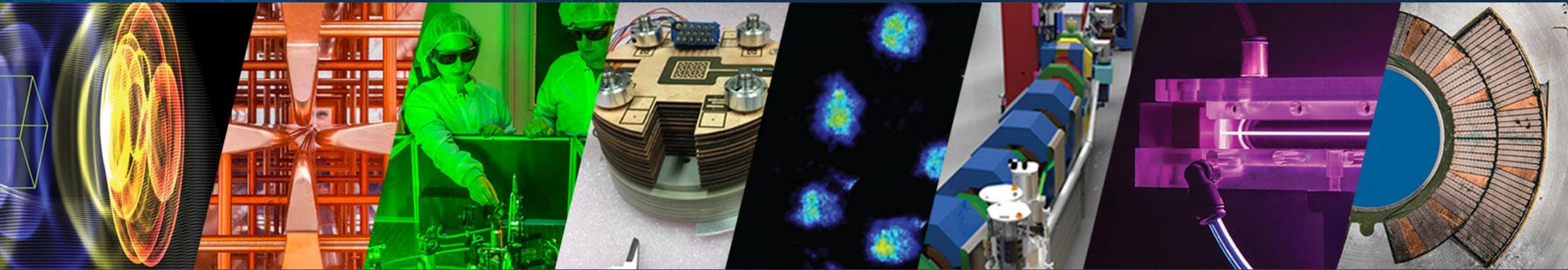


# A next generation, integrated community toolset for the modeling of linear colliders

Jean-Luc Vay, Arianna Formenti, Marco Garten, Axel Huebl, Remi Lehe,  
Chad Mitchell, Ji Qiang, Olga Shapoval, Edoardo Zoni



International Workshop on Future Linear Colliders, LCWS2024  
July 8-11, 2024 – The University of Tokyo, Japan



U.S. DEPARTMENT OF  
**ENERGY**

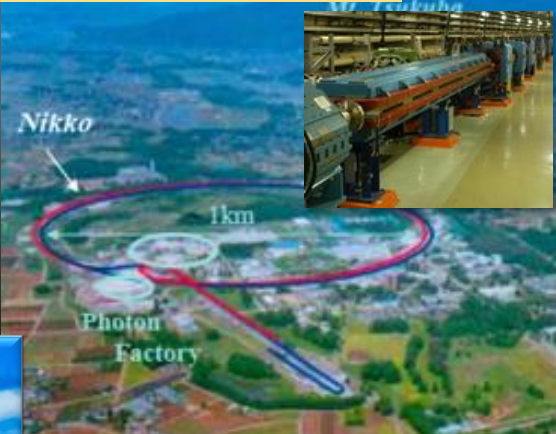
Office of  
Science

# Berkeley Lab has long been home to state-of-the-art modeling of particle accelerators

CERN (HL-)LHC, PS, SPS



KEK KEKB



FNAL Tevatron, PIP-II, IOTA, ...



LBNL ALS(-U), BELLA



SLAC LCLS(-II-HE), FACET(-II)



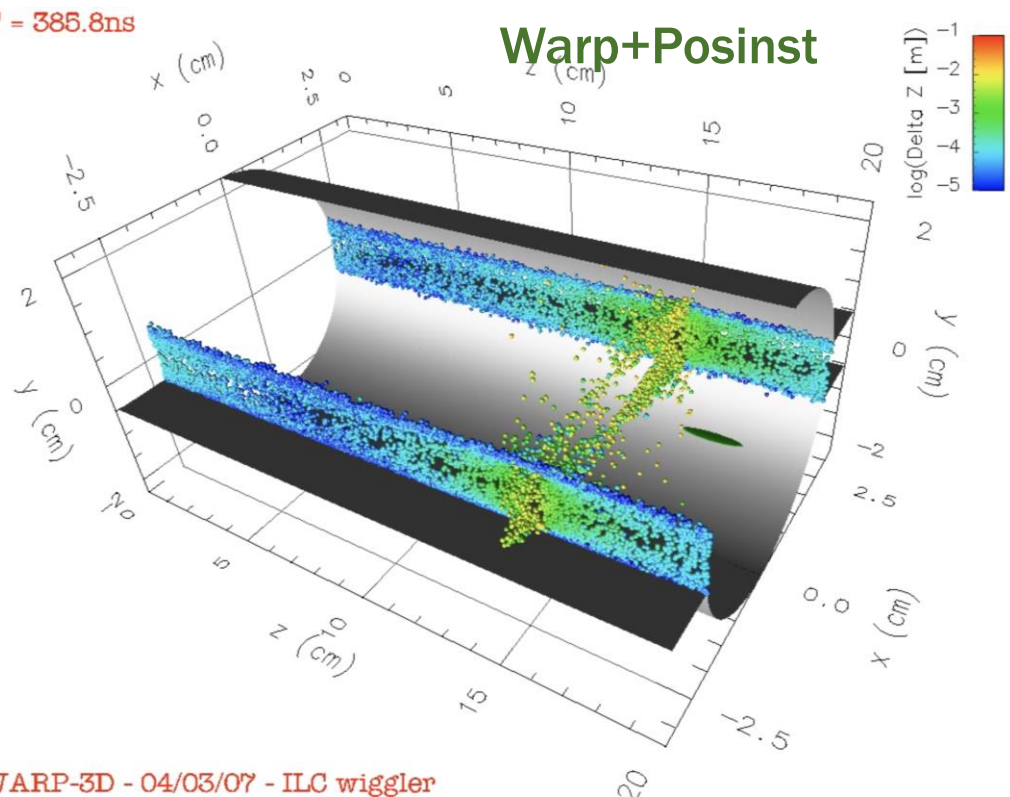
BNL RHIC, EIC



## Sample of applications

# Berkeley Lab has long been home to state-of-the-art modeling of particle accelerators

T = 385.8ns

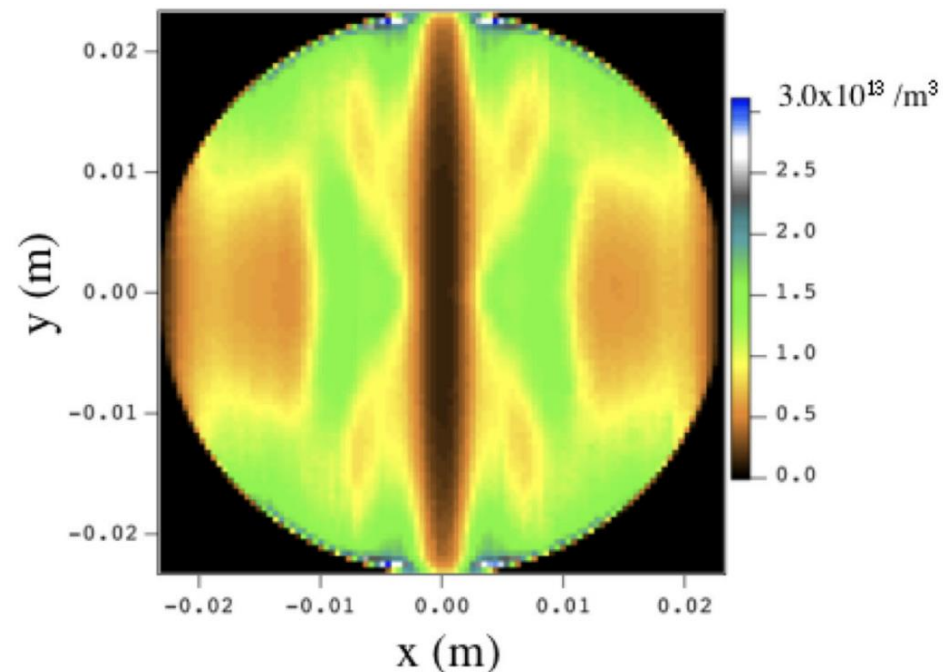


WARP-3D - 04/03/07 - ILC wiggler

## PIC calculations of the e-cloud in the ILC positron damping ring wigglers

C. M. Celata, M. A. Furman, J.-L. Vay, D. P. Grote, *Proc. PAC07* (2007)  
C. M. Celata, M. A. Furman, J.-L. Vay, D. P. Grote, *Proc. ELOUD07* (2007)

Posinst

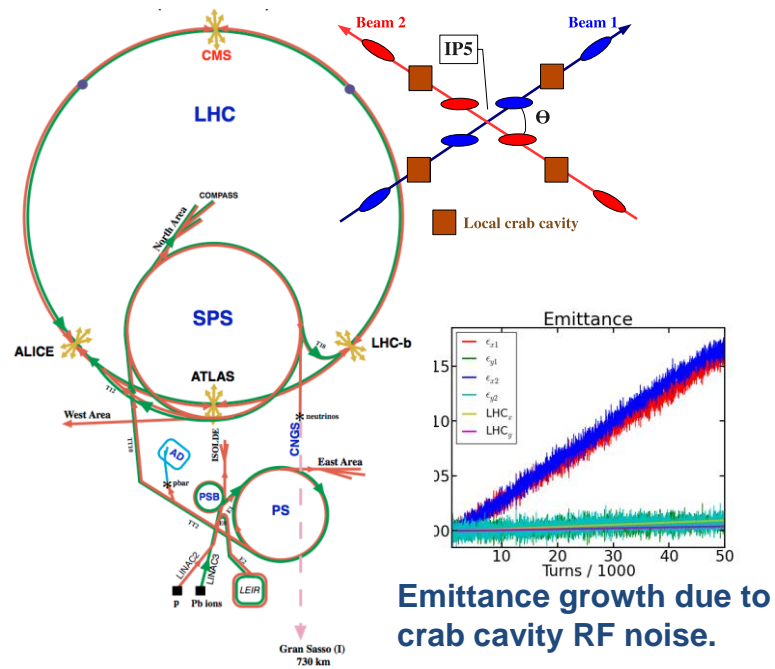


## Electron cloud cyclotron resonances in the presence of a short-bunch-length relativistic beam

C. M. Celata, M. A. Furman, J.-L. Vay, J. W. Yu, *PRAB* 11, 091002 (2008)

# Berkeley Lab has long been home to state-of-the-art modeling of particle accelerators

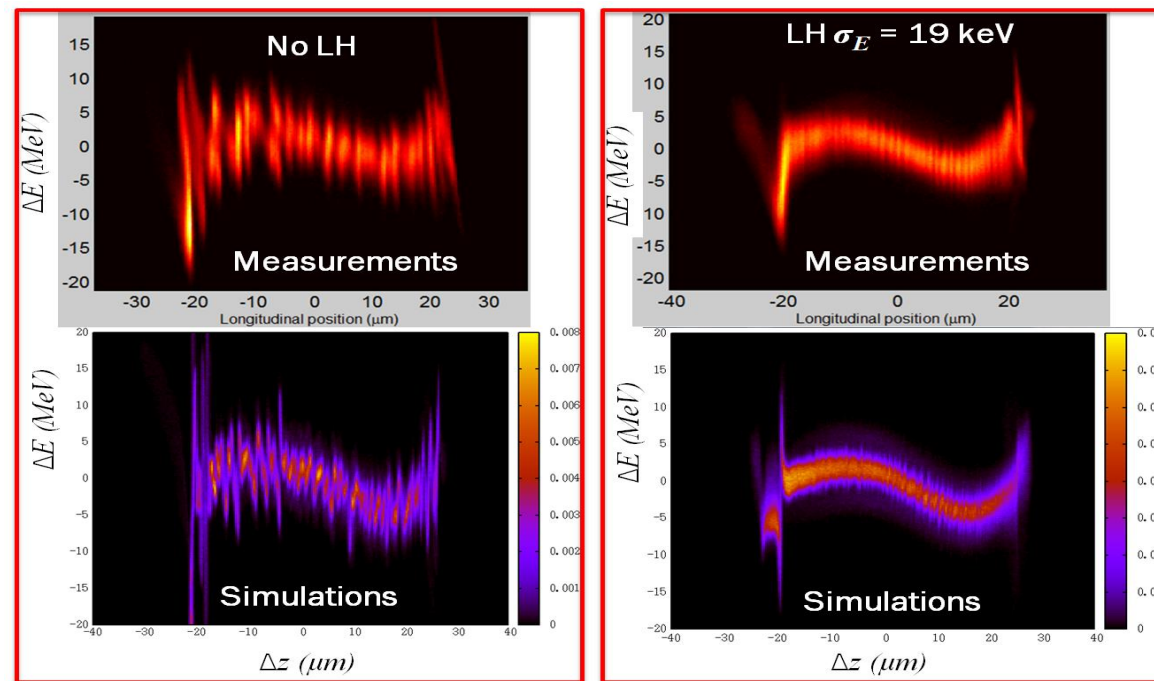
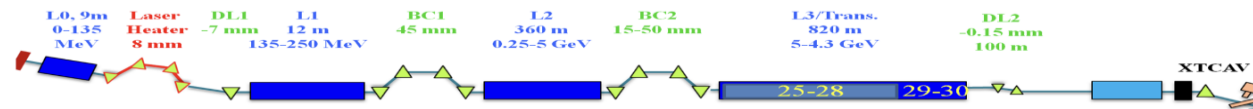
## BeamBeam3D



Strong beam-beam simulation of interactions in the LHC upgrade.

J. Qiang *et al.*, *Proc. IPAC 2015*, Richmond, VA (2015).

## Impact-T + Impact-Z (+ Genesis)



Start-to-end, one-to-one modeling reproduces microbunching in the LCLS X-ray FEL.

J. Qiang *et al.*, *Phys. Rev. Accel. Beams* **20**, 054402 (2017).

# The *Berkeley Lab Accelerator Simulation Toolkit* was created to coordinate Berkeley Lab codes



blast.lbl.gov  
Open Source



2014

## Codes:

- BeamBeam3D
- Impact-T, Impact-Z
- Marylie/Impact
- Posinst
- Warp

## Applications:

### Start-to-end accelerators

- beams, plasmas, lasers, structures
- rings, linacs, sources, injectors
- RF, plasma, dielectric acceleration
- conventional & plasma-based focusing
- e-cloud
- CSR
- cooling
- collisions
- beam-beam @ IP
- ...

## Codes were successfully used by accelerator community on major projects but:

- coordination of development was limited by legacy
  - amount of duplication was growing
  - number of additional physics modules, codes & contributions (from various institutions) was increasing
  - need to modernize codes for increasing number of levels of parallelism and GPUs
- ➔ **new (more inclusive) name & coordination of codes development.**

# The *B*eam, *p*Lasma & *A*ccelerator *S*imulation *T*oolkit followed to better coordinate & modernize codes



blast.lbl.gov  
Open Source

2024

## Codes:

- BeamBeam3D
- Impact-T, Impact-Z
- Marylie/Impact
- Posinst
- Warp
- FBPIC
- HiPACE++
- ImpactX
- LW3D
- Wake-T
- WarpX

## Applications:

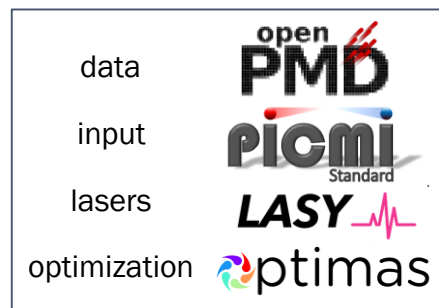
### Start-to-end accelerators

- beams, plasmas, lasers, structures
- rings, linacs, sources, injectors
- RF, plasma, dielectric acceleration
- conventional & plasma-based focusing
- e-cloud
- CSR
- cooling
- collisions
- beam-beam @ IP
- QED effects
- ...

### Plasma & fusion devices

and more (astrophysics, thermionics, microelectronics, ...)

## Standards:



## Codes were successfully used by accelerator community on major projects but:

- coordination of development was limited by legacy
  - amount of duplication was growing
  - number of additional physics modules, codes & contributions (from various institutions) was increasing
  - need to modernize codes for increasing number of levels of parallelism and GPUs
- ➔ new (more inclusive) name & coordination of codes development.

BLAST is a unique suite of interoperable codes & a multi-institutional international collaboration (>80 contributors, incl. from private sector).



# Developed by an international, multidisciplinary team

physicists + applied mathematicians + computational scientists + software engineers



*over 80*  
contributors,  
incl. from the  
private sector



Jean-Luc Vay



Ji Qiang



Arianna Formenti



Marco Garten



Axel Huebl



Rémi Lehe



Chad Mitchell



Ryan Sandberg



Olga Shapoval



Edoardo Zoni



Ann Almgren



Kevin Gott



Junmin Gu



Revathi Jambunathan



Andrew Myers



Weiqun Zhang



David Grote



Justin Angus



Kale Weichmann



Germany

Maxence Thévenet



Severin Diederichs



Alexander Sinn



Ángel Ferran Pousa



Rob Shalloo



France



Igor Andriyash



Switzerland



Lorenzo Giacomel



Lixin Ge



France



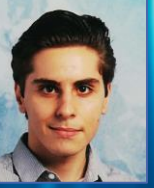
Henri Vincenti



Luca Fedeli



Thomas Clark



Pierre Bartoli



Franz Poeschel



Roelof Groenewald



# BLAST: a cutting-edge open-source simulation toolkit for end-to-end accelerator modeling

	Code	Year started	Dimensionality		Independent variable	Solver	Symplectic maps	ML surrogate elements	Language	Parallel (multimode) CPU and GPU	Multi-vendor GPU	Linac	Ring	Source	Wakefield accelerators	Beam-Beam	QED	E-cloud	IBS	CSR	Spin tracking	...
			t/s/ξ	ES/EM/QS																		
General purpose (legacy)	Warp	1989	2/3/RZ++	t/s/ξ	ES/EM/QS				For./Python	✓		✓	✓	✓							3D	...
	Impact-Z	1999	3	s	ES	✓			Fortran	✓		✓	✓								1D	...
	Impact-T	2002	3	t	ES				Fortran	✓			✓						✓			...
	Marylie/IMPACT	2006	3	s	ES	✓			Fortran	✓		✓	✓									...
Specialized	Posinst	2002	2	t	ES				Fortran									✓			...	
	BeamBeam3D	2003	2.5	t,s	ES	✓			Fortran	✓		✓			✓						...	
	FBPIC	2015	RZ++	t	EM				Python	✓	✓			✓						✓	...	
	LW3D	2018	3	t	LW				Fortran	✓											3D	...
New integrated	Wake-T	2019	RZ	ξ	QS				Python			✓		✓							1D	...
	WarpX	2016	1/2/3/RZ++	t	ES/EM		**		C++/Python	✓	✓	✓	*	**	✓	✓	✓		**		3D	**
	HiPACE++	2022	3	ξ	QS				C++/Python	✓	✓	✓			✓						✓	...
	ImpactX	2022	3	s	ES	✓	✓		C++/Python	✓	✓	✓	✓	✓				**		**	1D*, ML*	**

\* in development      ES=Electrostatic; EM=Electromagnetic; QS=Quasistatic; LW=Lienard-Wiechert; ML=Machine Learning Model  
 \*\* planned, seeking additional funding

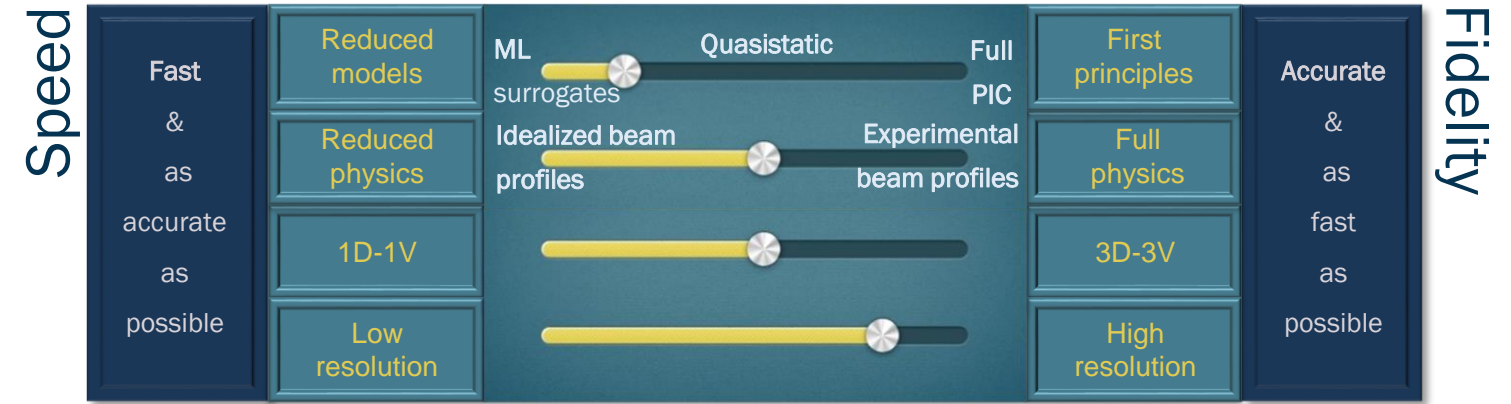
standards & workflows

Data: open PMD  
 Input: PICMI Standard  
 Lasers: LASY  
 Optimize: optimas



New codes are integrated around a common high-performance infrastructure w/CPU+GPU+ML support

- all types of colliders: Higgs factory, 10 TeV parton, muon, plasma-based, ...
  - tunability from fast modeling to detailed physics studies for collider design.
- E.g., for a plasma-based collider:  
 ML surrogate → Wake-T → HiPACE++ → WarpX

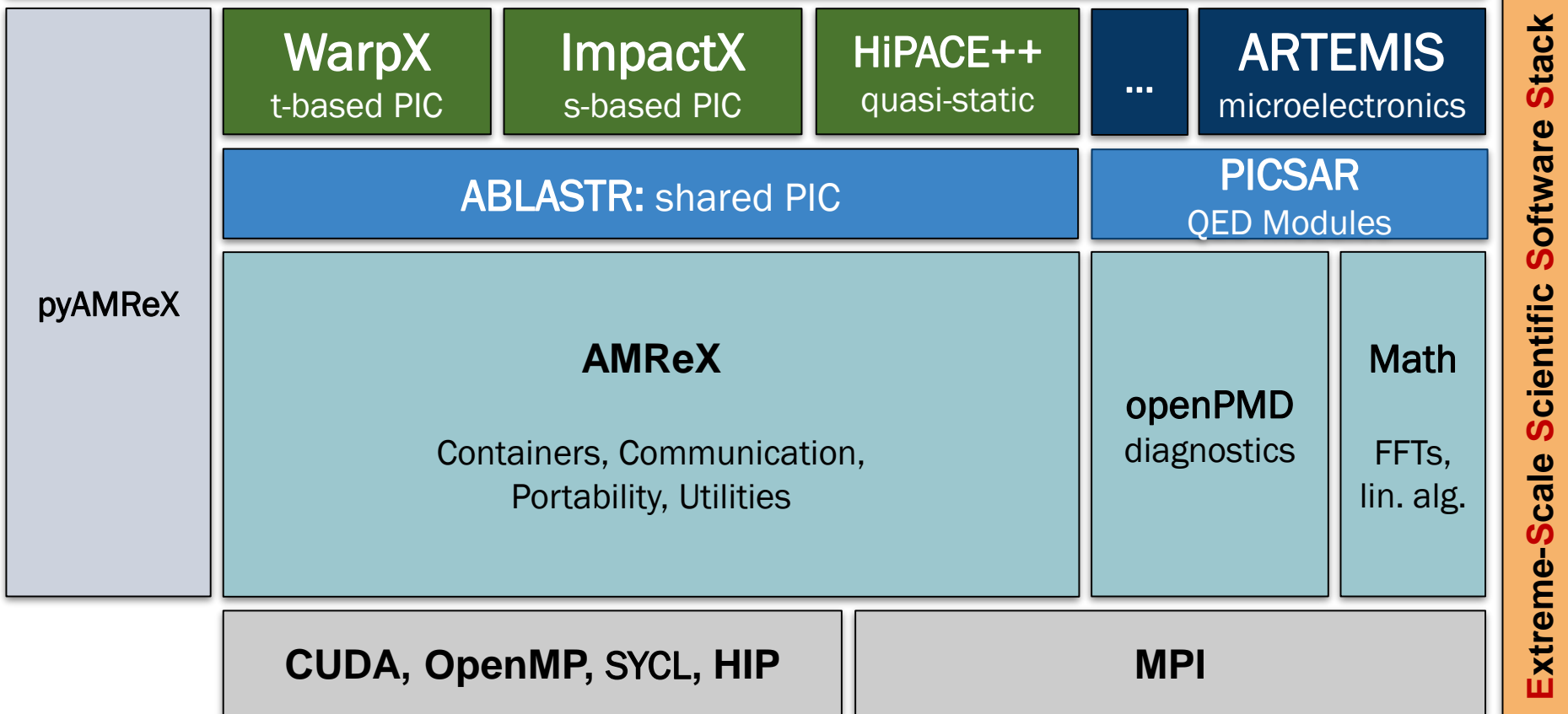




# AMP leverages the ECP technology (E4S) underpinning WarpX

→ high-performance, integrated suite for particle accelerator modeling (& more)

python™ user-steering, customization, workflows, AI/ML Frameworks (PyTorch, Tensorflow, ...)



E4S is unique in the world:  
→ advantage of unparallel performance & portability.

Python interface:

- Modular approach pioneered by Warp 20+ years ago  
→ Coupling to other codes (e.g., Posinst, ICOOL, ...)
- Access to powerful AI/ML tools

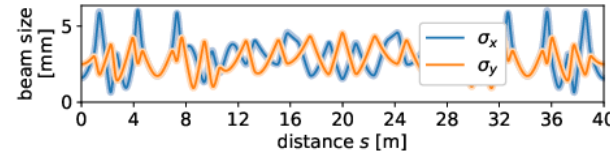
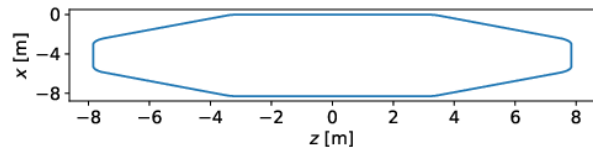


→ Propose to leverage for faster & larger scale modeling for colliders (all types) R&D

# ImpactX aims at high(er) performance modeling of RF Accelerator Modeling

## Beam-Dynamics in Linacs, Rings, Colliders

- intense beams, long-term dynamics
- **HEP colliders: Higgs factory, 10 TeV parton, ee, hh,  $\gamma\gamma$ , muons, ...**
- Example of benchmark against Impact-Z on IOTA ring beam dynamics

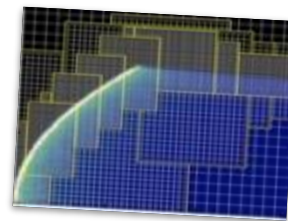


## Advanced Numerics

based on IMPACT suite of codes, esp. IMPACT-Z and MaryLie

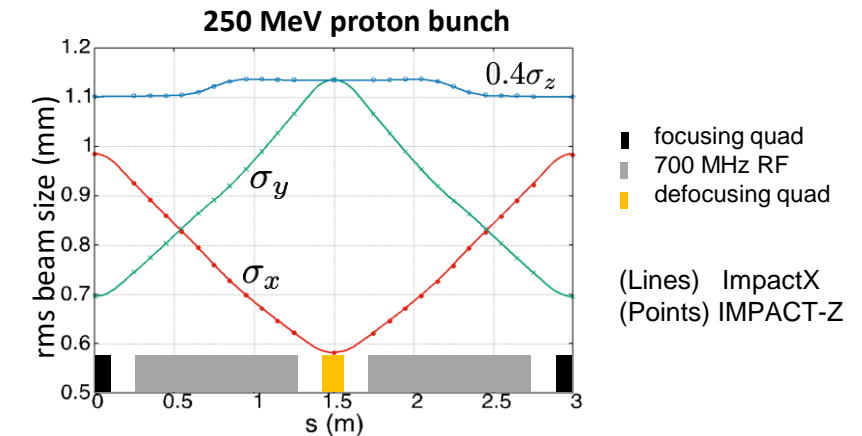
## Triple Acceleration Approach

- GPU support
- Adaptive Mesh Refinement
- AI/ML & Data Driven Models



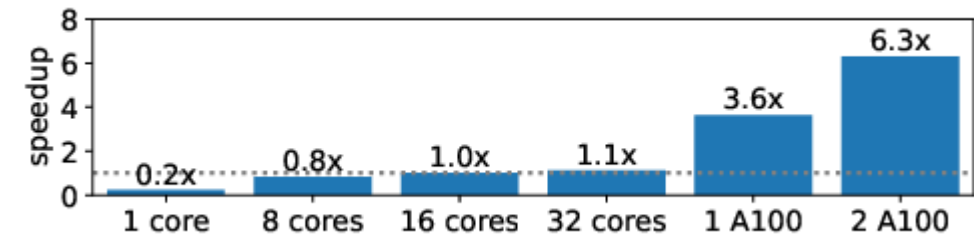
## Benchmarks & Validations

- 86 continuously run benchmarks
- code-to-code comparisons



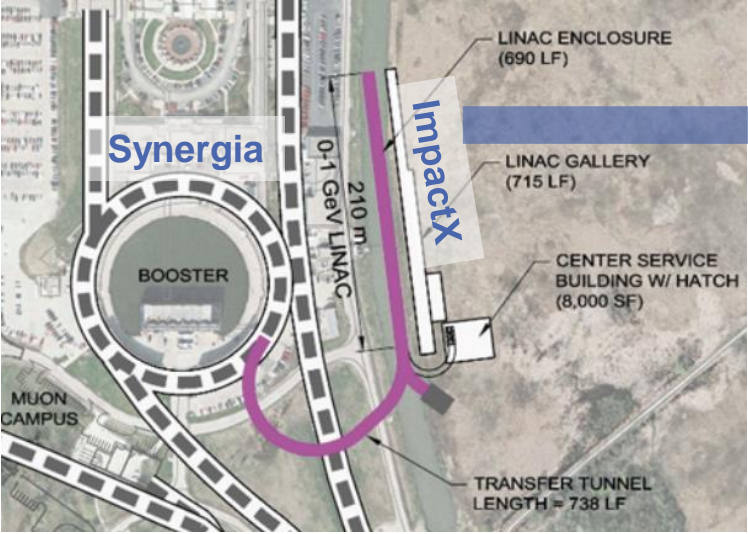
## Performance

- order-of-magnitude perf.  $\uparrow$  from GPUs



# ImpactX being applied to (& benchmarked on) high intensity beams in PIP-II

## Proton Improvement Plan II (Fermilab)

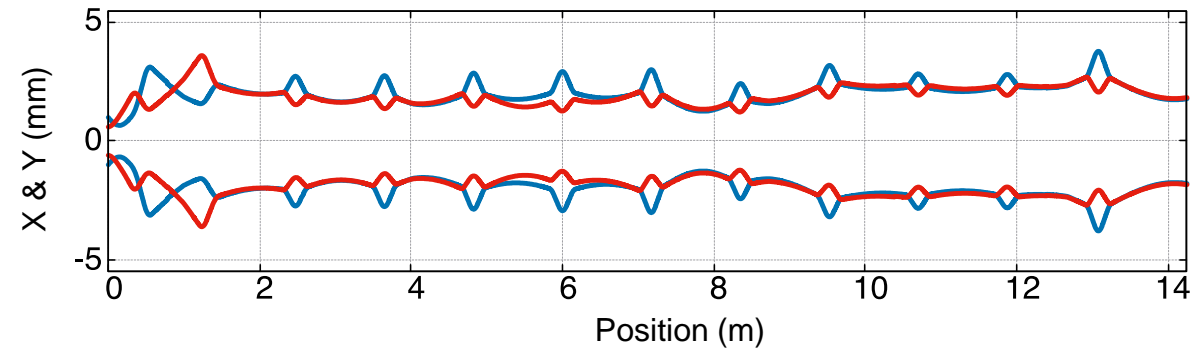
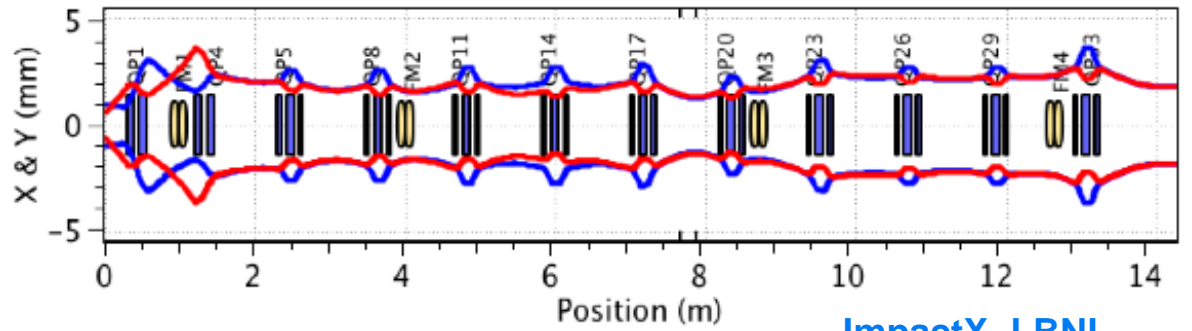


**Goal:**  
*start-to-end modeling  
 and design tuning  
 (virtual test stand)*

to Synergia for  
 modeling the Booster

## 5 mA beam modeling of the PIP-II linac

Beam envelopes MEBT transport section

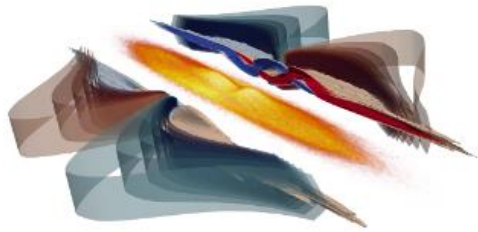


- Now: Benchmark against existing results using new parallel GPU-capable tools (ImpactX) for validation.
- Goal: Higher speed, resolution, and fidelity modeling.

# Samples of ongoing engagement in using next-generation BLAST tools to model future HEP colliders

## Collider interaction point modeling using WarpX:

- collaboration with SLAC, CEA Saclay
- benchmarked against GUINEA-PIG



WarpX simulation of beam-beam crossing at the interaction point of the ILC

See next talk by A. Formenti

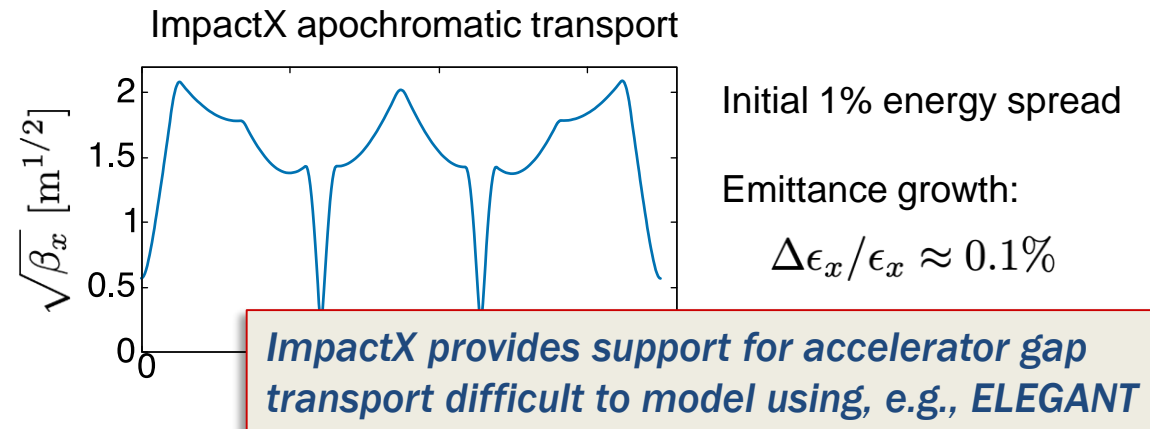
## Beam transport challenges for a laser plasma acceleration collider:

- Focusing of beams with large energy spread
- Compact transport → plasma lenses for focusing
- Preservation of beam emittance (100's of stages)
- Insensitivity to beam jitter and pointer errors

See *Advanced Accelerator Concepts 2 session: this afternoon*

## Exploring gap transport designs using ImpactX:

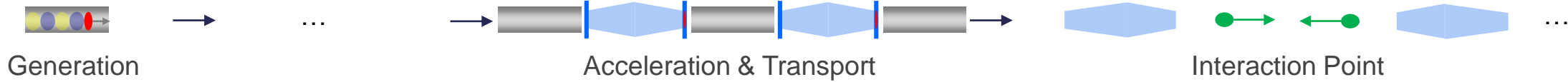
- collaboration with HALHF group (Univ. Oslo)
- aperiodic transport concept
- transversely-tapered plasma lens concept



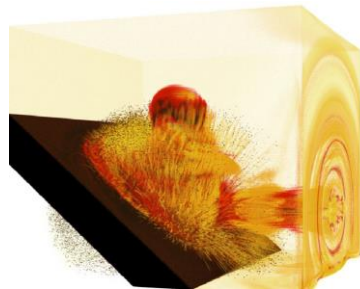
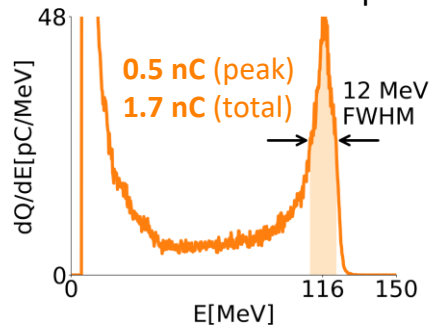
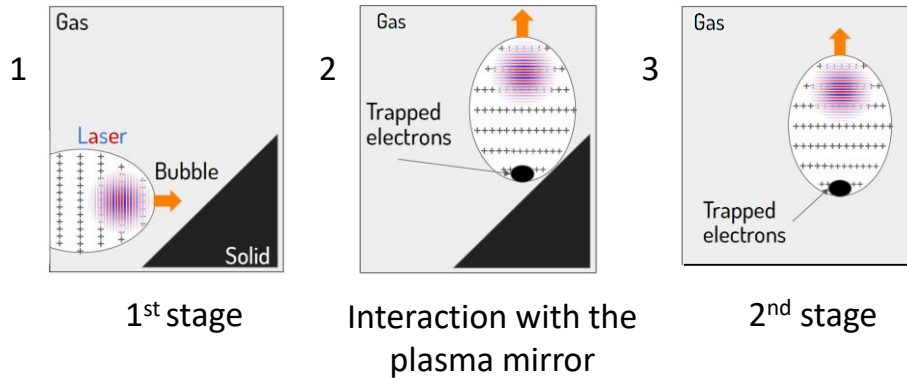
- C. Lindstrom and E. Adli, Phys. Rev. Accel. Beams 19, 071002 (2016)  
C. Lindstrom, EuroNNAc Special Topics Workshop 2022 (2022)  
C. Lindstrom *et al* incl. A. Huebl & C. Mitchell, paper in preparation

See talks on HALHF at this workshop

# BLAST codes also cover plasma-based collider modeling from source to interaction point

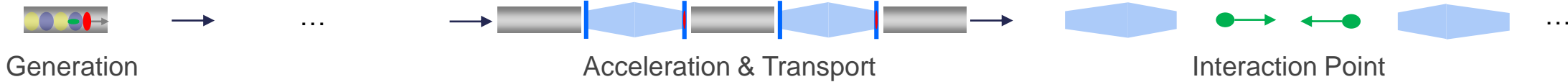


## Two-stage injection+acceleration w/ plasma mirror

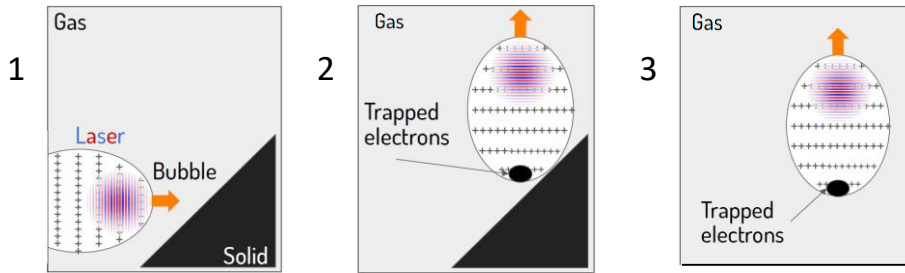


A success story of a multidisciplinary, multi-institutional team!

# BLAST codes also cover plasma-based collider modeling from source to interaction point



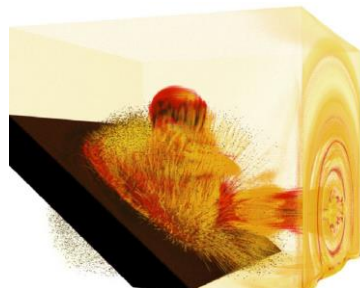
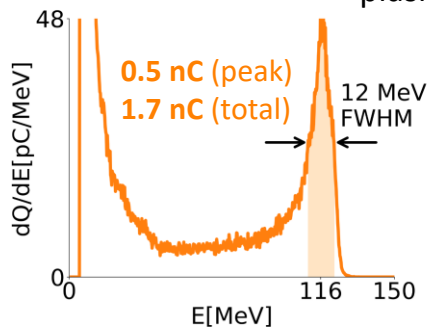
## Two-stage injection+acceleration w/ plasma mirror



1st stage

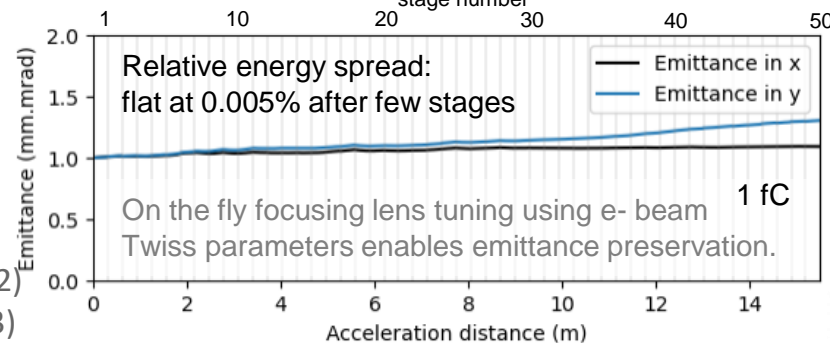
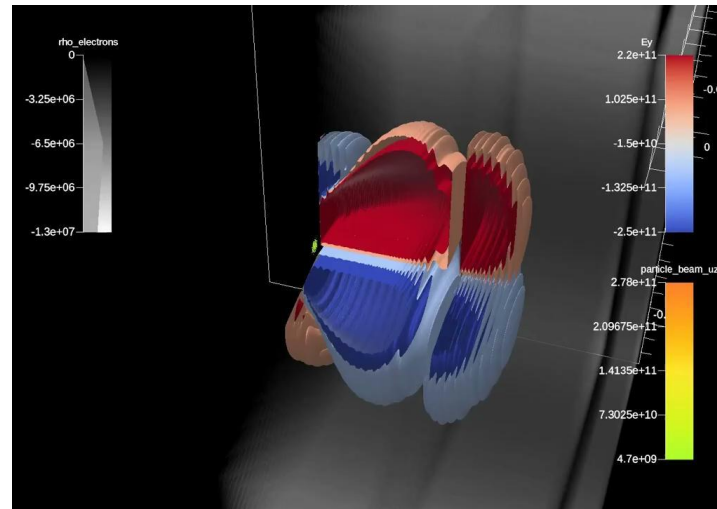
Interaction with the plasma mirror

2nd stage

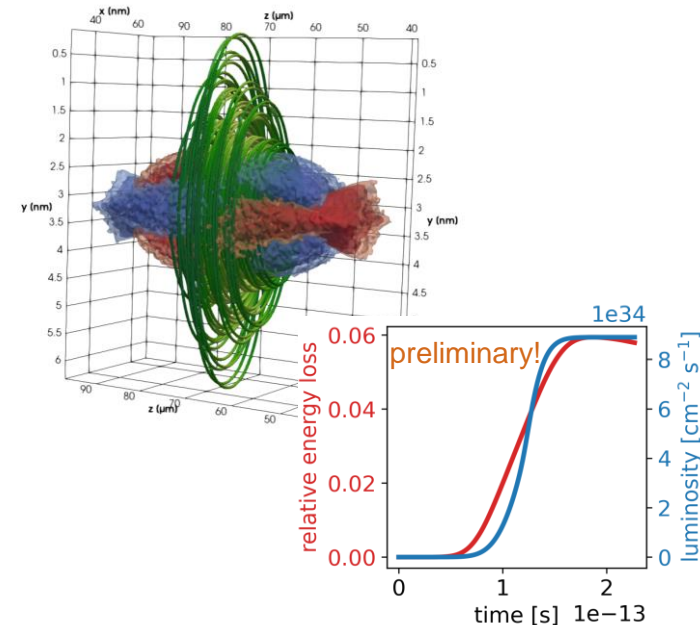


## 50 Multi-GeV LPA Stages in 3D

Ascent VTK-m *In Situ* Visualization of the first 15 stages:



## 10 TeV COM $e^-e^+$

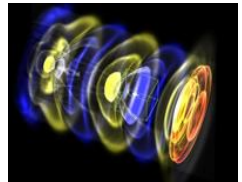
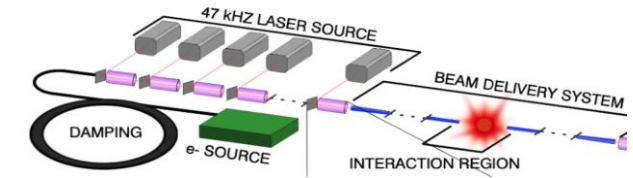
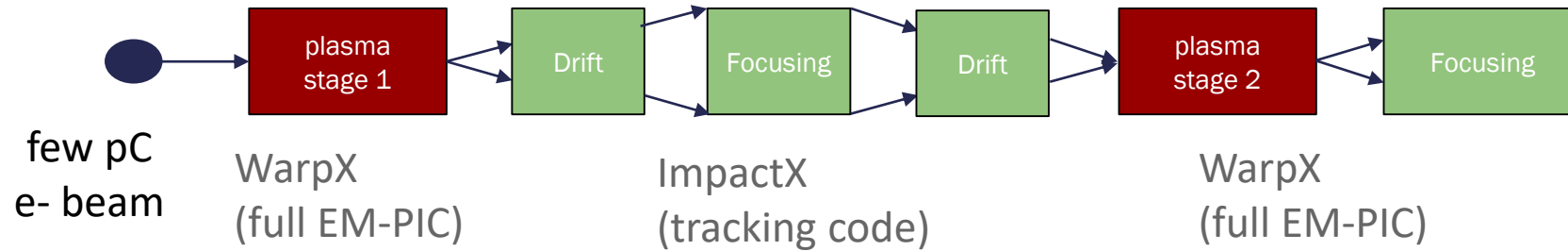


WarpX can now simulate flat, spherical, round and asymmetric beams in linear colliders:  
**ILC, C<sup>3</sup>, wakefield, HALHF, ...**  
 Next are more QED physics & **circular colliders: FCC-ee, Muons**

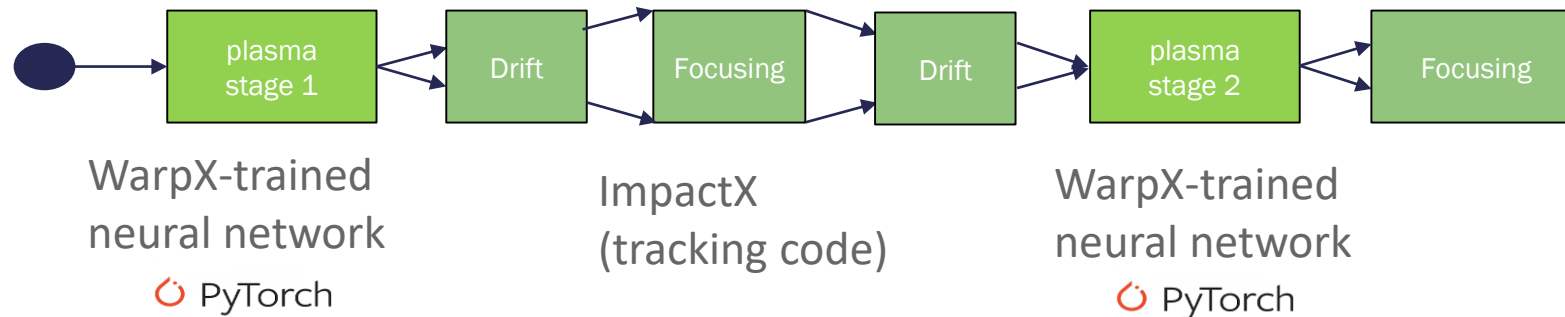
P5 Report: Exploring the Quantum Universe (2023)  
 J-L Vay et al., ISAV20 Keynote (2020) & PoP 28.2, 023105 (2021)  
 L Fedeli, A Huebl et al., SC22, ACM Gordon Bell Prize for WarpX (2022)  
 WarpX ECP MS FY23.1 & FY23.2 (2023); T Barklow et al., JINST (2023)  
 A Ferran Pousa et al., IPAC23, TUPA093 & PRAB (2023); CB Schroeder et al., JINST (2023)

# We also explore the training of ML surrogate models to speed-up simulations

- In a given lattice, some elements can be more **computationally-expensive** to model. Extreme example: laser-plasma acceleration stages



- Under certain conditions (here: negligible collective effects, specific range of parameters), computationally-expensive elements can be replaced by **ML models**, trained over **past simulations**.



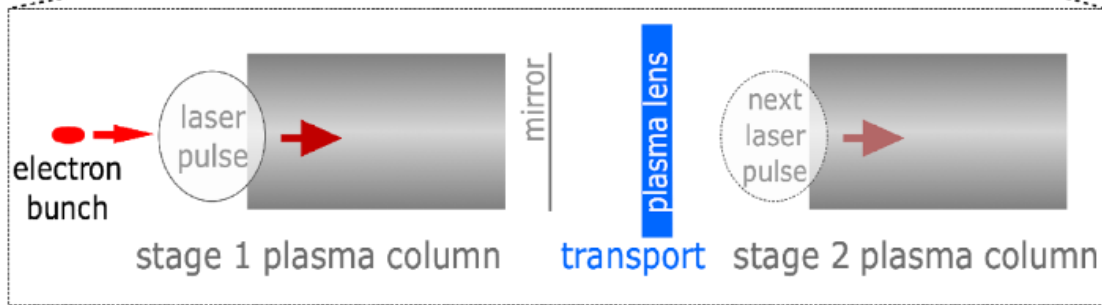
**Simulation time:** (with full geometry/physics)

hrs  
on several GPUs

<sec  
on 1 GPU

# We Exploit our High-Quality HPC Data for ML-Boosted Collider Design

**Central BLAST Code Interoperability:** Combine Plasma & RF Accelerator Elements for start-to-end modeling



**tightly-coupled LPA-neural networks inside ImpactX**

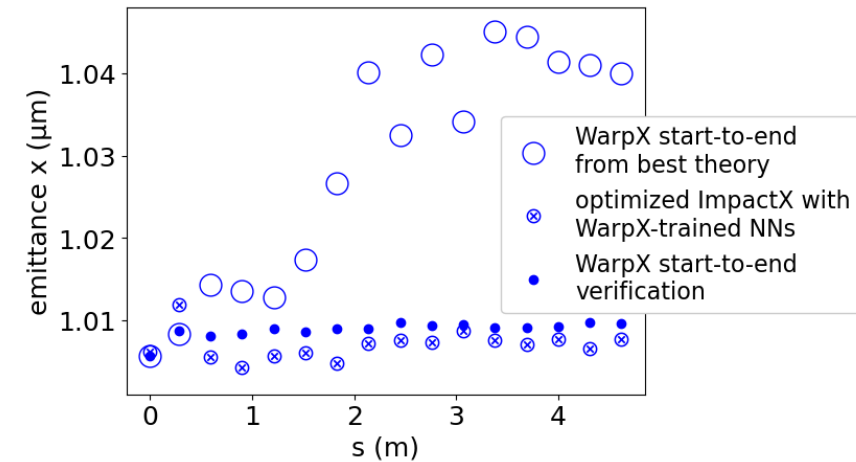


**WarpX start-to-end simulation**  
256 GPUs  
1 simulation / 5.1 hours



**ImpactX with WarpX-trained NNs**  
1 GPU  
2-4 simulations / sec

**LPA + Transport Optimization**  
with 1000s of evaluations





# Summary & Outlook

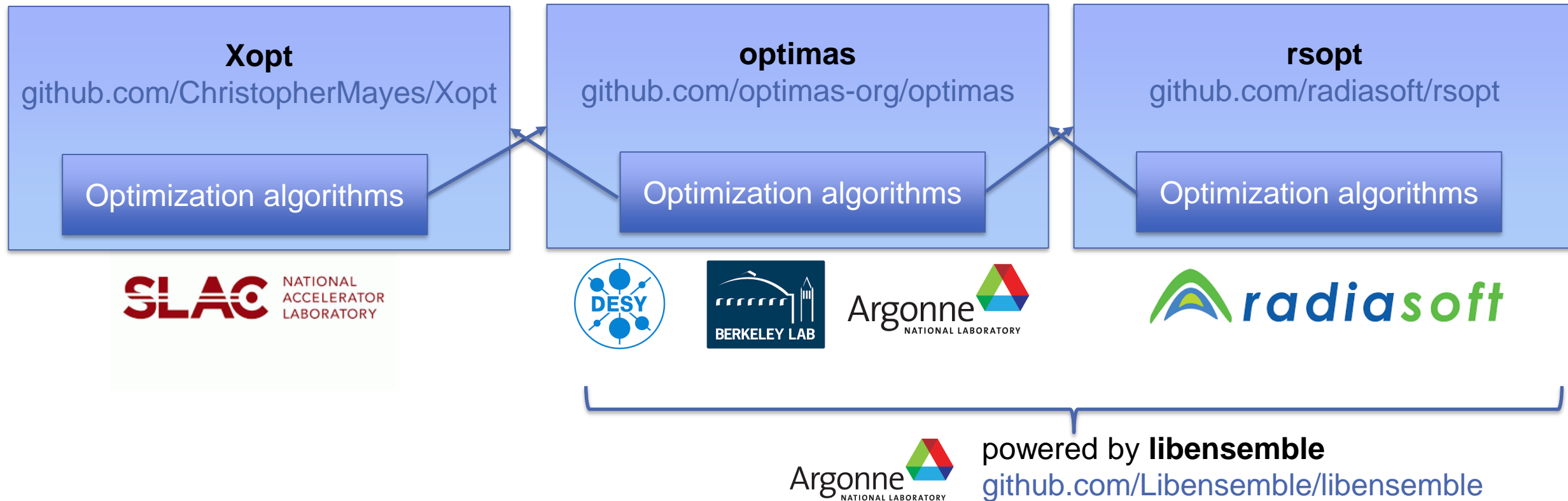
- Berkeley Lab has a long history in high performance modeling of particle accelerators
- We are leading the Beam, pLasma & Accelerator Simulation Toolkit (BLAST)
  - a coordinated effort that develops an integrated suite for start-to-end modeling of colliders
  - 80+ multidisciplinary team of contributors from labs, universities & private sector
  - leverages unique technology from the US DOE Exascale Computing Project that enables codes to be built on a common core that enables efficient simulations on CPUs and GPUs
  - common Python front-end enables user steering and direct coupling with optimization & AI/ML
- We are offering to use the new set of tools in support of the modeling of any type of future collider: Higgs factory, 10 TeV parton, muon, plasma-based, ...
- Feel free to use and contribute: <https://blast.lbl.gov/>



# Questions?

# We are fostering interoperability across open-source optimization software.

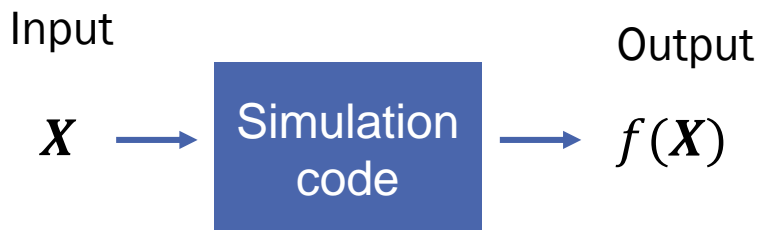
- Several **open-source optimization frameworks** are being used in the accelerator community (each with their respective strengths)



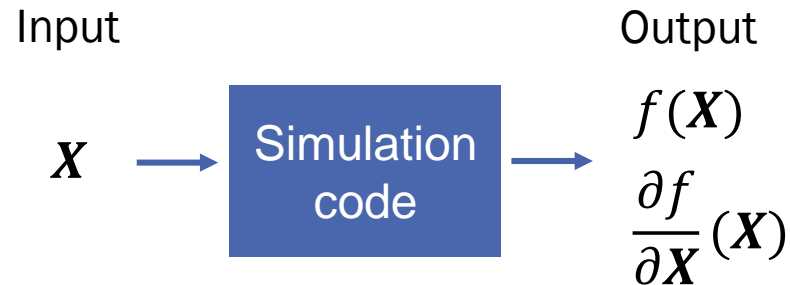
- Ongoing efforts by the developers to **standardize optimizers** and **foster interoperability**.

# Even tighter ML integration can be achieved with differentiable simulations

## Regular simulation code



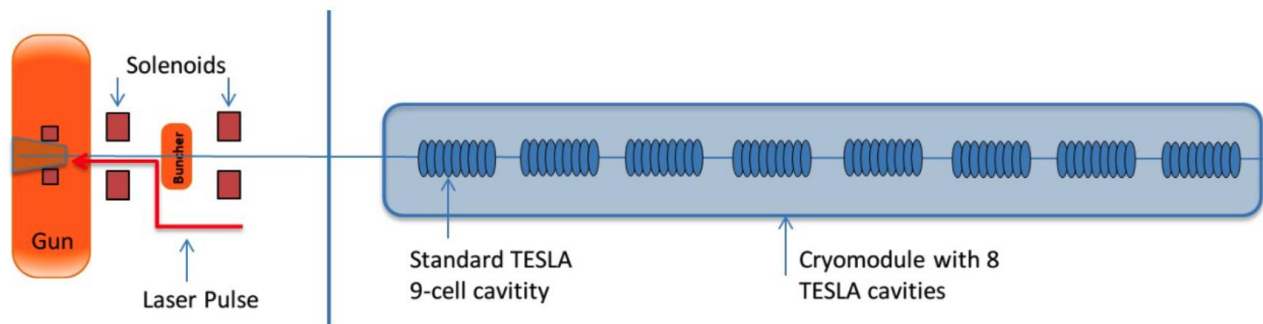
## Differentiable simulation code



### Example:

Input:  
accelerator parameters

$$\mathbf{X} = \begin{pmatrix} B_{solenoid} \\ \varphi_{RF\ cavity} \\ E_{RF\ cavity} \\ \sigma_{beam,i} \end{pmatrix}$$



$$f = \epsilon_{\perp}$$

$$\frac{\partial f}{\partial \mathbf{X}} = \begin{pmatrix} \frac{\partial \epsilon_{\perp}}{\partial B_{solenoid}} \\ \frac{\partial \epsilon_{\perp}}{\partial \varphi_{RF\ cavity}} \\ \frac{\partial \epsilon_{\perp}}{\partial E_{RF\ cavity}} \\ \frac{\partial \epsilon_{\perp}}{\partial \sigma_{beam,i}} \end{pmatrix}$$

# Differentiable codes have several advantages.

- **Sensitivity studies**

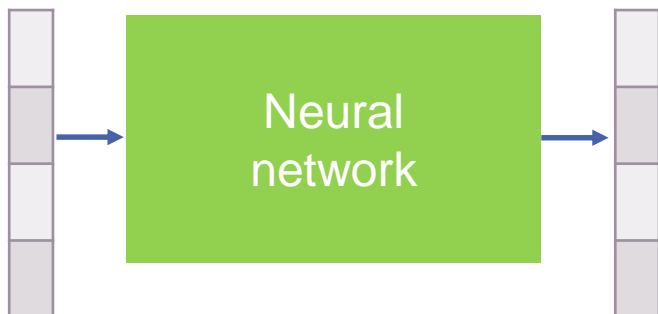
$\frac{\partial f}{\partial X}$  quantifies how sensitive the output is to the input.

- **Optimization in high-dimensional space** (e.g. of accelerator designs)

$\frac{\partial f}{\partial X}$  can be used in **gradient-based** optimizers, which often converge faster

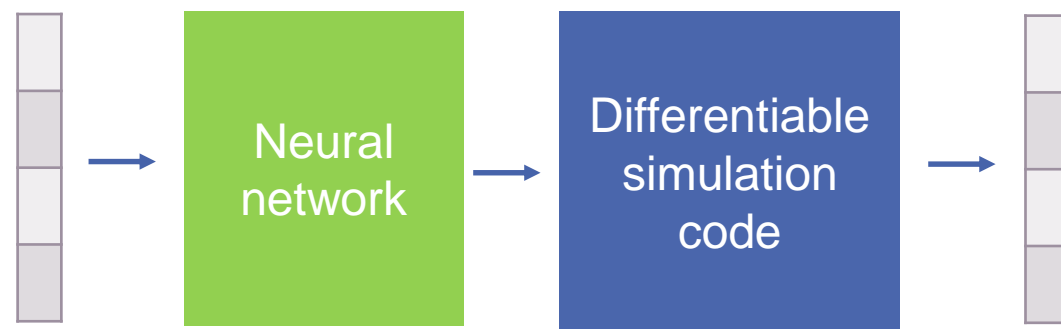
- **Allows training of a neural network that is combined with a differentiable code**

Traditional training of neural network

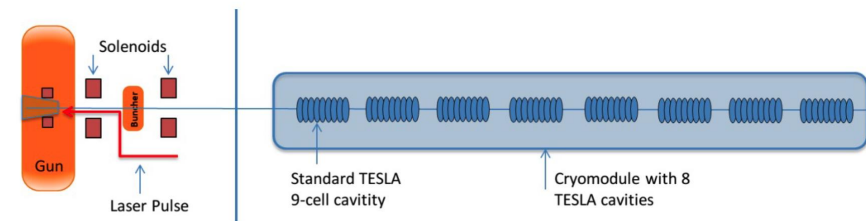


Input/output pairs, from a data set

Training of neural network combined with a code




Example: *R. Roussel et al., Phase Space Reconstruction from Accelerator Beam Measurements Using Neural Networks and Differentiable Simulations, PRL (2023)*



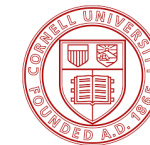
# We are exploring frameworks for differentiable codes.


- Several **algorithms** are available to make a code differentiable. e.g.  
*J. Qiang, Differentiable self-consistent space-charge simulation for accelerator design, PRAB (2023)*
- Several efforts to build differentiable **accelerator simulation codes**, based on **auto-differentiation frameworks**.

- pytorch  
 **Cheetah:** accelerator code based on pytorch  
[github.com/desy-ml/cheetah](https://github.com/desy-ml/cheetah)



- Julia  
 **Bmad-Jul** bosal to implement Bmad algorithms in Julia  
[github.com/bmad-sim](https://github.com/bmad-sim)



- Enzyme AD  
 Takes existing  $\partial_x$  code and makes it auto-differentiable at compile time.  
Could be leveraged to make BLAST codes (ImpactX, WarpX, ...) differentiable.

