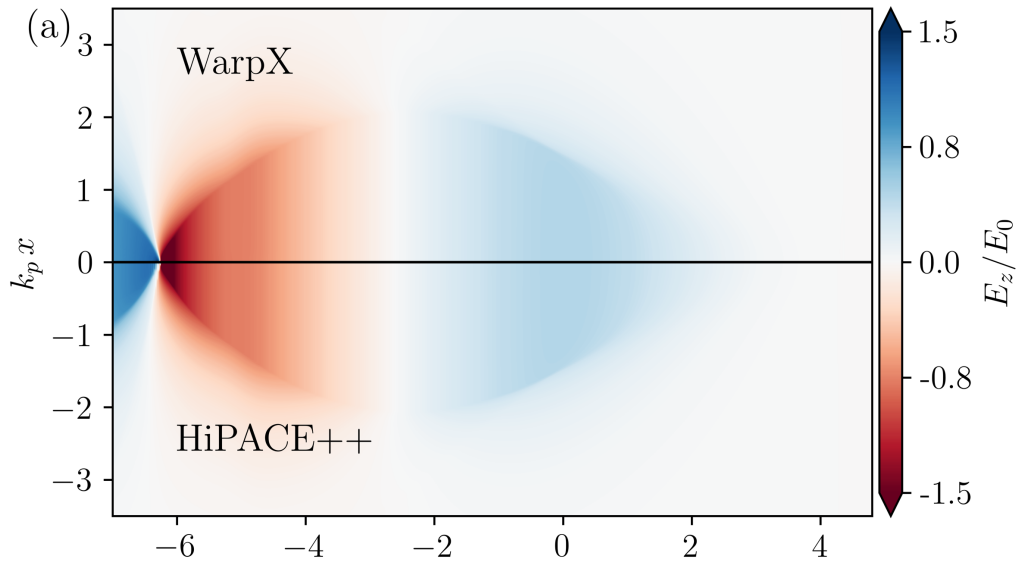
Standard 3D simulation



Costs: 1 hour on a laptop

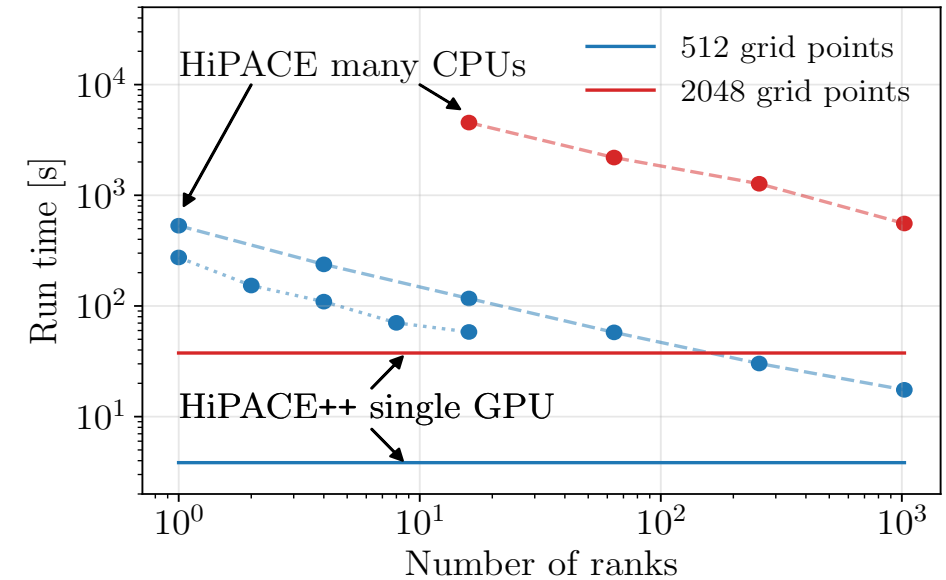
(NVIDIA RTX2070)
in FP32, **1024×1024×1024** grid
points, 10⁷ beam particles, **300**
time steps

HIPACE++ accuracy and performance benchmarks

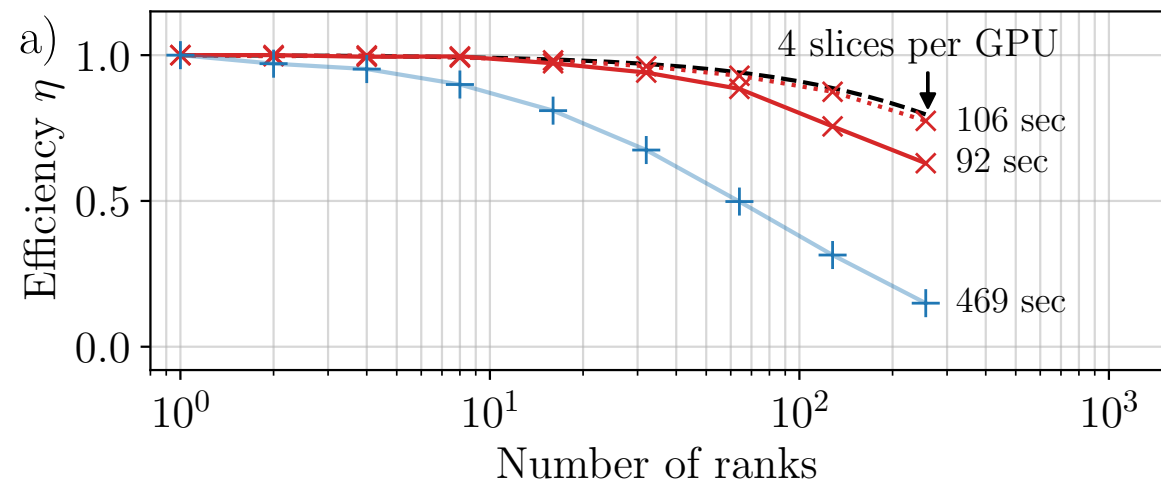


- Successfully benchmarked
- Large speedup w.r.t. CPU
- High-resolution simulations within minutes
- Production simulations possible on a laptop (float)

Slide from M. Thevenet (DESY)



➤ 1 GPU MUCH faster than many CPU cores



➤ Excellent scaling to hundreds of GPUs

Documentation & Testing is Part of Our Development



Online Documentation:
[warpX](https://warpX.readthedocs.io)|[hipace](https://hipace.readthedocs.io)|impactx.readthedocs.io

Open-Source Development & Benchmarks:
github.com/BLAST-ImpactX

USAGE

Run WarpX

Input Parameters

Python (PICMI)

Examples

- Beam-driven electron acceleration
- Laser-driven electron acceleration
- Plasma mirror
- Laser-ion acceleration
- Uniform plasma
- Capacitive discharge

For a complete list of all example input files, have a look at our [Examples/](#) directory. It contains folders and subfolders with self-describing names that you can try. All these input files are automatically tested, so they should always be up-to-date.

Beam-driven electron acceleration

AMReX [inputs](#):

- 2D case
- 2D case in boosted frame
- 3D case in boosted frame

All checks have passed

24 successful and 1 neutral checks

✓	macOS / AppleClang (pull_request)	Successful in 40m	Required	Details
✓	Windows / MSVC C++17 w/o MPI (pull_request)	Successful in 58m		Details
✓	CUDA / NVCC 11.0.2 SP (pull_request)	Successful in 31m	Required	Details
✓	HIP / HIP 3D SP (pull_request)	Successful in 29m		Details
✓	Intel / oneAPI DPC++ SP (pull_request)	Successful in 38m		Details
✓	OpenMP / Clang pywarpX (pull_request)	Successful in 37m	Required	Details

264 physics benchmarks run on every code change of WarpX
262 physics benchmarks for ImpactX (Sept. 2026)

Rapid and easy installation on any platform:



conda install
-c conda-forge warpX



spack install warpX
spack install py-warpX



cmake -S . -B build
cmake --build build --target install



python3 -m pip install .



brew tap ecp-warpX/warpX
brew install warpX



module load warpX
module load py-warpX