



A Novel Linac Simulation Code AVAS

Institute of Modern Physics, Chinese Academy of Sciences

Linac Physics Group

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Mini-Workshop on Beam Dynamics Simulation for Particle Accelerators

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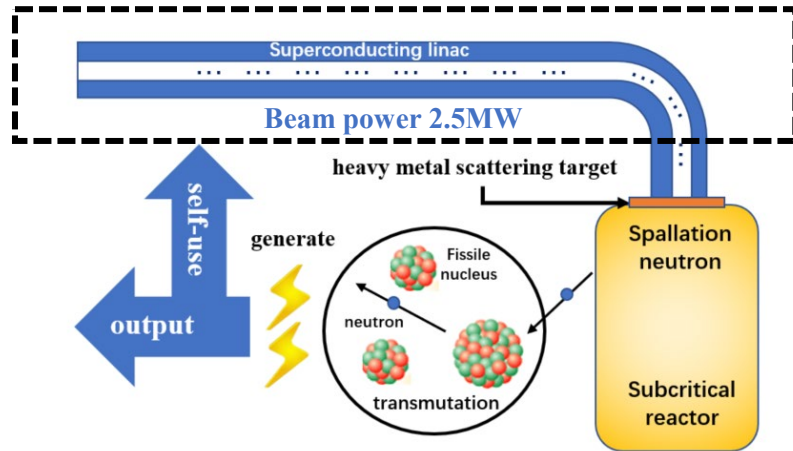
Verification and Application

6

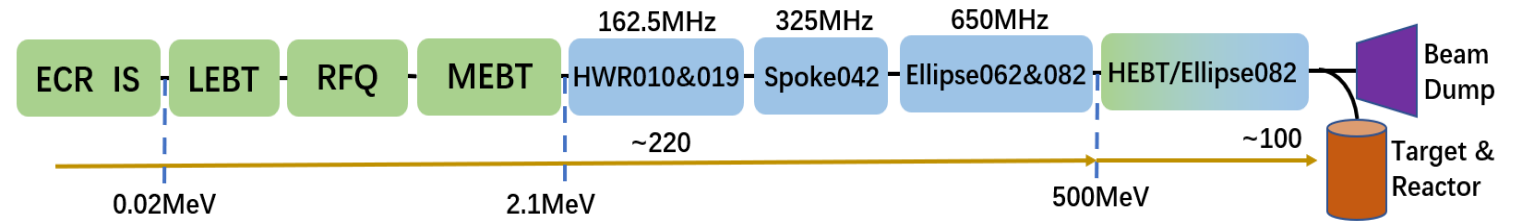
Summary

Background and Challenges

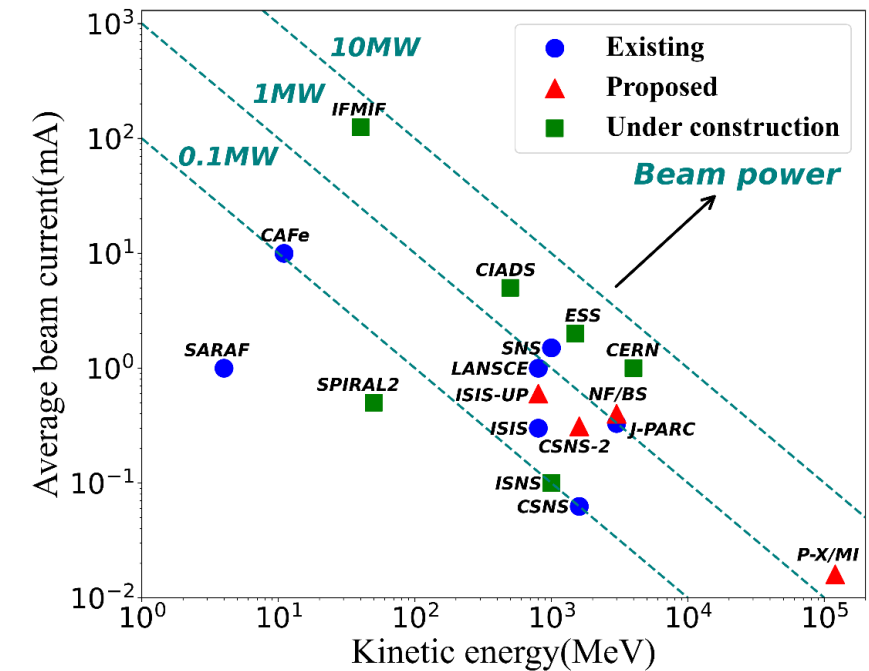
CiADS Schematic



To meet the ever-growing application demands, the energy and beam intensity of accelerators continue to increase.



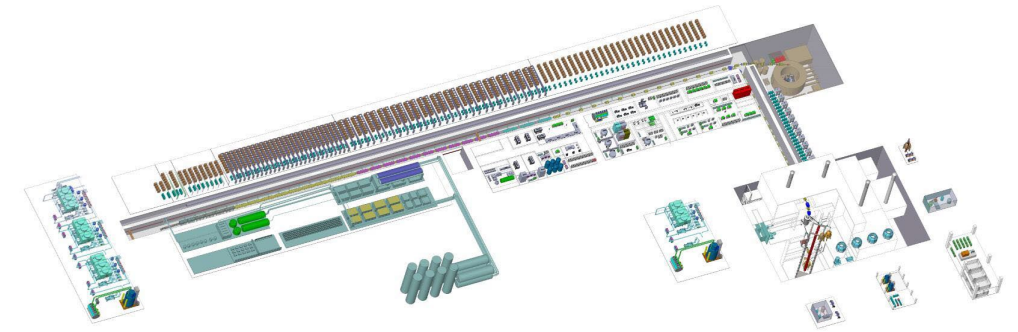
Structure layout of CiADS superconducting linac



Background and Challenges

Commonly Used Beam Simulation Programs

Program	Laboratory	Language	Core Algorithm
TraceWin	CEA Saclay	C++, Python	PIC(PICNIC)
RF-Track	CERN	C++, Octave, Python	Particle to Particle, CIC (Green function method)
PyORBIT	ORNL	C++, Python	PIC(FFT)
IMPACT	LBNL	Fortran90, Python	PIC (Green function method)
Parmila	LANL	Fortran	PIC(SCHEFF)
BEAMPATH	SLAC	Fortran77	PIC(FFT)
TRACK	ANL	-	PIC(FFT)
WARP3D	LLNL	Fortran	PIC(FFT)



Challenges for Accelerator Simulation

1.Computational Complexity: High-intensity ion accelerators exhibit strong nonlinear space charge effects, requiring minutes to hours per simulation.

2.Simulation-Reality Mismatch: Discrepancy between numerical simulations and actual accelerators, making it difficult to closely integrate numerical simulations with accelerator operations.

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Summary

Modified PICNIC algorithm: S-PICNIC

1. Fully utilize the symmetry of bunch

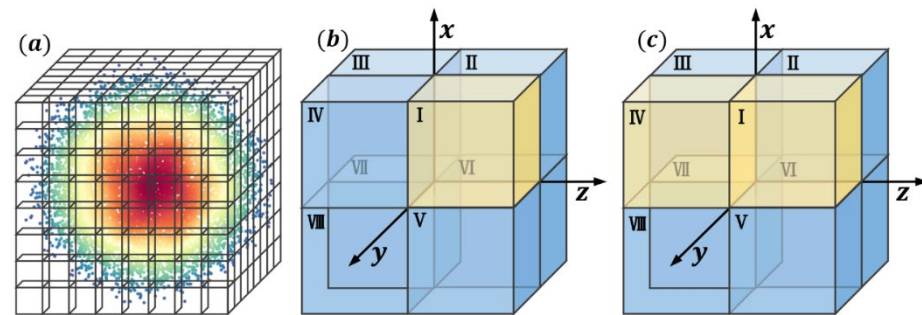
The solution of space charge field is divided into two parts:

- the solution of symmetric space charge field
- the correction of asymmetric field component

2. Neglecting operations that contribute less to the space charge field

- Setting thresholds: $TH = RTH \times \text{Central node charge}$
- Ignoring the contribution of charges below the threshold to the space charge field

The higher the bunch symmetry, the faster the solving speed.



Evaluation Metrics

- To quantify the precision loss from ignoring low-charge grid points

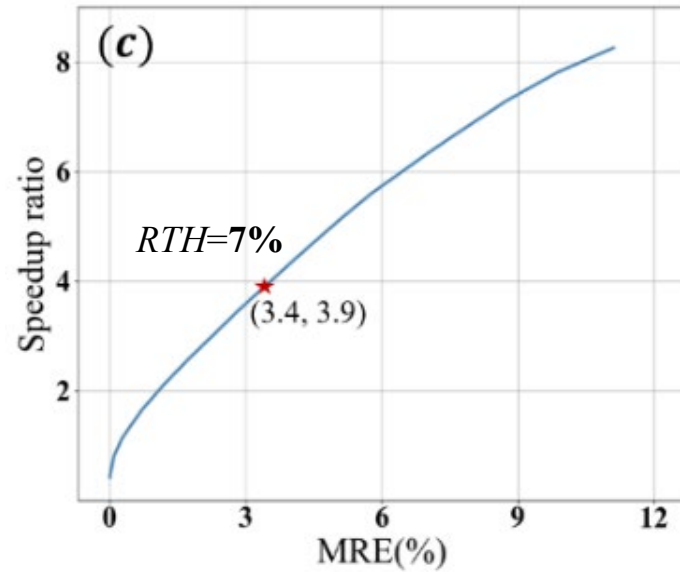
MRE = $(MRE_p + MRE_v)/2$ Quantify the difference between two bunch

$$MAE_p = \frac{1}{3} \left(\sqrt{\frac{\frac{1}{N_p} \sum_{i=1}^{N_p} (x_i - x_{0i})^2}{xSize}} + \sqrt{\frac{\frac{1}{N_p} \sum_{i=1}^{N_p} (y_i - y_{0i})^2}{ySize}} + \sqrt{\frac{\frac{1}{N_p} \sum_{i=1}^{N_p} (z_i - z_{0i})^2}{zSize}} \right)$$

$$MAE_v = \frac{1}{3} \left(\sqrt{\frac{\frac{1}{N_p} \sum_{i=1}^{N_p} (v_{xi} - v_{x0i})^2}{v_xSize}} + \sqrt{\frac{\frac{1}{N_p} \sum_{i=1}^{N_p} (v_{yi} - v_{y0i})^2}{v_ySize}} + \sqrt{\frac{\frac{1}{N_p} \sum_{i=1}^{N_p} (v_{zi} - v_{z0i})^2}{v_zSize}} \right)$$

Test result

- Change RTH from 0.05% to 20%, observation of the relation between speed and error



$$\text{Speed up ratio} = \frac{T_1}{T_2}$$

T_1 and T_2 are the time taken by the standard algorithm and the modified algorithm to solve the SC field, respectively

simulation conditions

Macro-particle number: 99940

Intensity: 5mA

Frequency: 162.5MHz

Initial energy: 1.52MeV

length: 13.07m

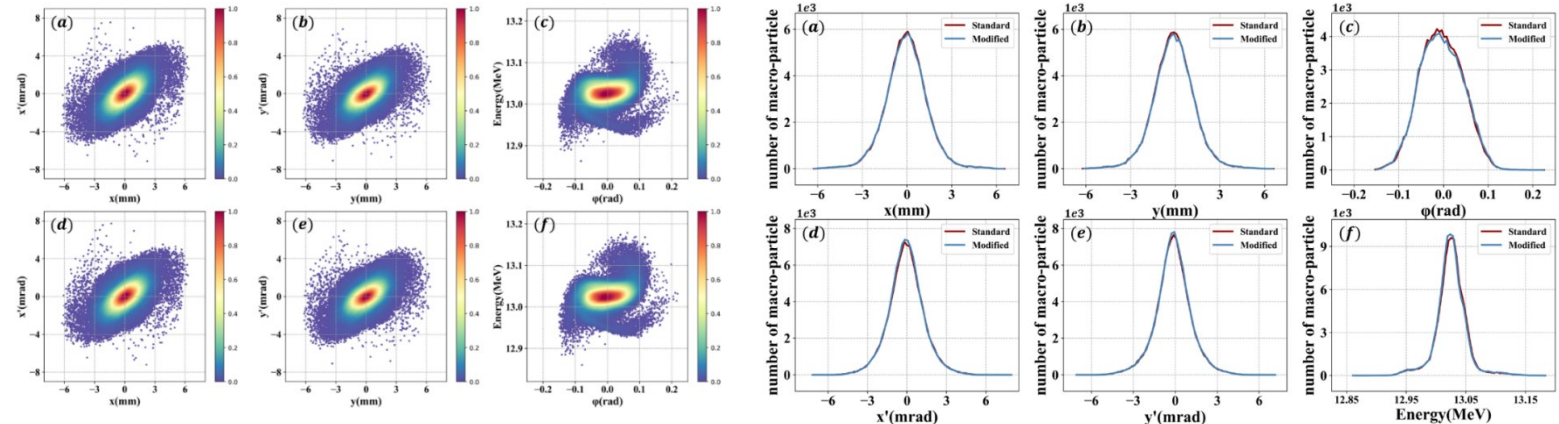
Grid: $22 \times 22 \times 22$

$\Delta t = 1/(\text{frequency} * 100)$

$RMS_x = 1.48mm$

$RMS_y = 0.66mm$

	$Emit_x$	$Emit_y$	$Emit_z$	Energy
PICNIC	0.2281	0.2153	0.0336	13.0278
S-PICNIC	0.2315	0.2179	0.0337	13.0271



Comparison of simulation results between two algorithms when $MRE = 3.4\%$

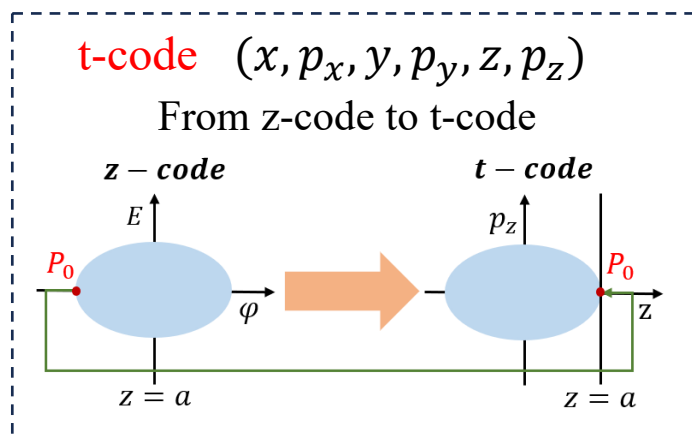
T-code and Z-code

- T-code with time as independent variable ($t \rightarrow t + \Delta t$), and the program saves the distribution of bunch at time t
- Z-code with position as independent variable ($z \rightarrow z + \Delta z$), and the program saves the distribution of bunch on the $z=a$ plane.

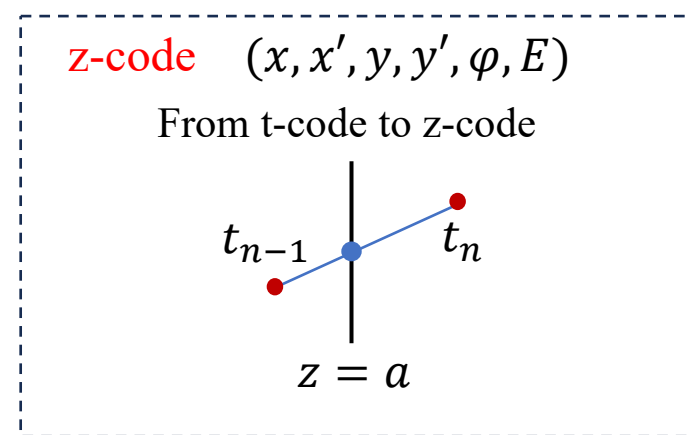
Using conditions

- Z-code is used in hard-edged components to use the transfer matrix.
- T-code is used in field map components for more accurate space charge effects calculation.

1. High-precision conversion between z-code and t-code



Key: Ensure that the parameters of the particle passing through the $z=a$ plane are consistent with those before the Conversion.

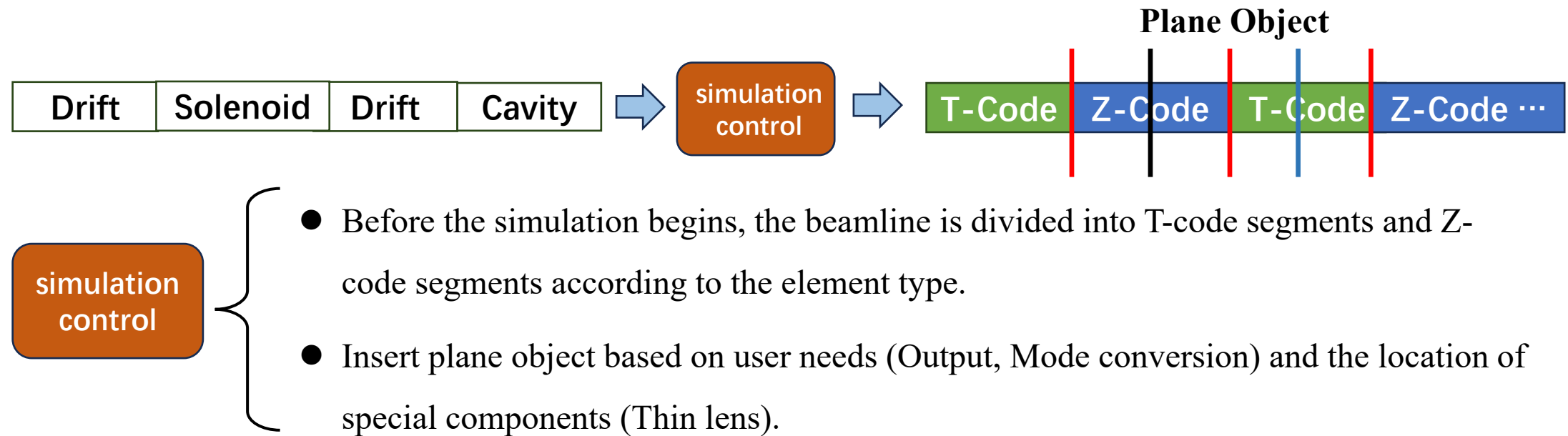


Key: by backing up the bunch information from the previous moment, interpolate to obtain the particle parameters at the $z = a$ plane

2.Code design that adapts to switching between two modes

Define plane object

- Functions such as bunch output, mode conversion, and thin lens elements are integrated into plane object.
- The bunch will stop at a plane when it encounters a plane object, and will continue to transfer from the plane after executing the operation



Unify the implementation of functions and elements across different modes.

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Summary

AVASX: An ultra-high-speed GPU-based version using CUDA.

Computational bottleneck

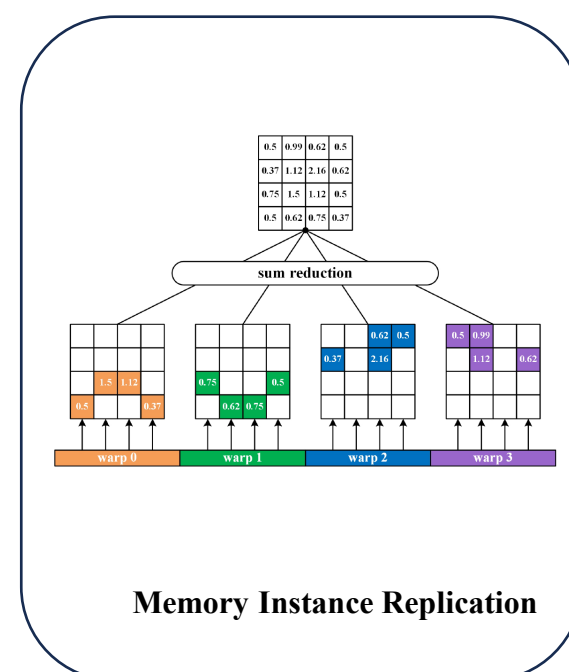
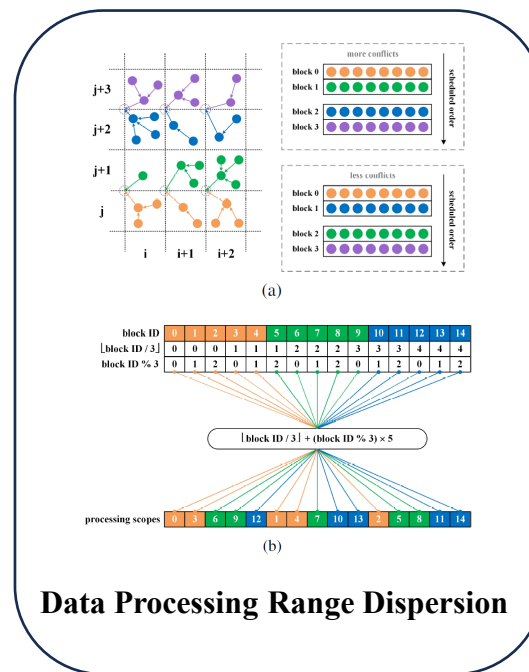
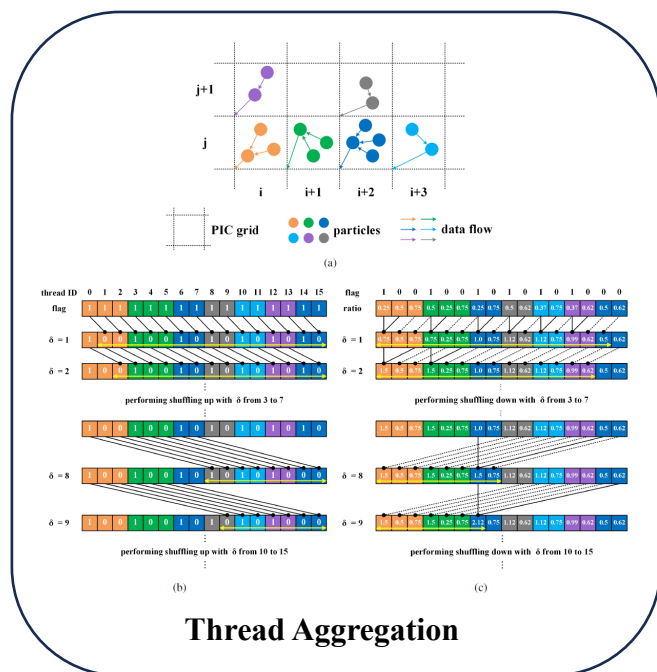
Particle-in-cell (PIC) methods require **numerous particles** to accurately capture the underlying physics.



Solution

GPU-Accelerated High-Performance Beam Dynamics Simulation

Key Challenges: Memory Access Conflicts in Atomic Operations on GPU Global Memory





Performance Evaluation

Memory workload metrics, compute and memory throughputs, execution durations, and performance improvements of different kernels in charge deposition calculation.

Employed kernel	Threads to L1 cache			L1 cache to L2 cache		Excessive sector rate (%)	Compute throughput (%)	Memory throughput (GB/s)	Duration (us)	Performance improvement (%)
	Memory requests	Requested sectors	Hit rate (%)	Memory requests	Requested sectors					
kernel-3	250,000	7,832,161	0%	7,691,097	7,832,161	9.9	6.92	51.96	463.01	-
kernel-4	250,016	7,814,976	0%	7,673,911	7,814,976	9.9	14.31	51.96	463.74	-0.16
kernel-5	251,904	7,822,528	0%	7,681,463	7,822,528	9.9	14.37	52.07	463.49	-0.1
kernel-6	250,000	7,832,737	0%	7,692,102	7,832,737	9.9	7.55	56.82	426.5	8.56
kernel-7	250,000	7,869,016	0%	7,842,952	7,869,016	not provided	5.87	19.48	1230	-62.36
kernel-8	250,016	997,324	0%	992,204	997,324	0.01	29.95	167.78	143.55	222.54
kernel-9	251,904	1,004,876	0%	999,756	1,004,876	0.01	56.05	307.71	78.4	490.57
kernel-10	250,016	997,324	0%	992,204	997,324	0.01	45.79	250.32	96.67	378.96
kernel-11	251,904	1,004,876	0%	999,756	1,004,876	0.01	64.63	356.47	68.06	580.3

580% Improvement in Charge Deposition Performance

A100 GPU vs. 56 Core CPU

CAFe

Length: 16.27m $f = 162.5\text{MHz}$

Intensity : 0.27mA $\Delta t = \frac{1}{100*f}$

CiADS

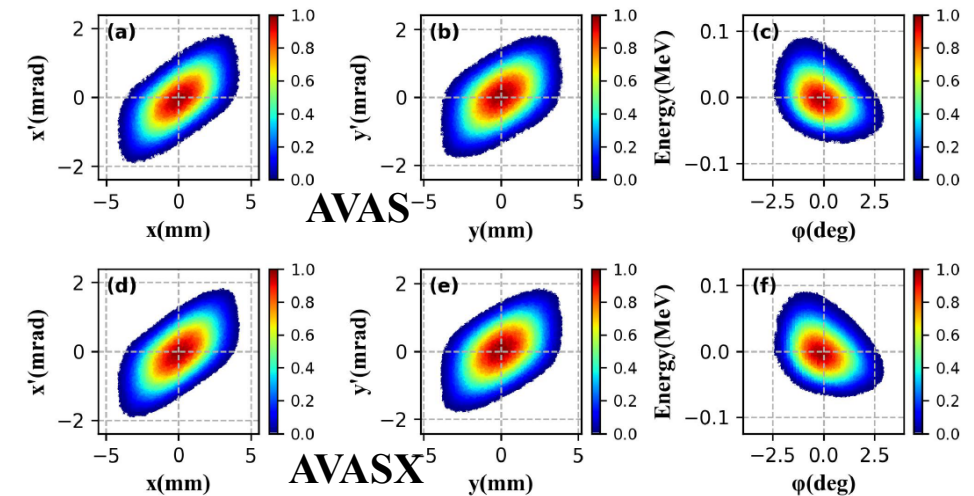
Length: 202.75m $f = 162.5\text{MHz}$

Intensity : 5mA $\Delta t = \frac{1}{100*f}$

$N_p = 1,000,000$	CAFe	CiADS
AVASX	16.58 s	45.98 s
AVAS	2.17 h	7.04 h
Speedup	472	550

Verification

Phase space distributions of the bunch at the exit of the CAFe superconducting section



Bunch parameters at the exit of the CAFe superconducting section, simulated by AVASX and AVAS.

Code	Particle number	Emittance ($\pi \cdot \text{mm} \cdot \text{mrad}$)			Bunch size (mm)			Energy (MeV)
		$emit_x$	$emit_y$	$emit_z$	$size_x$	$size_y$	$size_z$	
AVASX	1,000,000	0.148	0.152	0.134	1.585	1.486	0.914	16.95
	10,000,000	0.149	0.152	0.134	1.585	1.486	0.914	16.95
AVAS	1,000,000	0.148	0.152	0.134	1.584	1.485	0.914	16.95
	10,000,000	0.149	0.152	0.134	1.584	1.486	0.914	16.95

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Physics-Guided Deep Learning Architecture for Beam Phase Space Super-Resolution

Key Challenge

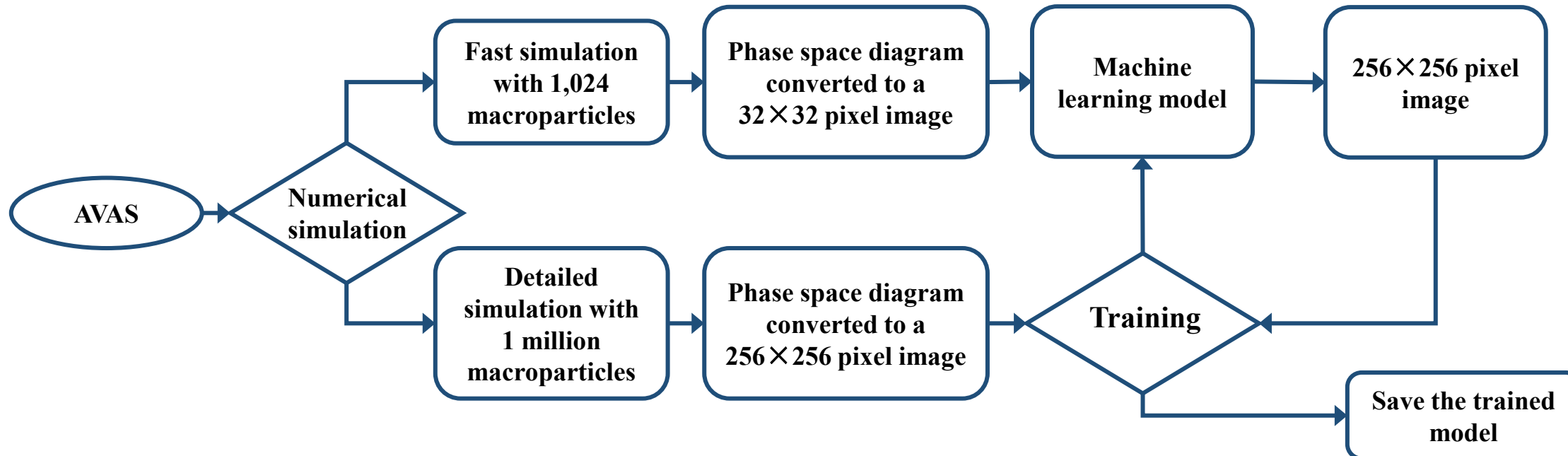
Incorporating Physical Constraints into Machine Learning to Enhance Model Generalization and Interpretability



Method

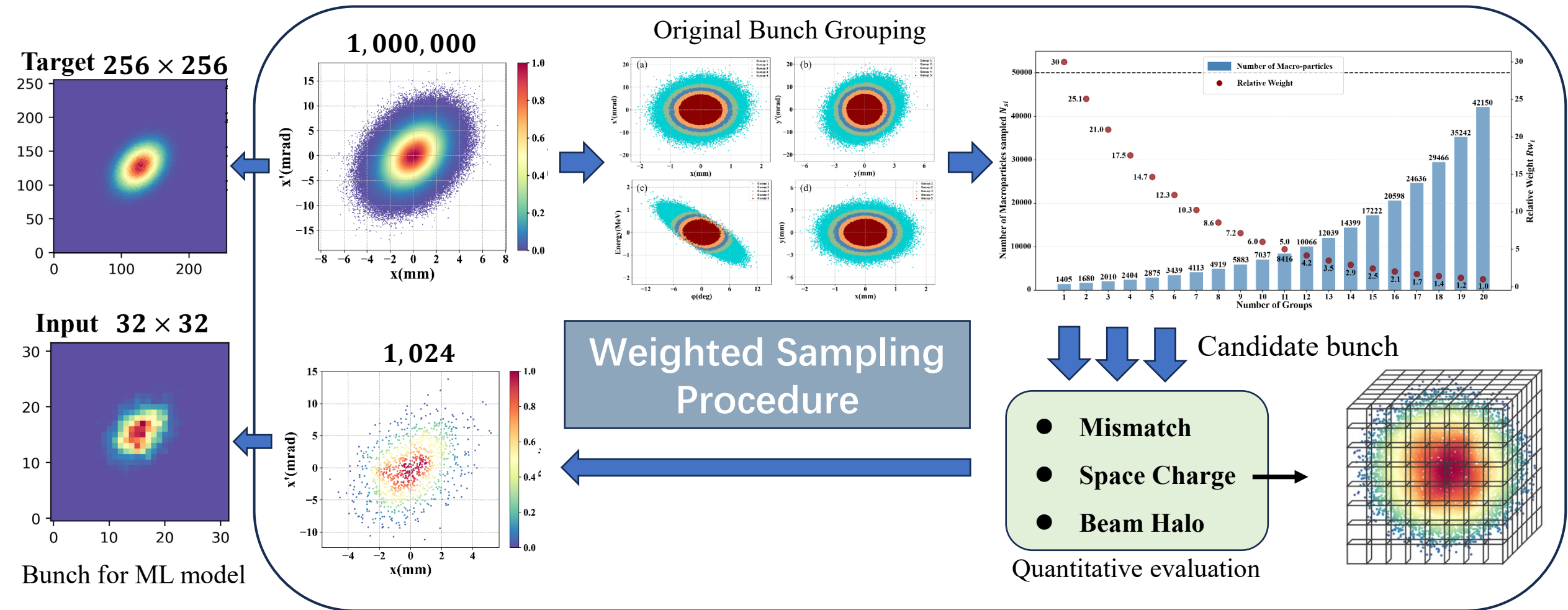
Coarse-grained simulation results obtained through rapid numerical simulations are employed to guide machine learning networks in generating refined results consistent with large-scale multi-particle simulations.

Numerical Simulation Guidance + Machine Learning Model



Physics-Guided Deep Learning Architecture for Beam Phase Space Super-Resolution

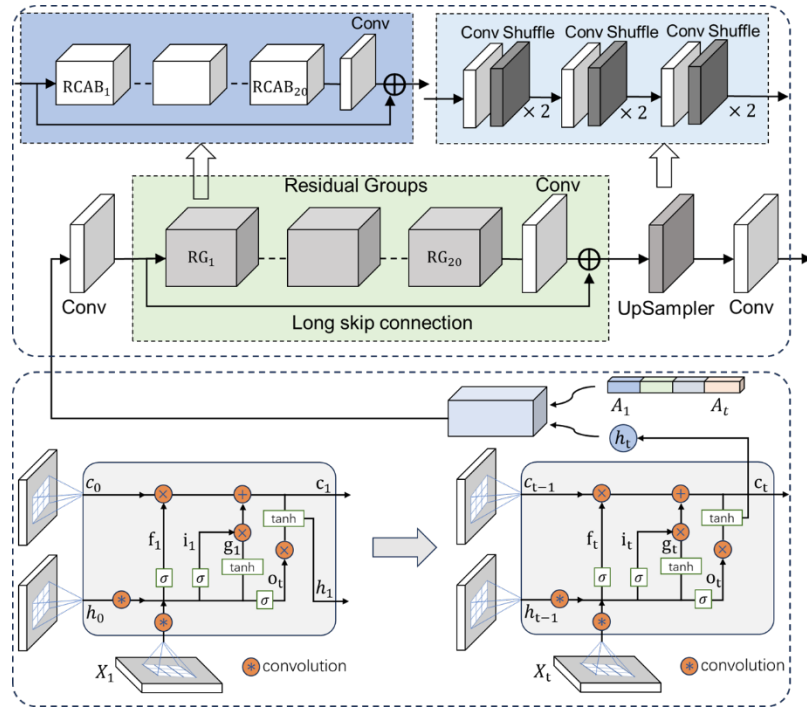
● Numerical Simulation Guidance



Machine Learning Techniques

Physics-Guided Deep Learning Architecture for Beam Phase Space Super-Resolution

Machine Learning Model



Sample size

Training set	Independent test set
800×200	200×200

800 simulation runs, each outputting 200 bunch distributions along the beamline

Data set

CAFe Superconducting Section

Macro-particle number: 1,000,000
1024

Intensity: 5mA
Frequency: 162.5MHz
Length: 13 m

Error boundaries of elements

Element Type	Cavity	Solenoid
Rotation Error	$\mp 4mrad$	$\mp 2mrad$
Translation Error	$\mp 1.0mm$	$\mp 0.5mm$
Field amplitude Error	$\mp 1\%$	$\mp 4\%$
Phase Error	$\mp 1deg$	

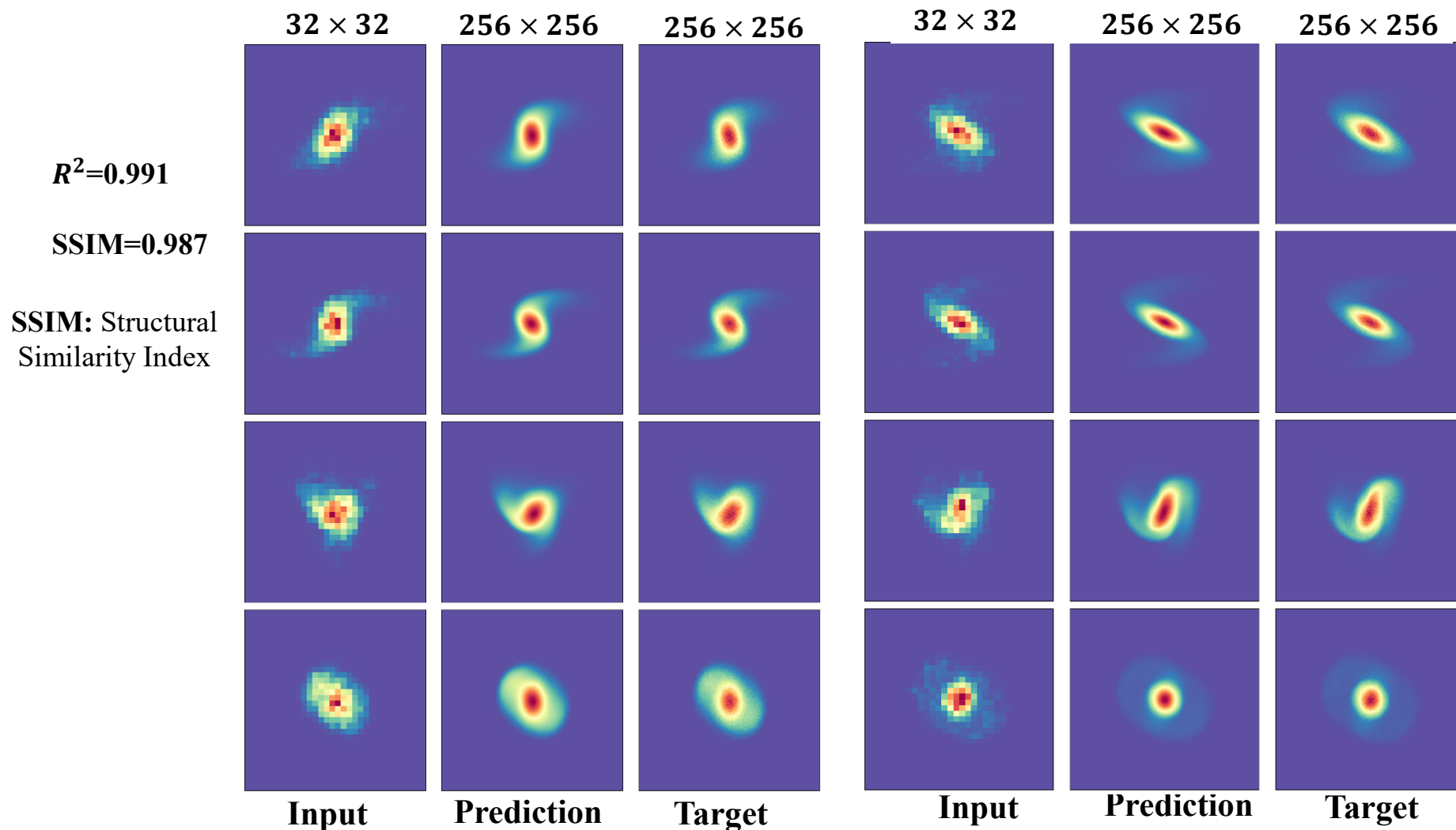
Initial bunch error

Current Error	Beam mismatch	Emittance growth
$\mp 2.5mA$	50%	50%

Physics-Guided Deep Learning Architecture for Beam Phase Space Super-Resolution

Numerical results

Generalization (Doubled error)



Summary

1.The number of particles was reduced from 1,000,000 to 1,024.

2.Next, real beam diagnostic data will be incorporated to reconstruct actual beam information.

$R^2 = 0.981$

SSIM=0.984

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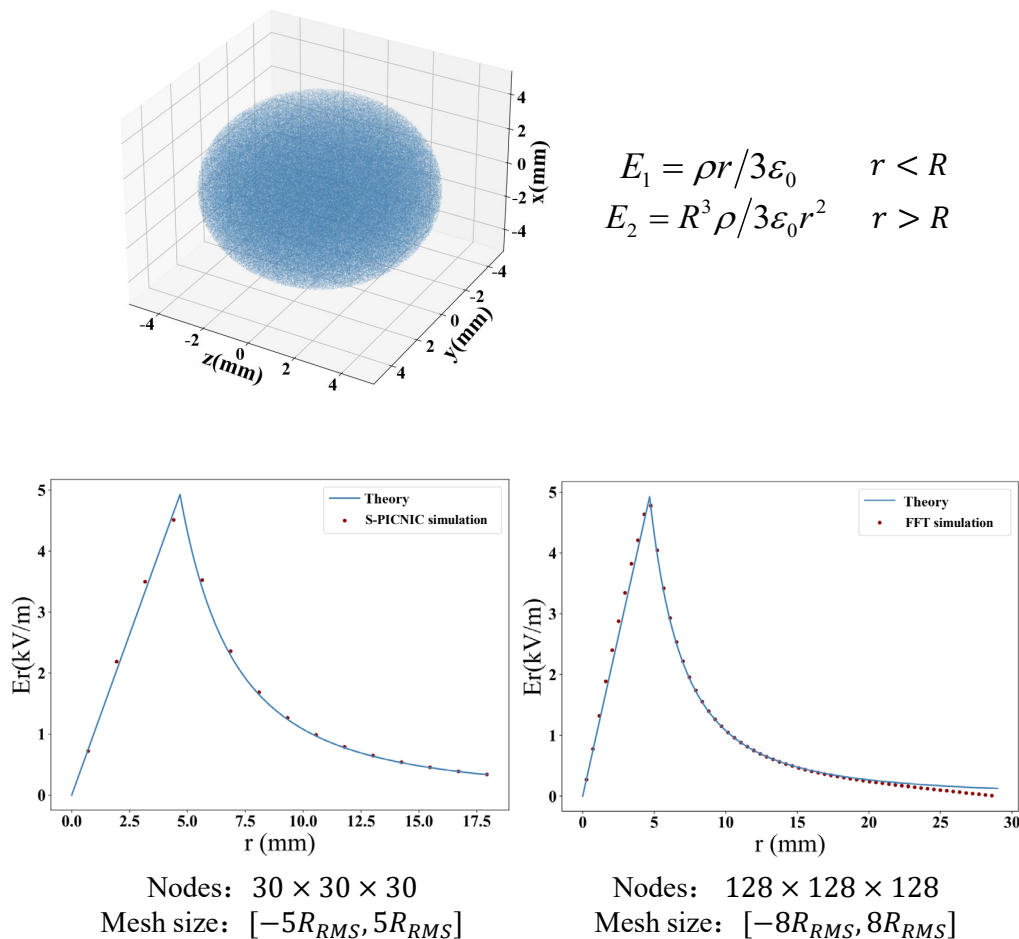
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Summary

Analytical Solution Test

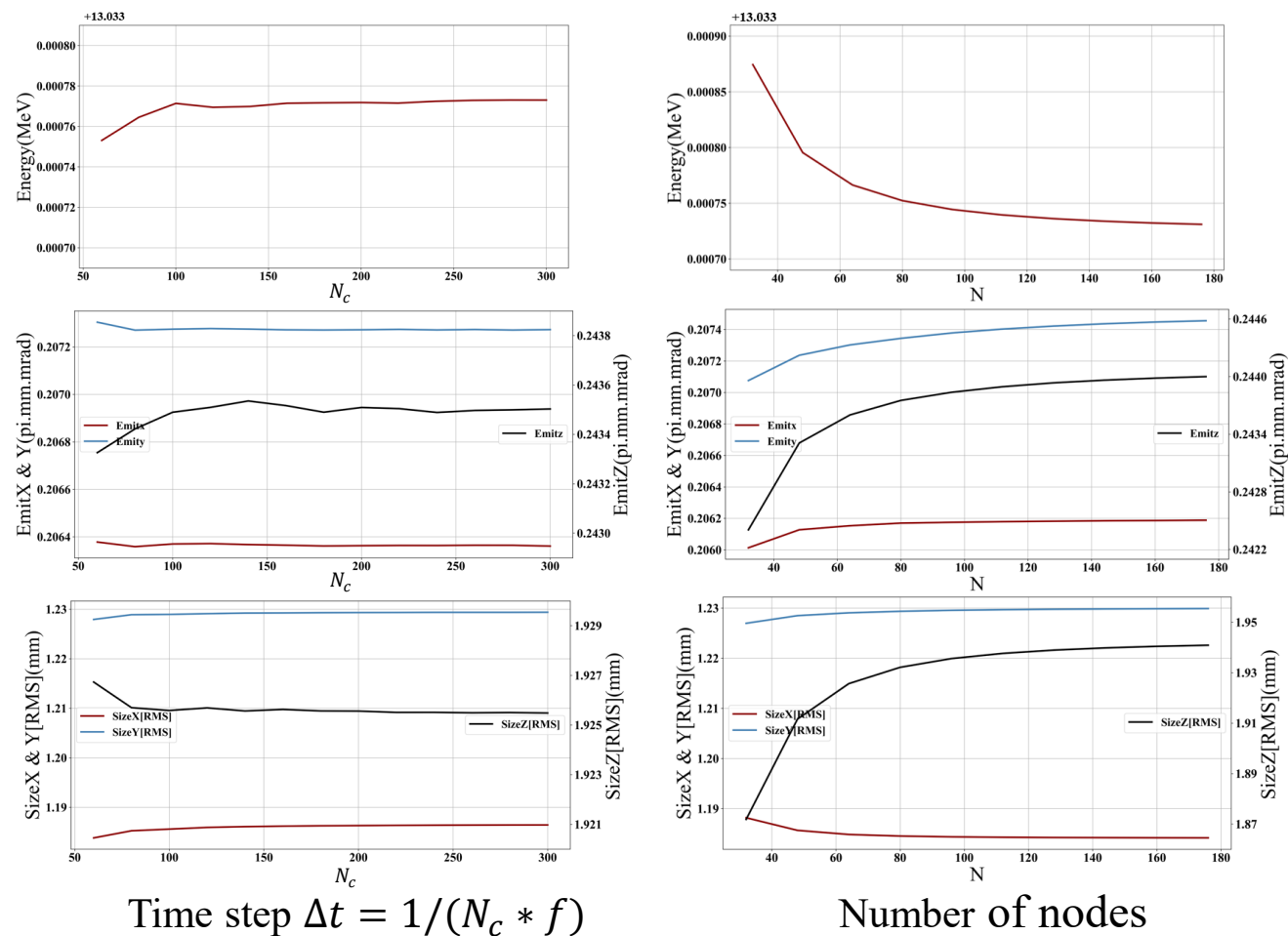
Radial Field Distribution of a Uniform Spherical Bunch



Convergence test

Beam Transport Simulation in the CAFE Superconducting Section

$I = 5\text{mA}$, $\text{Freq} = 162.5\text{MHz}$, $E = 1.5\text{MeV}$, Total length 13.07m



Benchmark

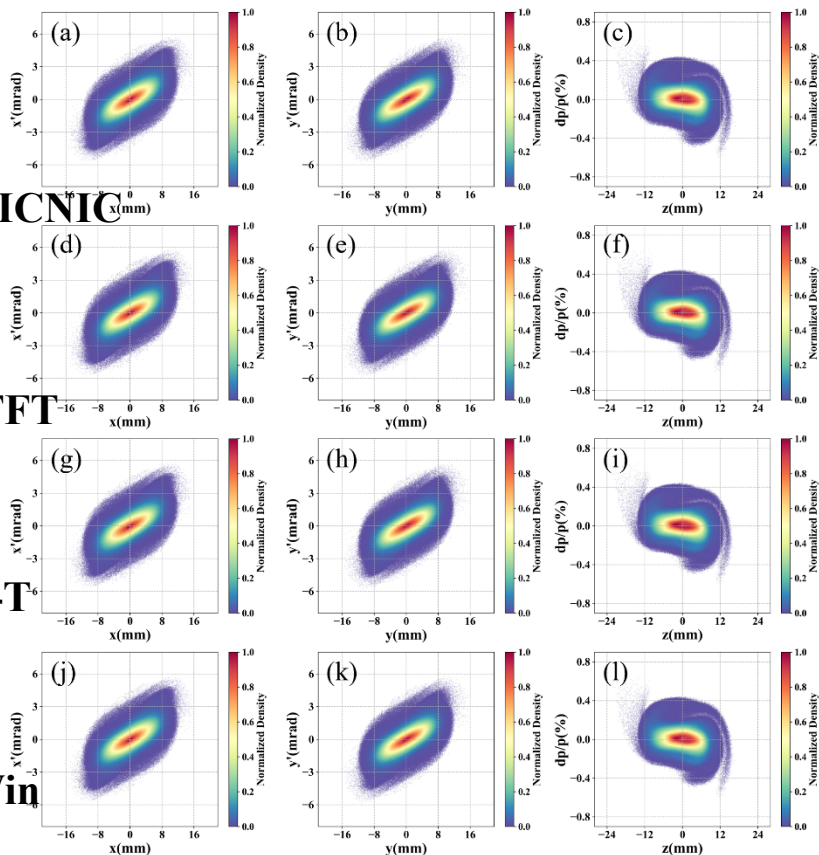
Beam Transport Simulation Through the 100 mA Beamline

AVAS-SPICNIC

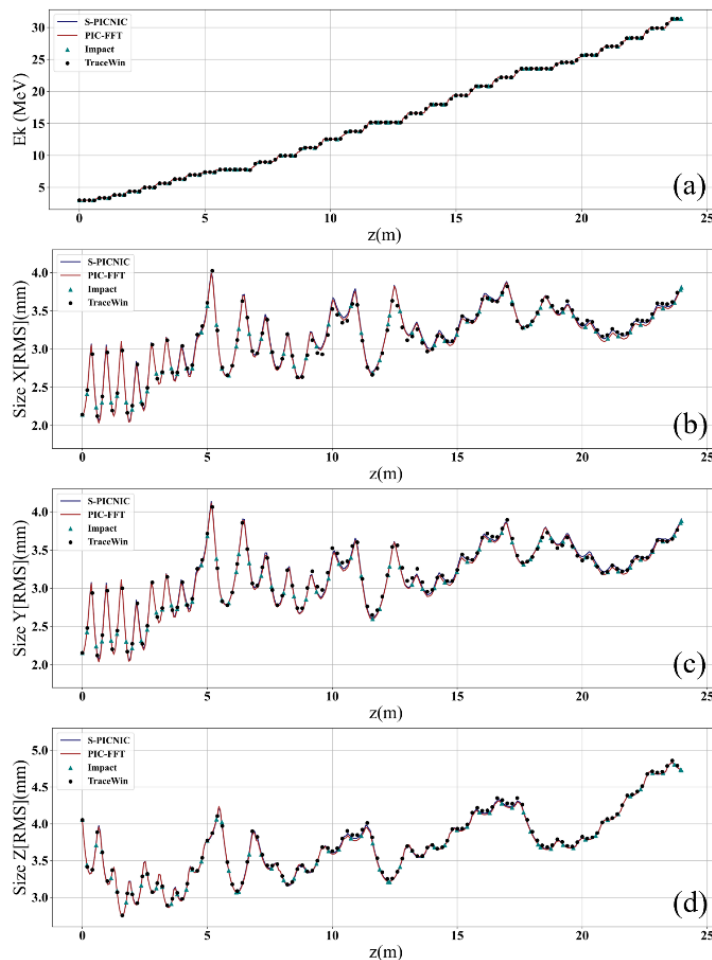
AVAS-FFT

Impact-T

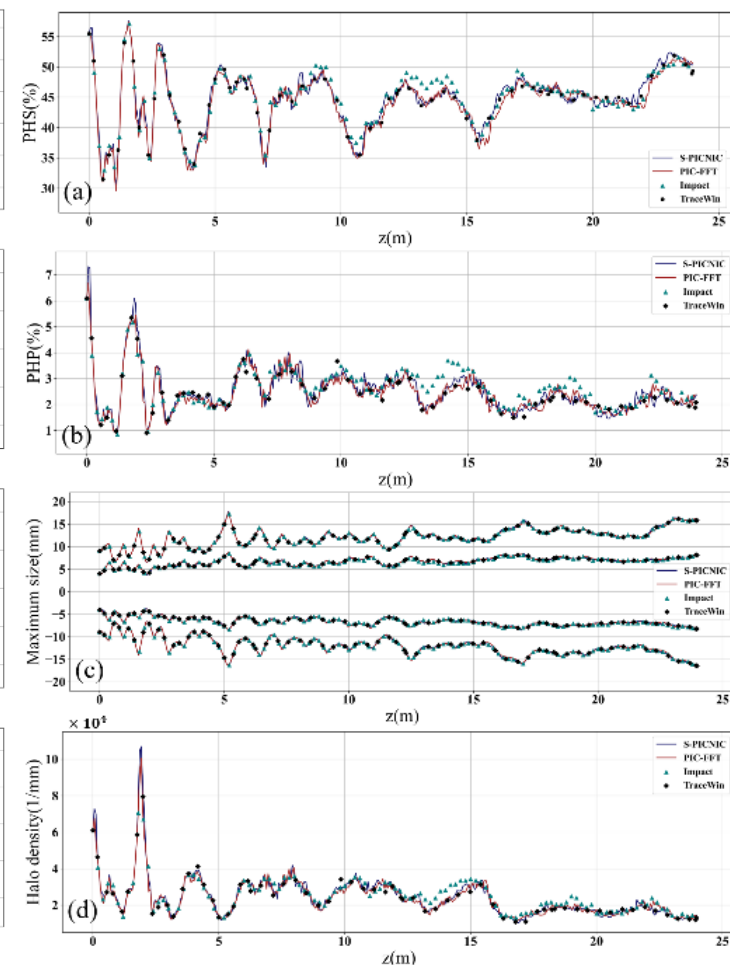
TraceWin



Phase space of exit bunch



Envelope Parameter Deviation < 1.5%

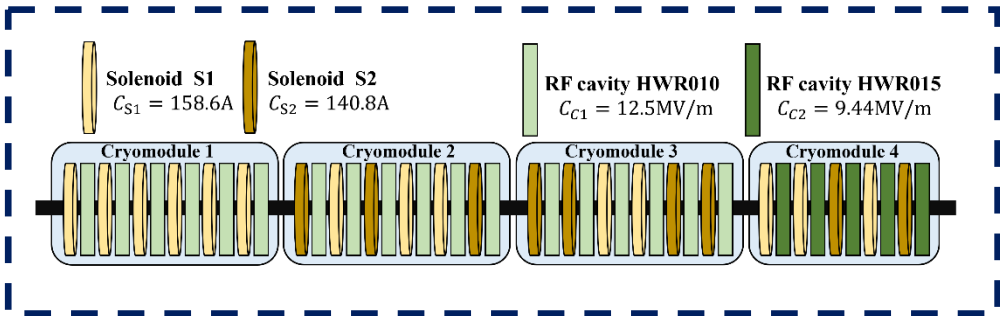


Beam Halo Parameter Deviation < 3.5%

Verification

Experimental Verification

Numerical Simulation

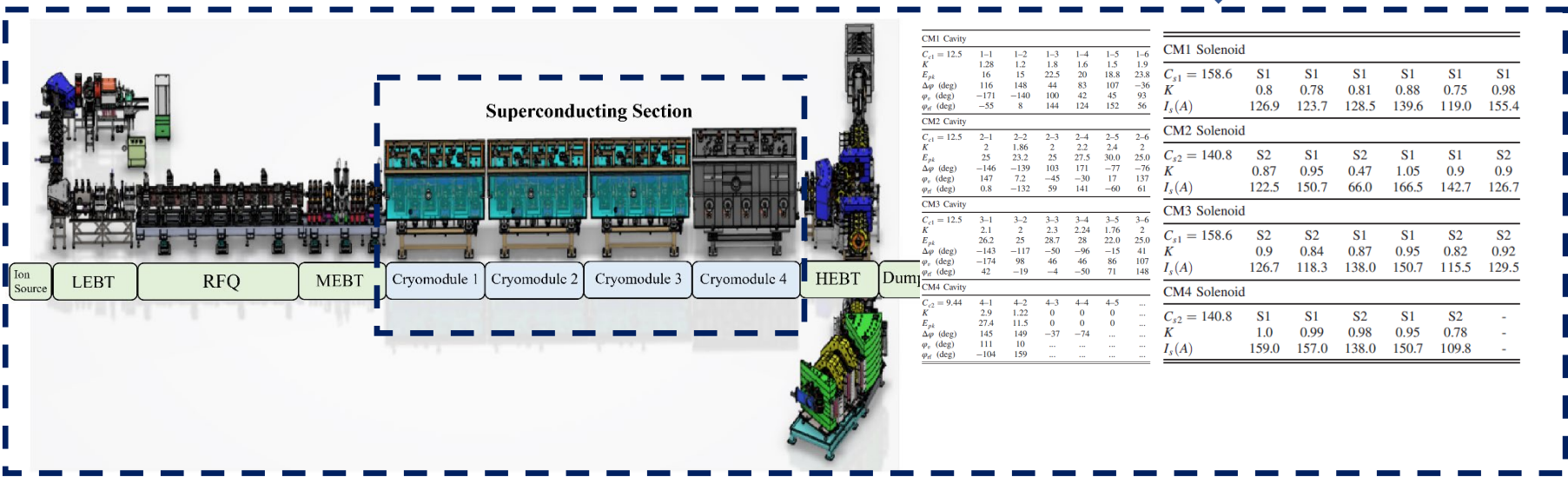


Parameter Calibration

$$\begin{cases} I_{s,i} = C_{s,i}K_i \\ E_{pk,i} = C_{c,i}K_i \end{cases}$$

$$\varphi_{rf} = \varphi_v + \Delta\varphi + n \times 360.$$

Preset Accelerator Operating Parameters



$^{40}\text{Ar}^{12+}$

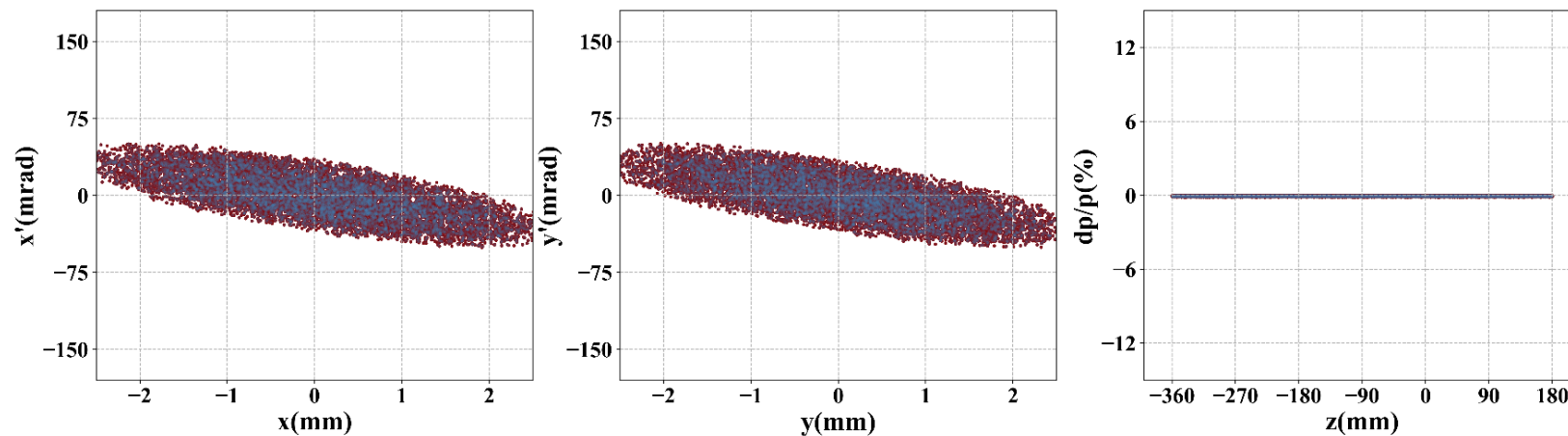
Simulated Energy:
186.6 MeV

Measured Energy :
187.5 MeV

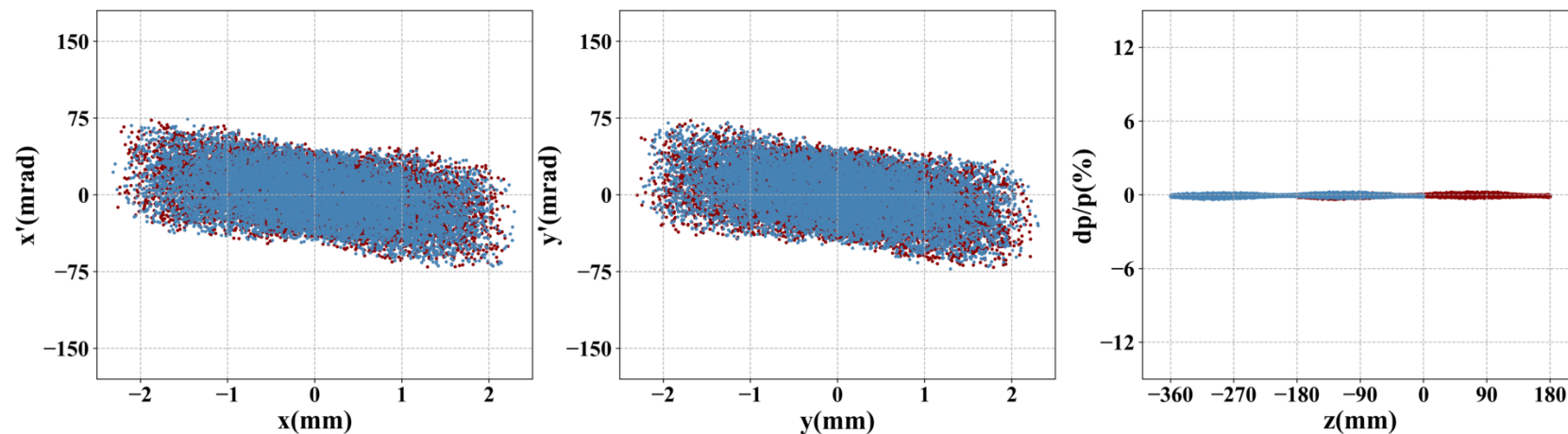
Energy Deviation <
0.5%, with No
Observable Beam Loss

Multi-bunch transmission

Initial Bunch



**Simulation of DC beam
transport consisting of mixed
protons and negative hydrogen
ions in an RFQ**



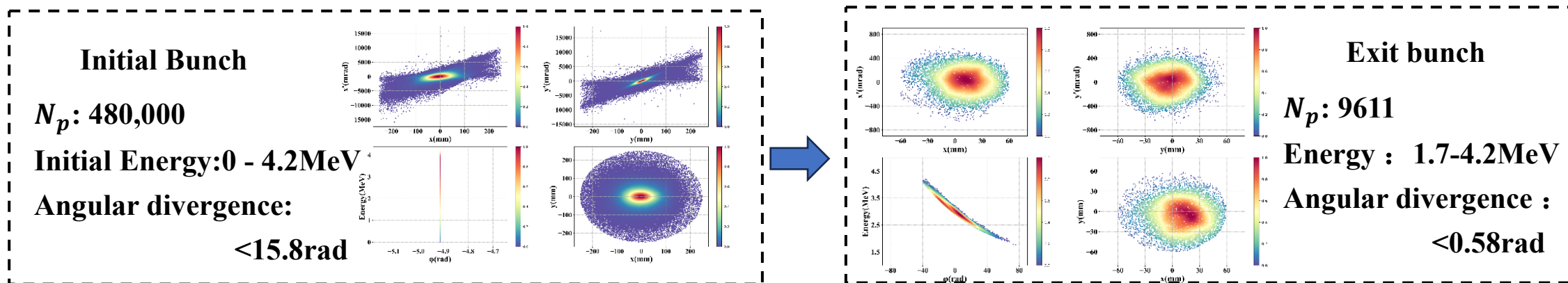
Display H^- particles within $[-360^\circ, 0^\circ]$

Display protons within $[-180^\circ, 180^\circ]$

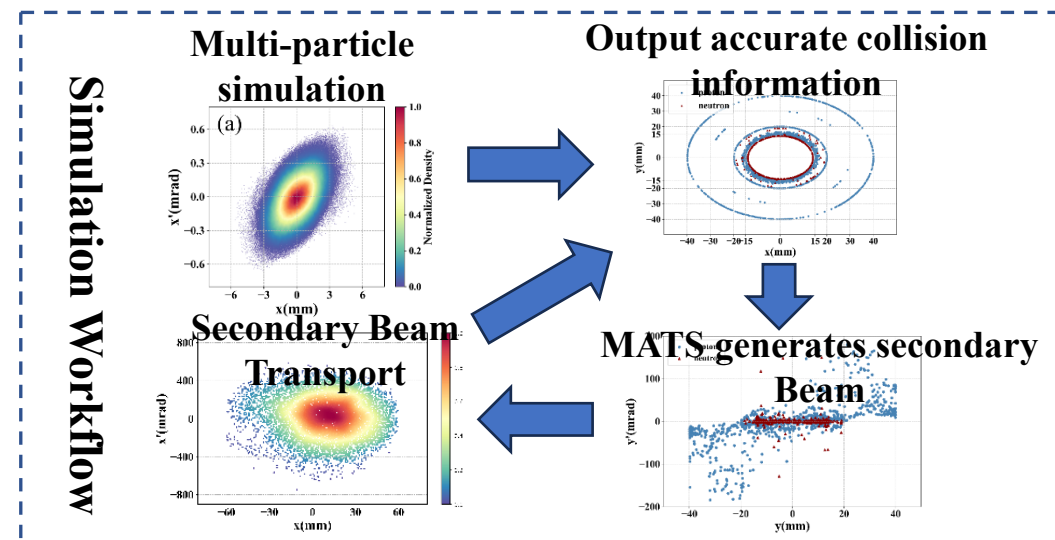
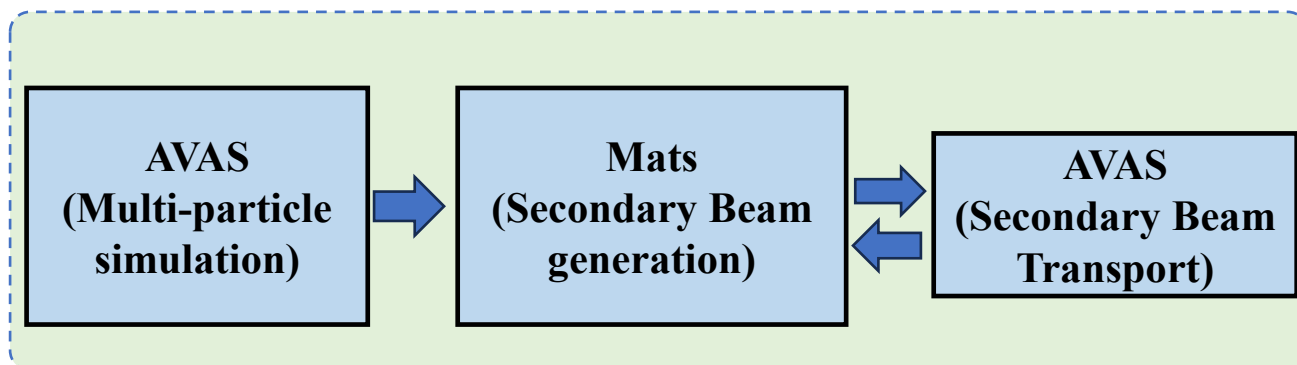
Display only particles within one period

Muon Beam Transport

Simulate beam transport with large energy spread and angular divergence

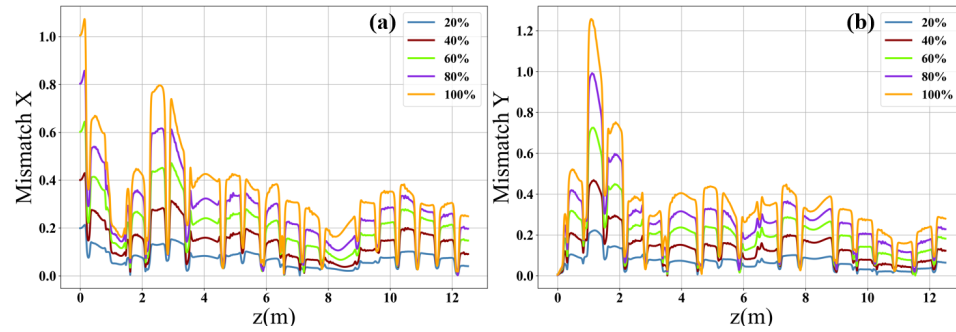


Multi-physics Simulation



Error study

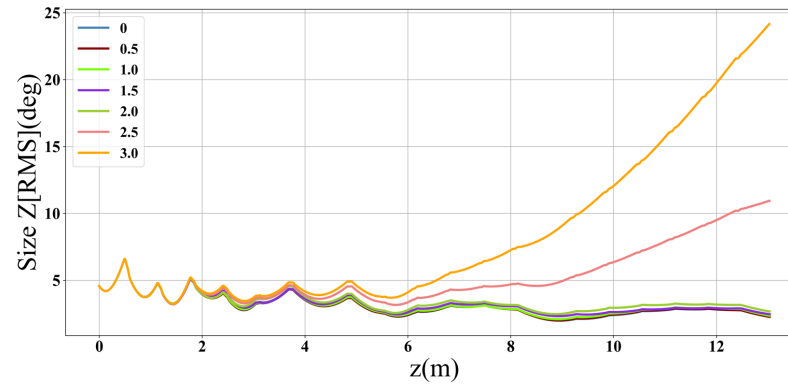
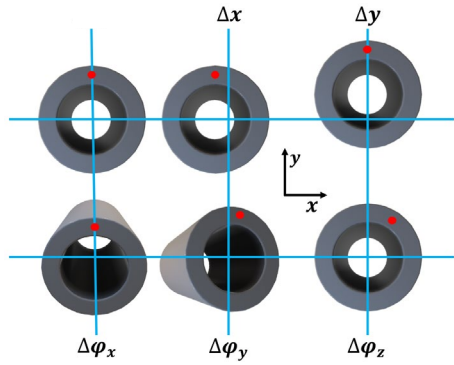
Beam parameter sensitivity analysis



Relationship between entrance bunch mismatch and mismatch along the beamline

Analysis of elements error effects on beam transport

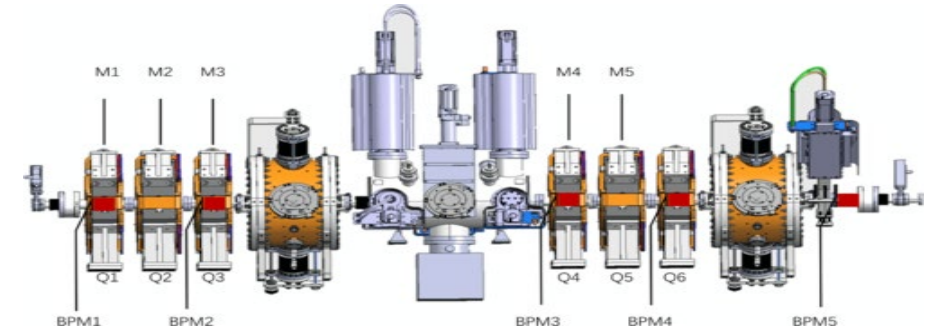
Translation and rotation errors of elements



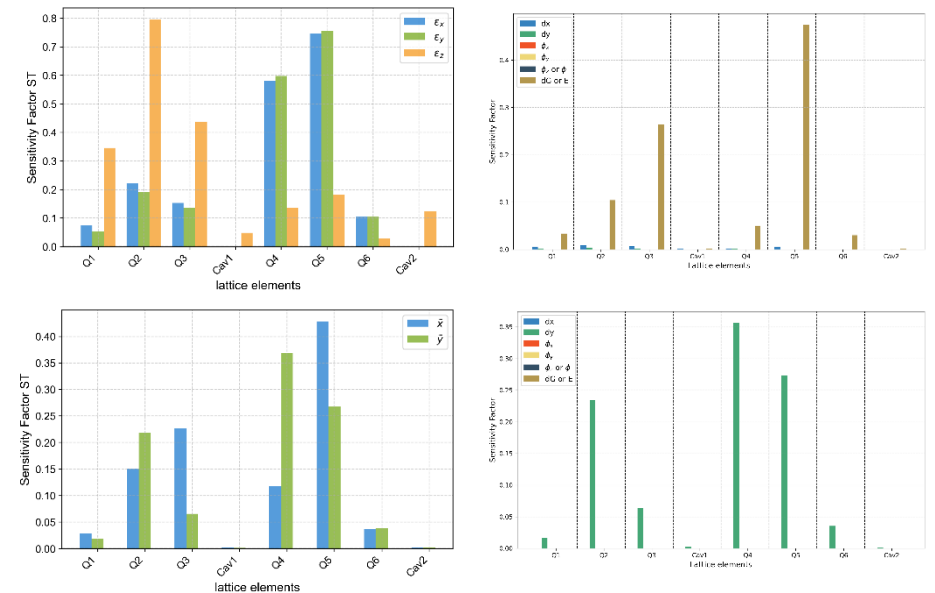
Variation of longitudinal phase width along the beamline with different RF cavity Epk and phase errors

Engineering analysis

Sensitivity-based quantitative analysis of the error-beam quality relationship

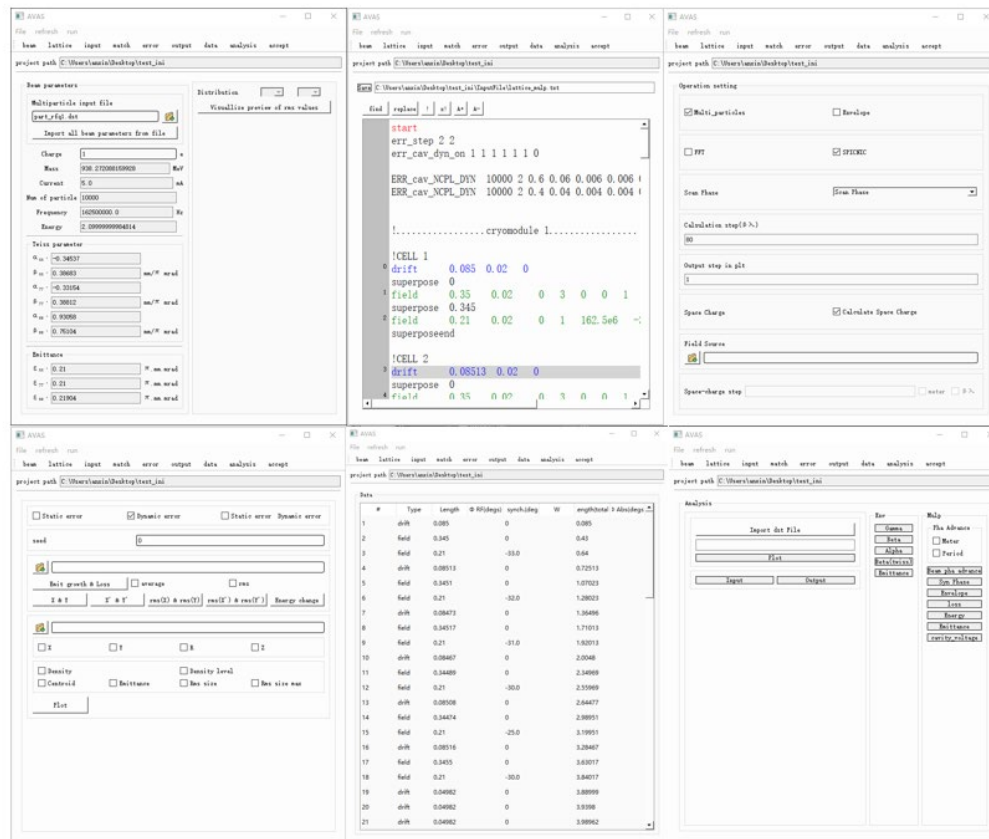


Engineering analysis of the CAFE2 MEBT section

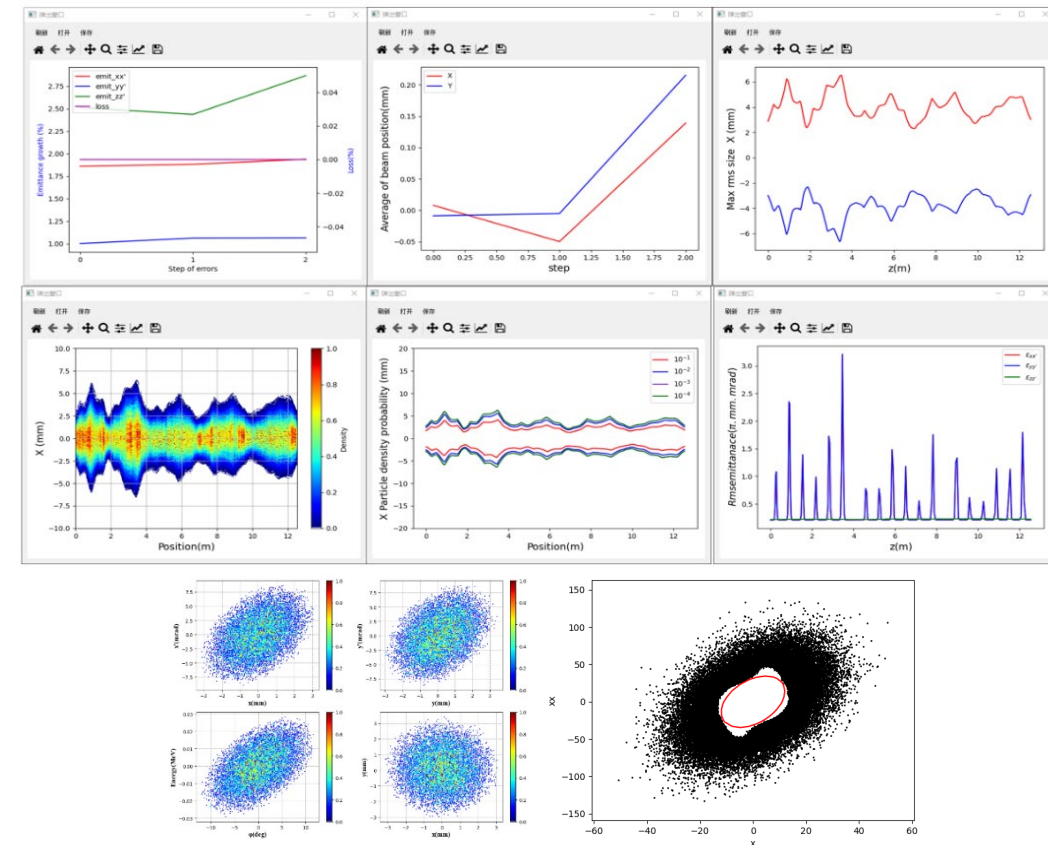


Desktop software

AVAS is a cross-platform software supporting Windows and Linux systems, incorporating a variety of accelerator elements. It features comprehensive capabilities including multi-particle tracking simulation, error analysis, acceptance calculation, transport of particles with large energy spread and angular divergence, and mixed-species beam transport.



Graphical user interface



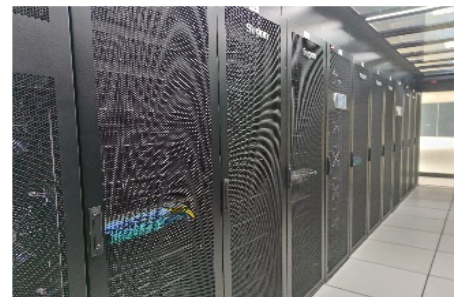
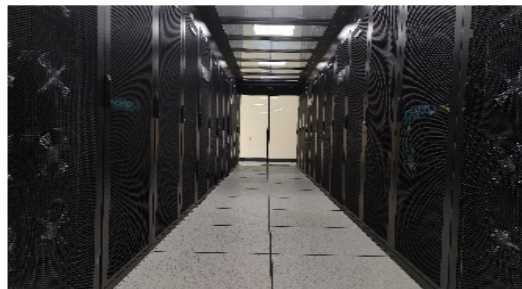
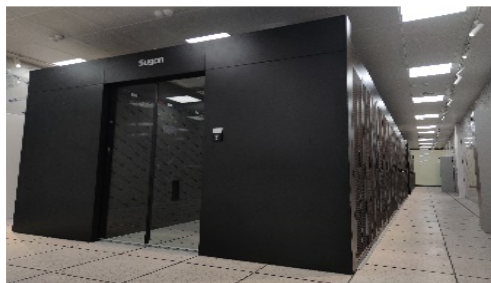
Data post-processing

Cloud simulation platform

Computing resources of Dongjiang Laboratory Supercomputing Center

CPU Infrastructure: 14,000+ cores providing peak performance of 1.820 PFLOPS in single precision (FP32) and 0.910 PFLOPS in double precision (FP64);

GPU Infrastructure: 72 NVIDIA A100 40GB SMX4 and 88 A100 40GB PCIe accelerators with aggregate performance exceeding 0.456 PFLOPS (FP32) and 0.228 PFLOPS (FP64).



CiADS superconducting segment

N_p : 1,000,000

Length: 202.75m

Beam intensity: 5mA

Frequency: 162.5MHz

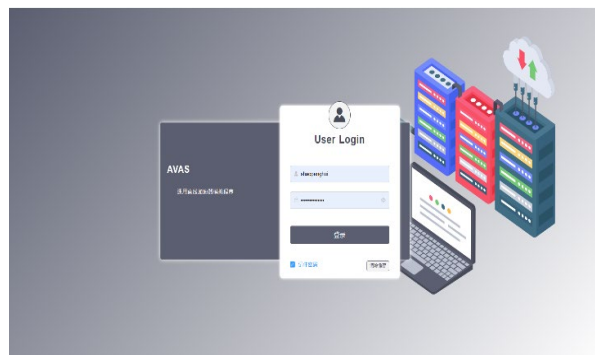
Initial energy: 2.1MeV

time step: $\Delta t = 1/(f * 100)$

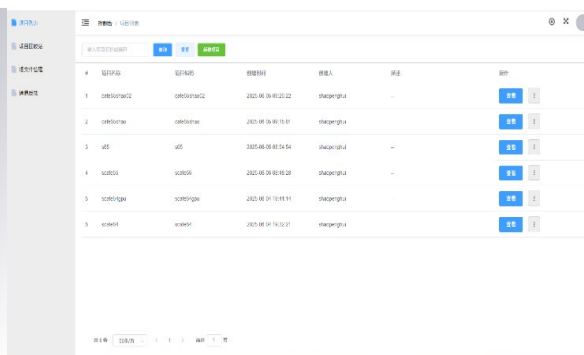
Output energy: 621.1MeV

Simulation time

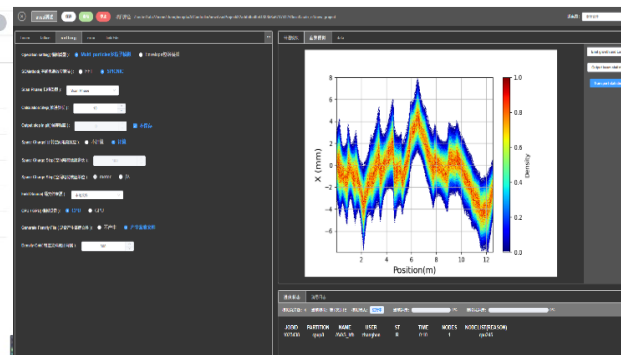
7.04h $\rightarrow$ 45.98s



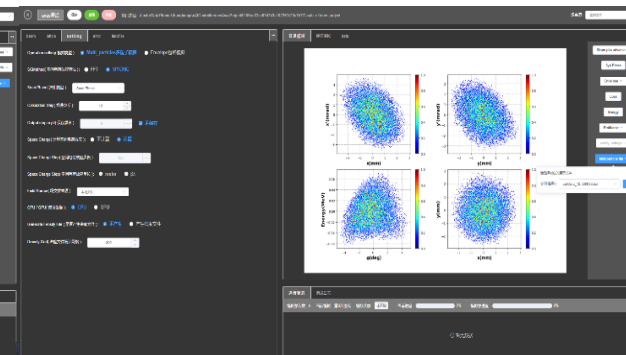
User login



Project management



Parameter Settings



Data post-processing



Summary



群聊: avas用户交流群



该二维码7天内(11月6日前)有效, 重新进入将更新

Key Innovations

- **S-PICNIC Algorithm:** $3.9\times$ speedup by exploiting bunch symmetry
- **Hybrid T-code/Z-code:** Seamless switching between tracking modes for optimal accuracy
- **GPU Acceleration (AVASX):** $470\text{-}550\times$ faster than CPU, reducing simulation time from hours to seconds
- **Physics-Guided ML:** Super-resolution model ($R^2=0.991$) enabling fast multi-particle predictions

Verification & Applications

- Validated against analytical solutions, TraceWin, Impact-T ($< 1.5\%$ deviation)
- Experimental verification: $< 0.5\%$ energy error, no beam loss
- Applications: Error analysis, multi-bunch transport, mixed-species beams, muon beamlines

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lizhongyi@gdlhz.ac.cn

(Function development)

tianyuan08@impcas.ac.cn

(GPU- Accelerated)

jinchao@impcas.ac.cn

(Physical modeling)



Thanks for your attention

Email: jinchao@impcas.ac.cn