

# Efficient 6-dimensional phase space reconstructions from experimental measurements using generative machine learning

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U.S. DEPARTMENT OF  
**ENERGY**

Stanford  
University



NATIONAL  
ACCELERATOR  
LABORATORY

# Modern Challenges in Beam Control

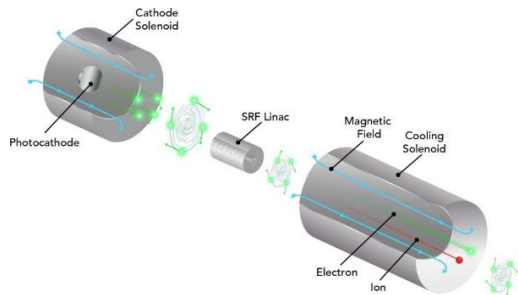
Modern accelerator developments require **precise control** over **high dimensional** beam structure

$$x = (x, p_x, y, p_y, z, p_z)^T$$

**RMS quantities** → **Exact beam structure**

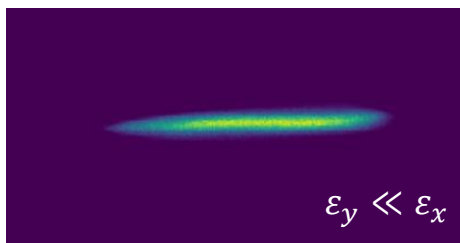
● Macroscopic scale    
 ● Mesoscopic scale    
 ● Microscopic scale

Magnetized beams for hadron cooling @ EIC



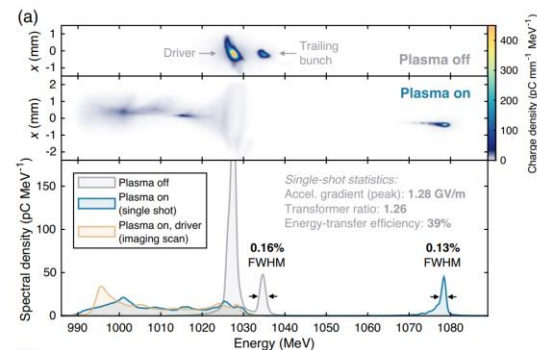
Wijethunga, S. A. K., et al. NIMA 1051 (2023): 168194.

Flat beams for colliders



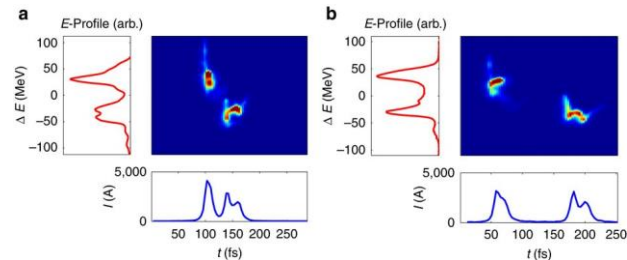
Chen, P., and K. Yokoya. (1991): 91-2.

Beam shaping for high efficiency PWFA



C. A. Lindström, et al. Phys. Rev. Lett. **126**, 014801

Free electron lasers



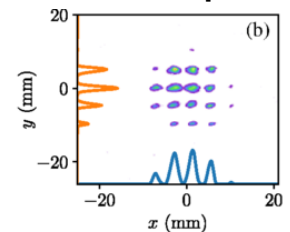
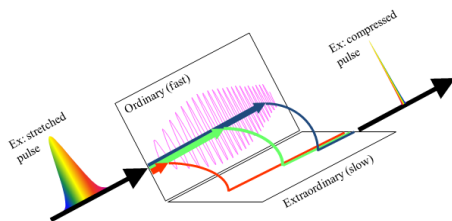
Marinelli, A., et al. Nature communications 6.1 (2015): 6369.

# Methods for 6D Beam Shaping

## Argonne Wakefield Accelerator (AWA)

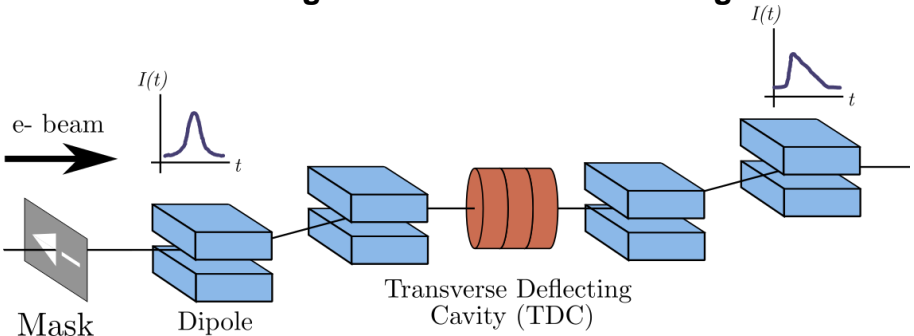


## Temporal laser pulse shaping    Transverse laser pulse shaping



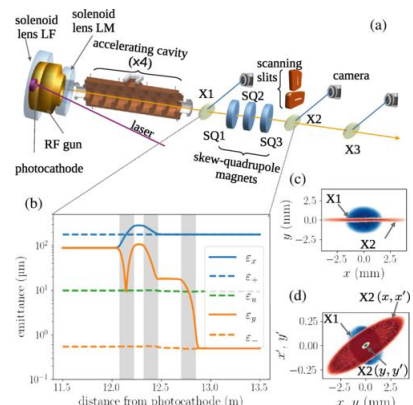
Halavanau, A., et al. PRAB 22.11 (2019): 114401.

## Transverse-to-longitudinal emittance exchange



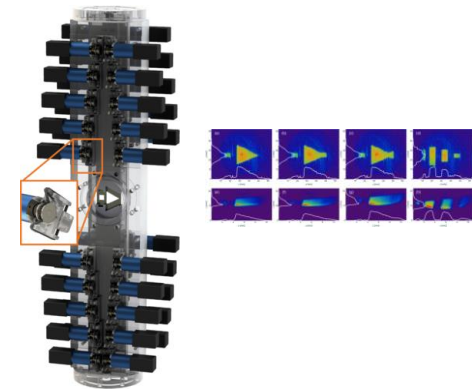
Emma, P., et al. PRAB 9.10 (2006): 100702.

## Round-to-flat beam transformation



Xu, T., et al. PRAB 25.4 (2022): 044001.

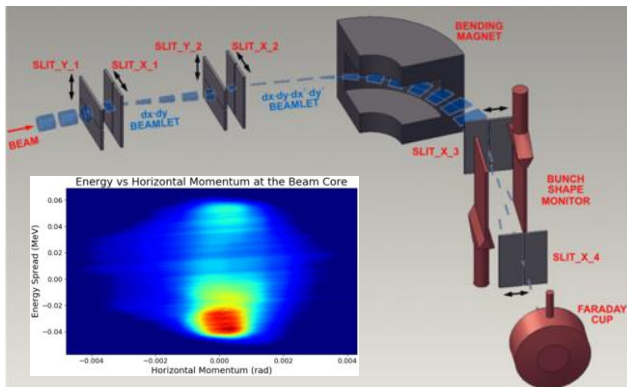
## Adjustable transverse beam masking



Majernik, N., et al. PRAB 26.2 (2023): 022801.

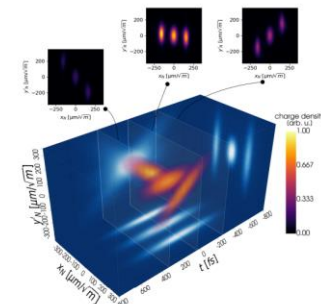
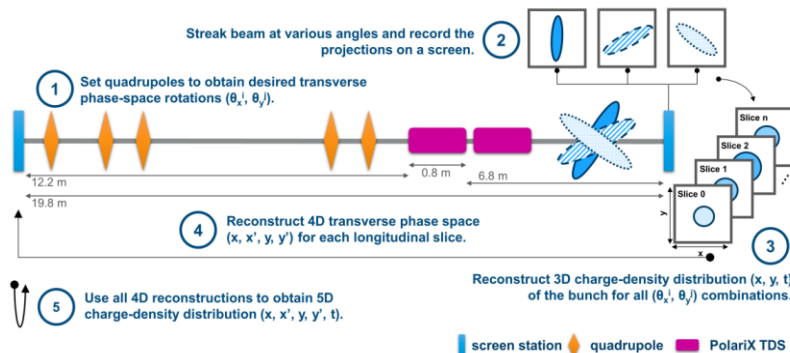
# Measuring 6D Beam Distributions

Multiple scanning slits with dipole and bunch shape monitor @ SNS (ORNL)



Cathey, Brandon, et al. PRL 121.6 (2018): 064804.

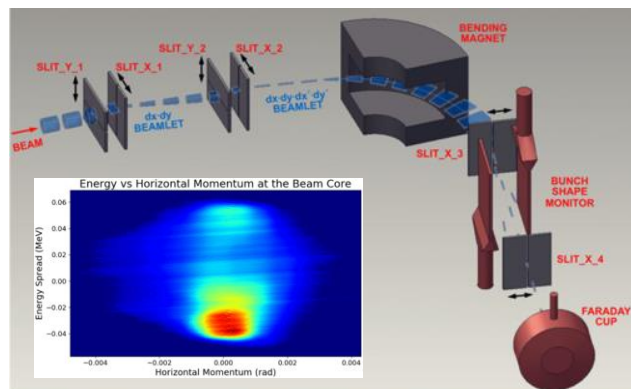
5D tomography with polarizable X-band TDC @ ARES (DESY)



Jaster-Merz, S., et al Proc. IPAC'22 (2022): 279-283.

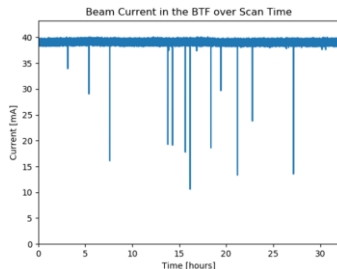
# Progress on Measuring 6D Beam Distributions

Multiple scanning slits with dipole and bunch shape monitor @ SNS (ORNL)

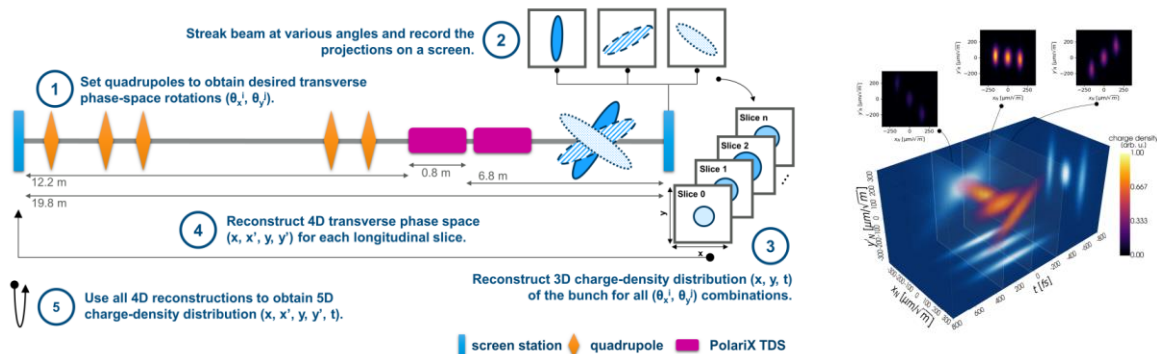


Cathey, Brandon, et al. PRL 121.6 (2018): 064804.

5 million (!) measurements ~ 32 hrs



5D tomography with polarizable X-band TDC @ ARES (DESY)

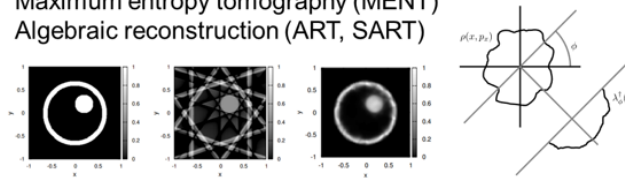


Jaster-Merz, S., et al Proc. IPAC'22 (2022): 279-283.

960 measurements ~ 28 hrs

Tomographic reconstruction ~ 24 hrs

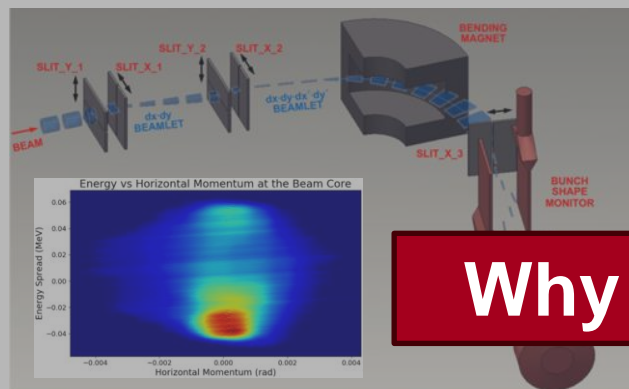
- Maximum entropy tomography (MENT)
- Algebraic reconstruction (ART, SART)



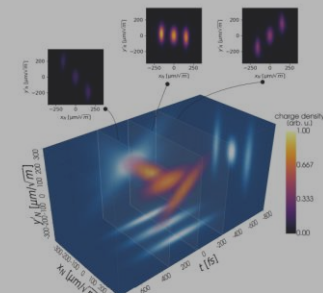
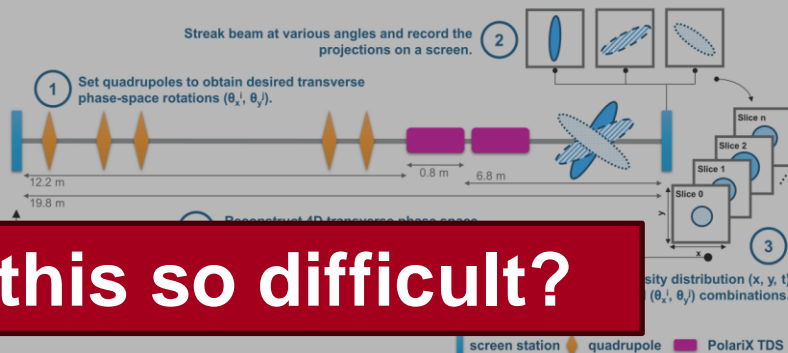
Hock K. and Ibison M., JINST, 2013

# Progress on Measuring 6D Beam Distributions

Multiple scanning slits with dipole and bunch shape monitor @ SNS (ORNL)



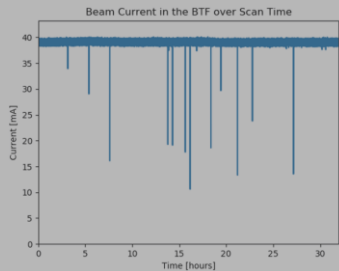
5D tomography with polarizable X-band TDC @ ARES (DESY)



**Why is this so difficult?**

Cathey, Brandon, et al. PRL 121.6 (2018): 064804.

5 million (!) measurements ~ 32 hrs

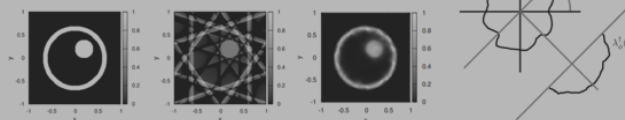


Jaster-Merz, S., et al Proc. IPAC'22 (2022): 279-283.

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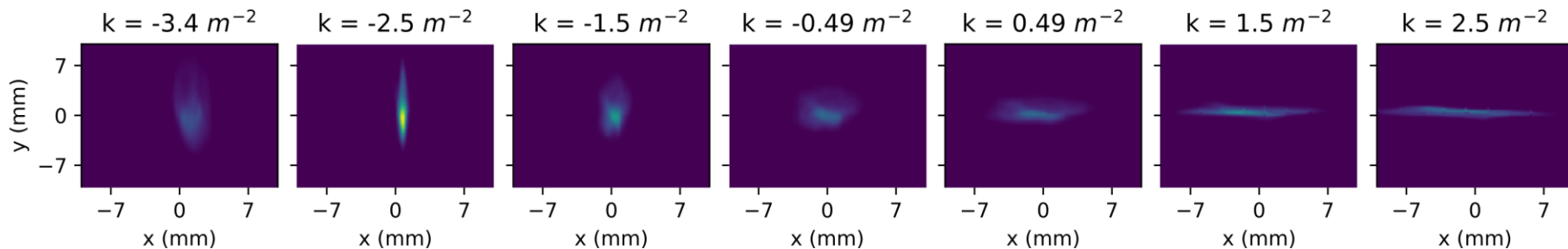
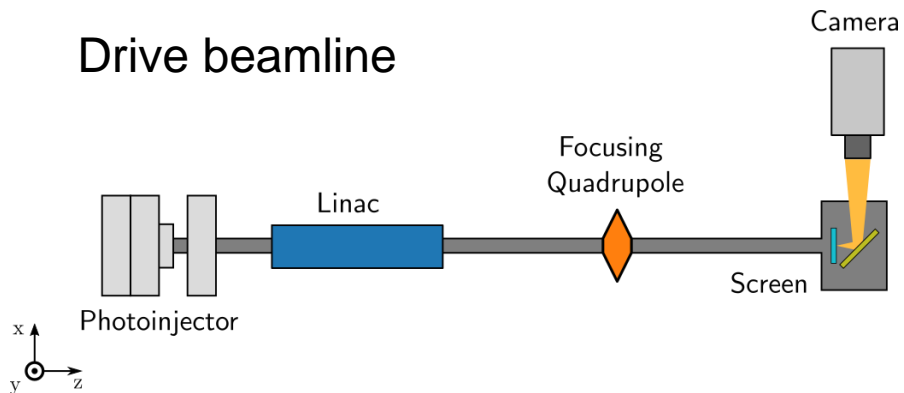
- Maximum entropy tomography (MENT)
- Algebraic reconstruction (ART, SART)



# Example: Quadrupole Scan @ AWA

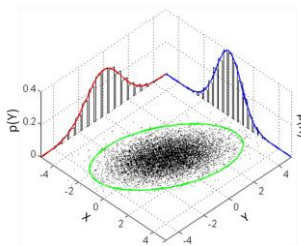


## Drive beamline





# Using Optimization to Reconstruct Distributions

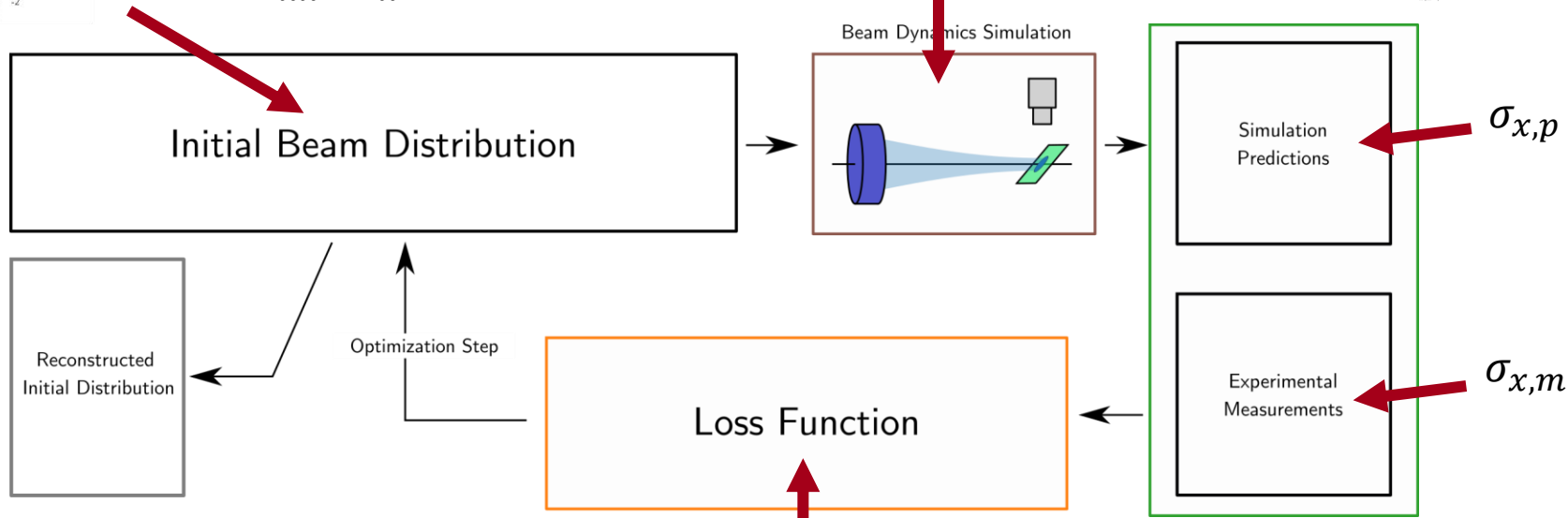
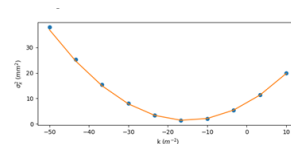


Second order beam moments

$$\sigma_x, \sigma_{xx'}, \sigma_{x'}$$

Analytical beam dynamics model

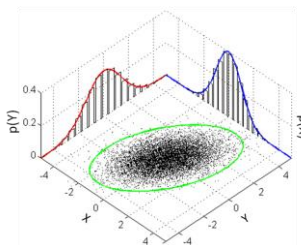
$$\sigma_{x,p}^2 = (1 + dlk)^2 \sigma_x^2 + 2(1 + dlk) \sigma_{xx'} + d^2 \sigma_{x'}^2$$



$$l = \sum (\sigma_{x,p} - \sigma_{x,meas})^2$$



# Using Optimization to Reconstruct Distributions

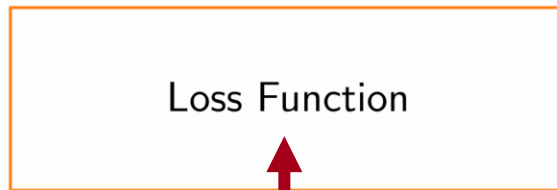
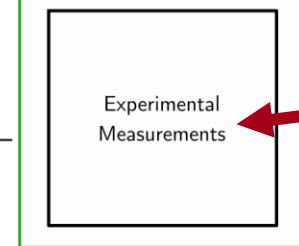
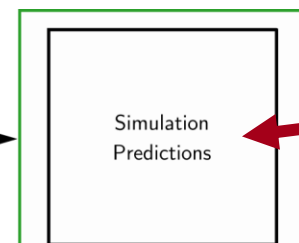
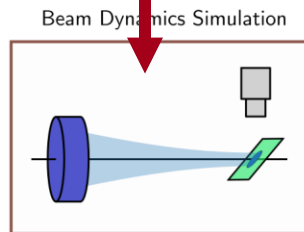
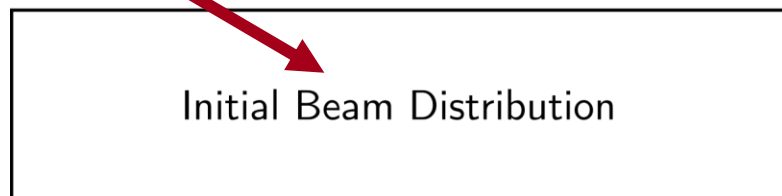
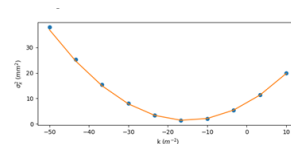


Second order beam moments

$$\sigma_x, \sigma_{xx'}, \sigma_{x'}$$

Analytical beam dynamics model

$$\sigma_{x,p}^2 = (1 + dlk)^2 \sigma_x^2 + 2(1 + dlk) \sigma_{xx'} + d^2 \sigma_{x'}^2$$



$$l = \sum (\sigma_{x,p} - \sigma_{x,meas})^2$$

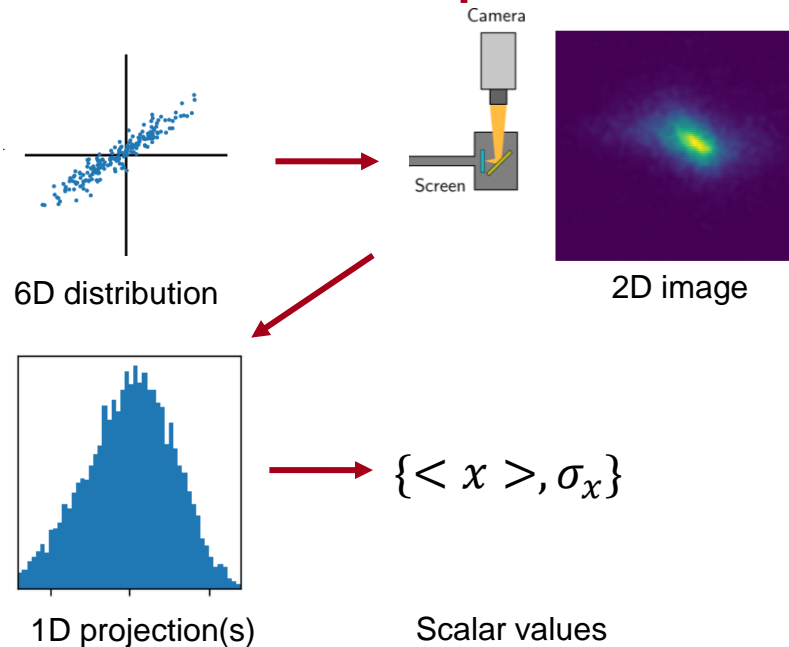


Optimization Step

Conventional analysis of measurements loses **A LOT** of beam information

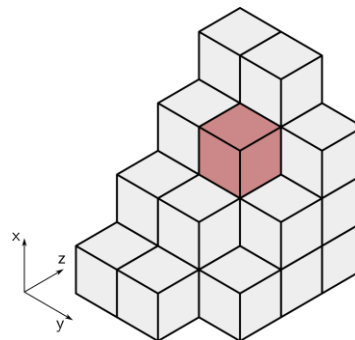
# Why Is This So Difficult?

## Information compression



Often required by analysis constraints (analytical tractability, optimization simplicity, etc.)

## Costs of detailed beam representations



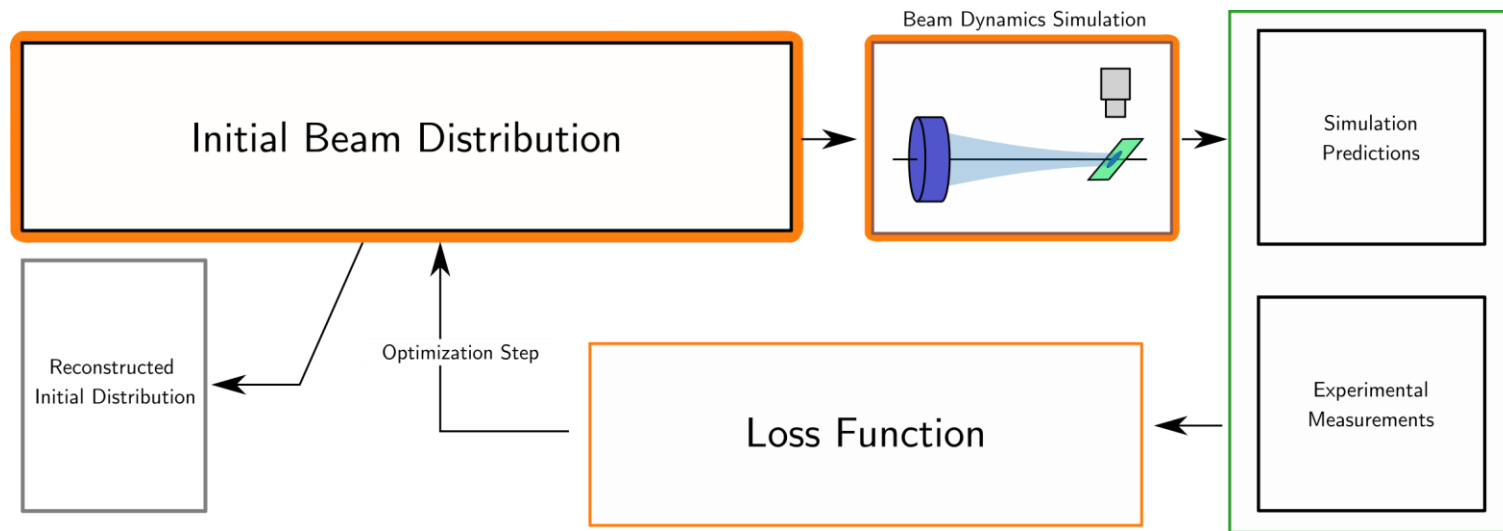
Histogramming scales poorly with number of dimensions,  $N_{bins} \propto n^D$

Reasonable resolution,  $n = 100$   
For a 6D distribution,  $N_{bins} = 10^{12}!$

# The Road to Efficient 6D Reconstructions

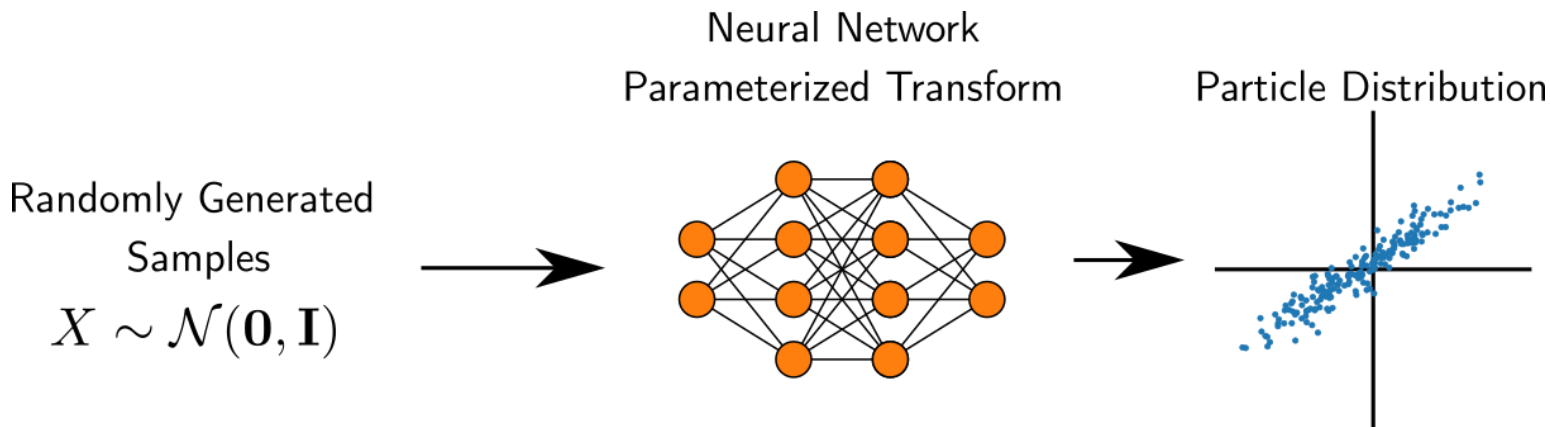
(1) Use **generative machine learning** to represent complex 6D beam distributions.

(2) Implement **differentiable** beam dynamics simulations to enable learning



# Machine Learning Based Beam Representations

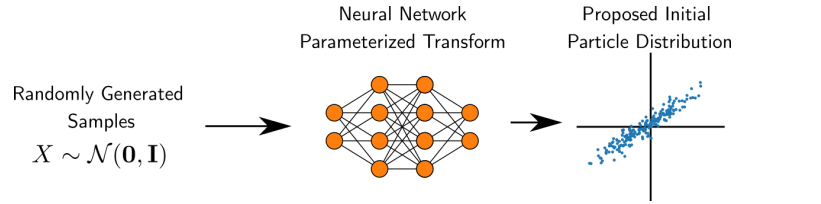
Use a **generative** machine learning model to create **arbitrary beam distributions**



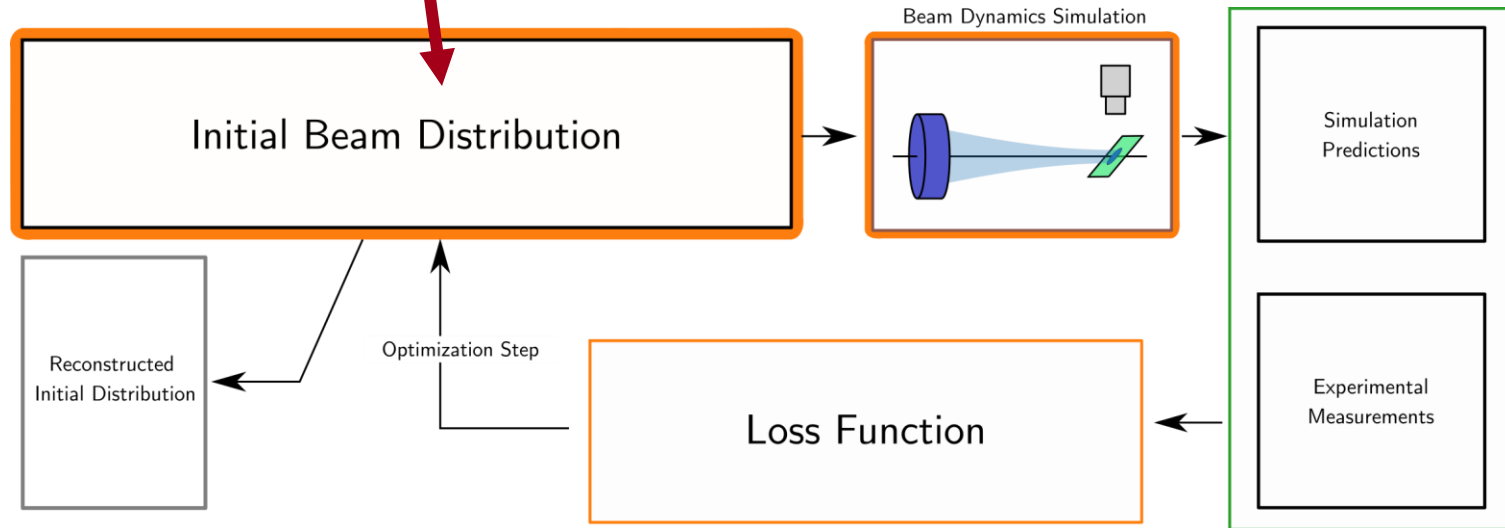
$O(\sim 1000)$  parameters of the neural network control the distribution of particles in 6D phase space

Learn the NN parameters  $\rightarrow$  learn the beam distribution

# The Road to Efficient 6D Reconstructions

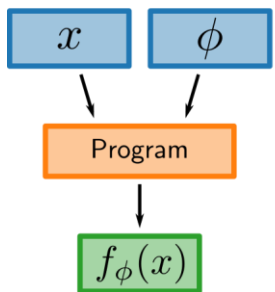


(2) Implement **differentiable** beam dynamics simulations to enable learning

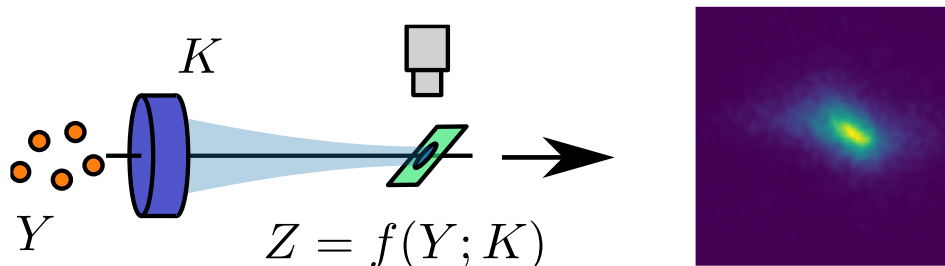
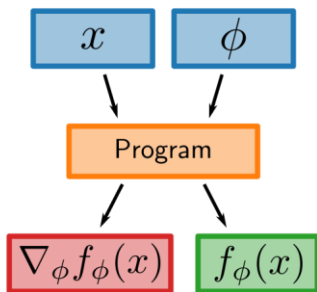


# Differentiable Beam Physics Simulations

Standard Programming



Differentiable Programming



**Differentiable sims** keep track of derivative information during **every** calculation step.

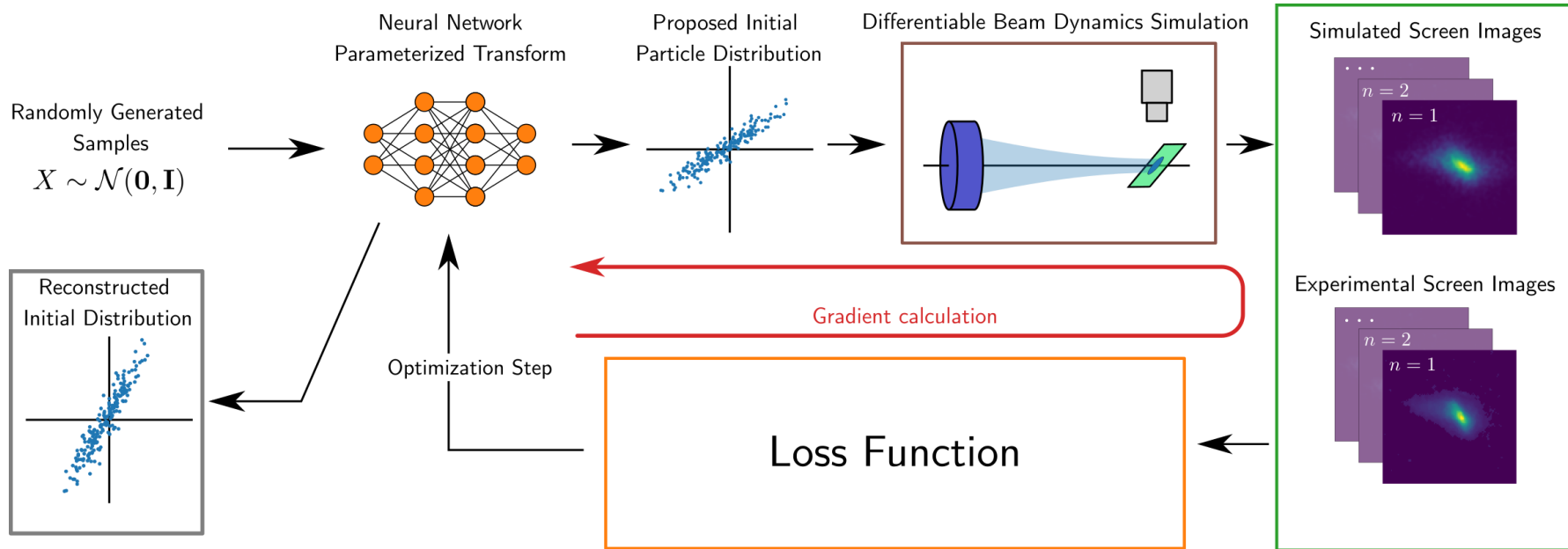
Enables **cheap gradient evaluations** which enable optimization of >10k free parameters.

$$Q^{(i,j)} = \text{KDE}(Z)$$

$$\frac{\partial Z}{\partial Y}, \frac{\partial Z}{\partial K}, \frac{\partial \sigma_Z}{\partial K}, \dots, \frac{\partial Q^{(i,j)}}{\partial Y}, \frac{\partial Q^{(i,j)}}{\partial K}$$

Allows us to extract information from the **individual pixels** of an image.

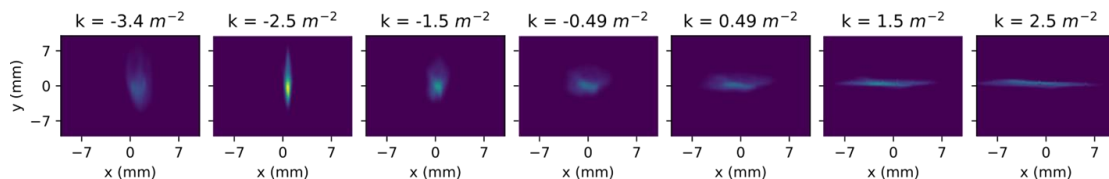
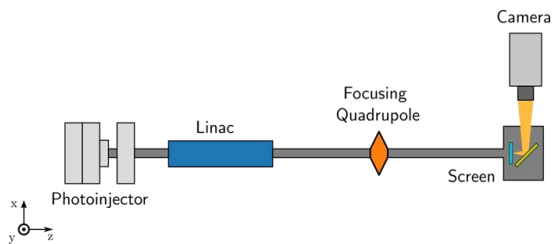
# Generative Phase Space Reconstruction (GPSR)



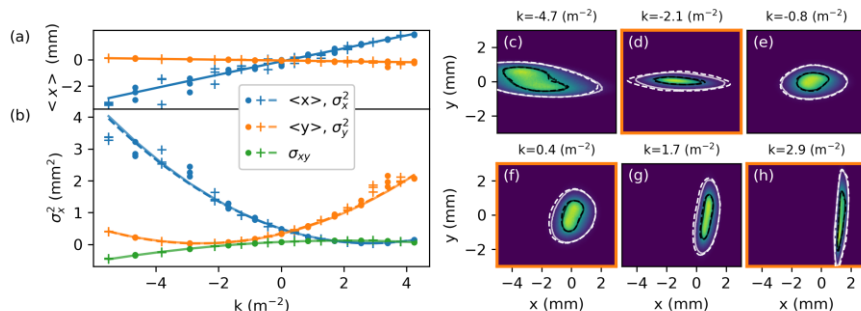
Now we can use **gradient descent** to optimize (train) beam parameters  $\rightarrow$  **necessary** for learning  $O(\sim 1000)$  parameters



# Four-dimensional reconstructions using GPSR

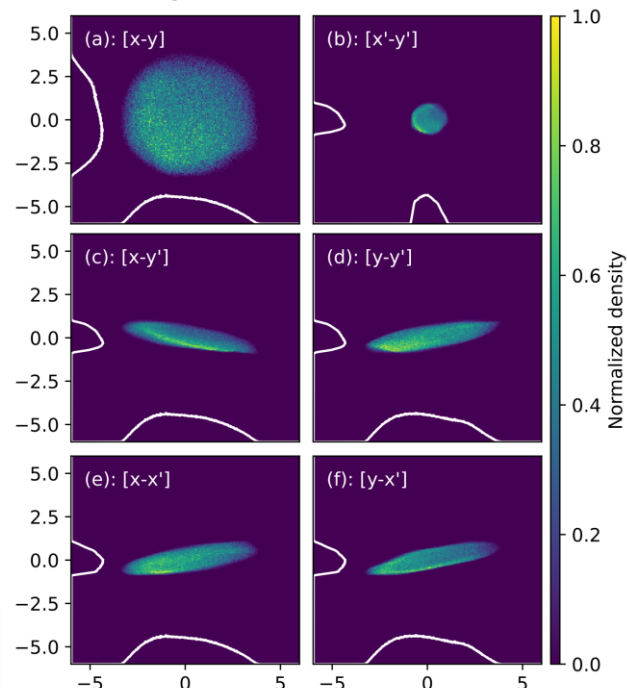


Proof of concept reconstruction from quadrupole scan



Roussel, R, et al. PRL (2023)

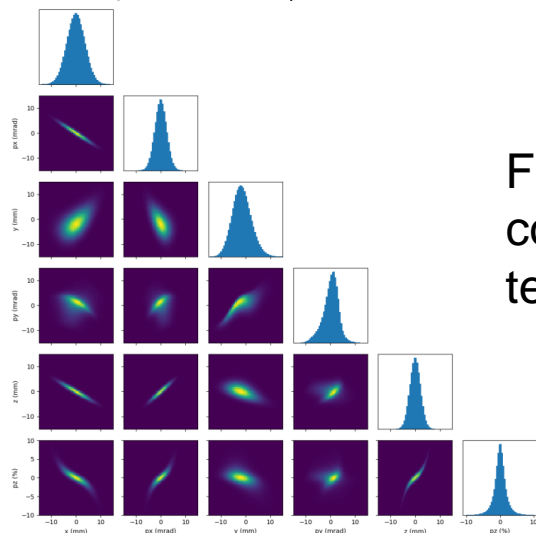
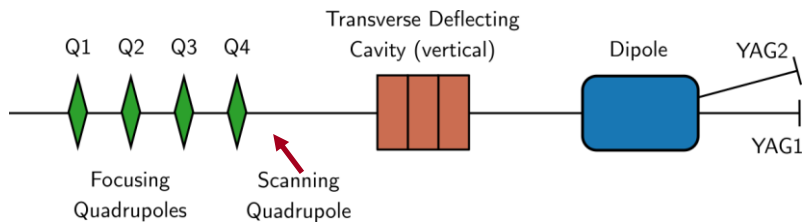
## Application to magnetized / flat beams



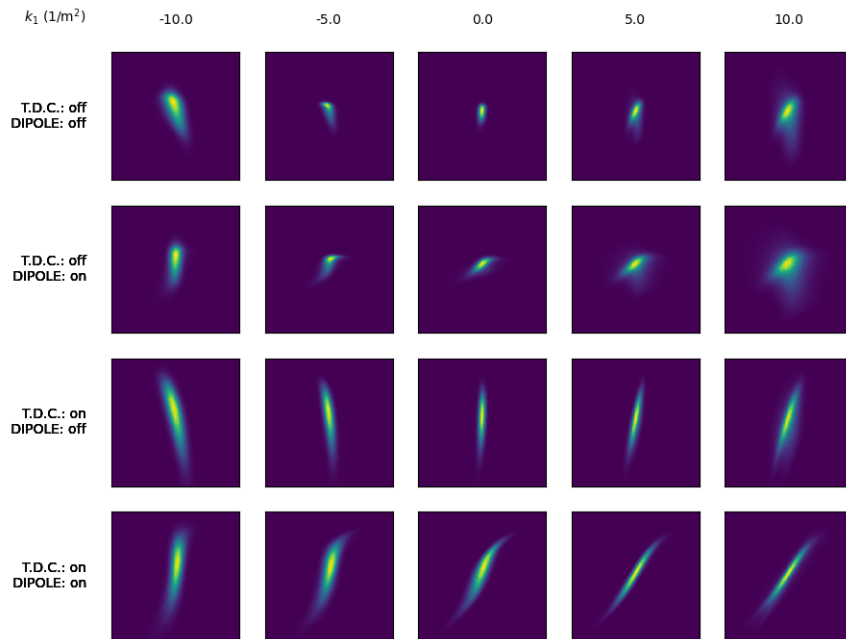
Kim, S., et al. PRAB (2024)

# 6D Phase Space Reconstruction (Simulation)

Logical next step, combine 4D reconstructions from quad scans with single shot longitudinal phase space diagnostics

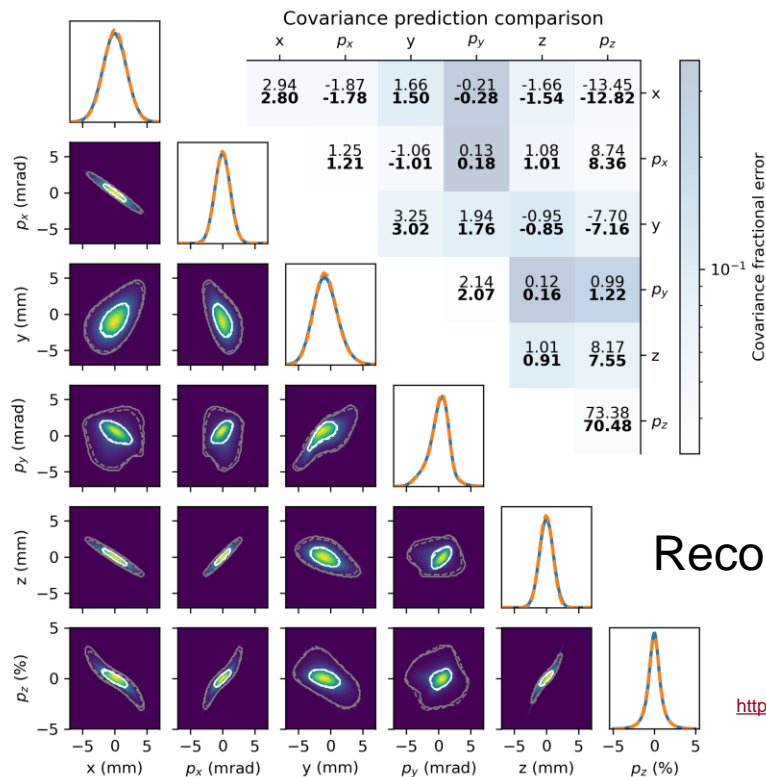
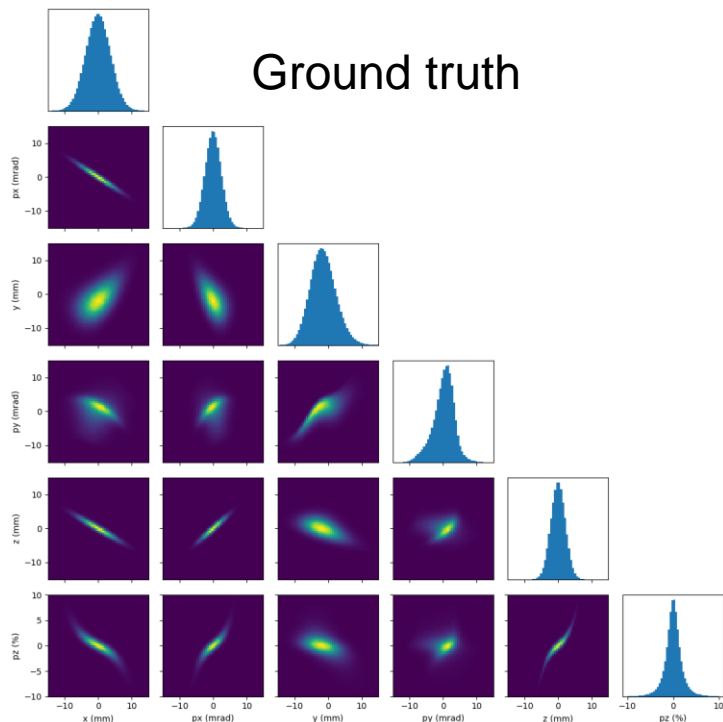


Fully correlated test beam



20 simulated measurements

# Reconstruction Results (Simulation)

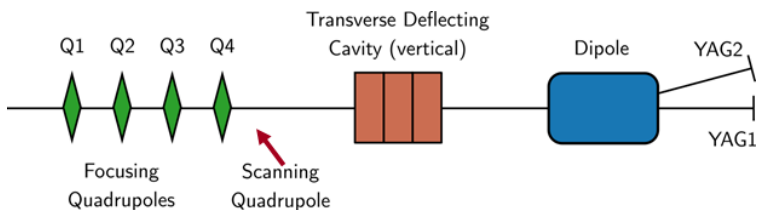


### Reconstruction

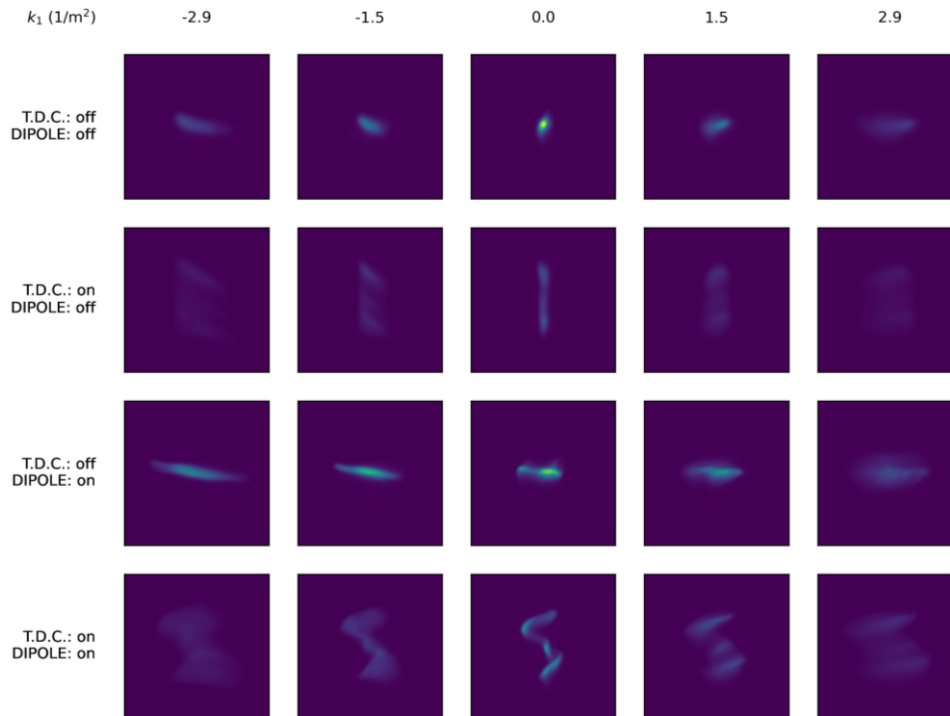
Roussel, R, et al.  
<https://arxiv.org/abs/2404.10853>  
 (Submitted to PRAB)

Successful reconstruction with 20 measurements, ~8 min computation time

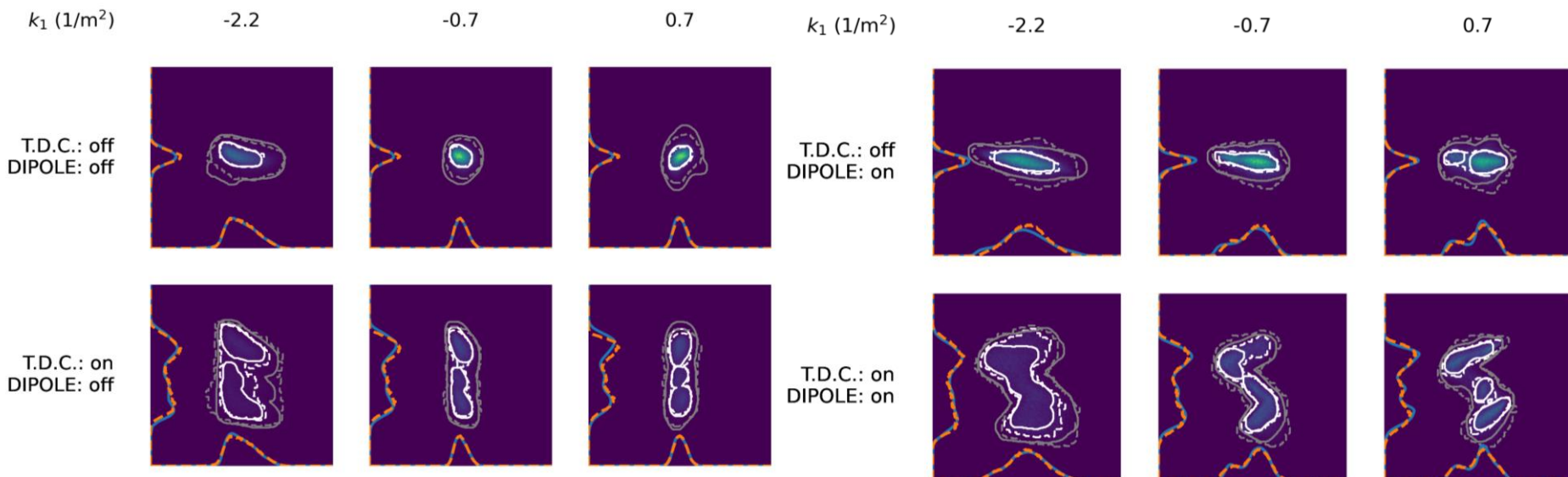
# Experimental Results from AWA



## Measurement data

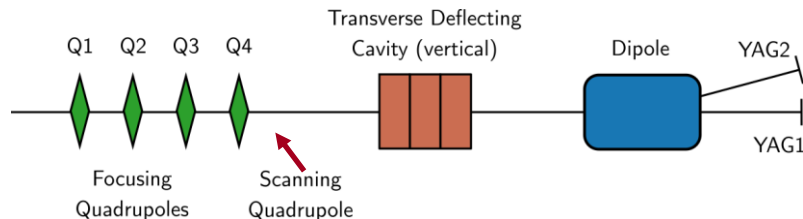


## Reconstruction predictions on test data



— Measured      - - Predicted

# Experimental Results from AWA



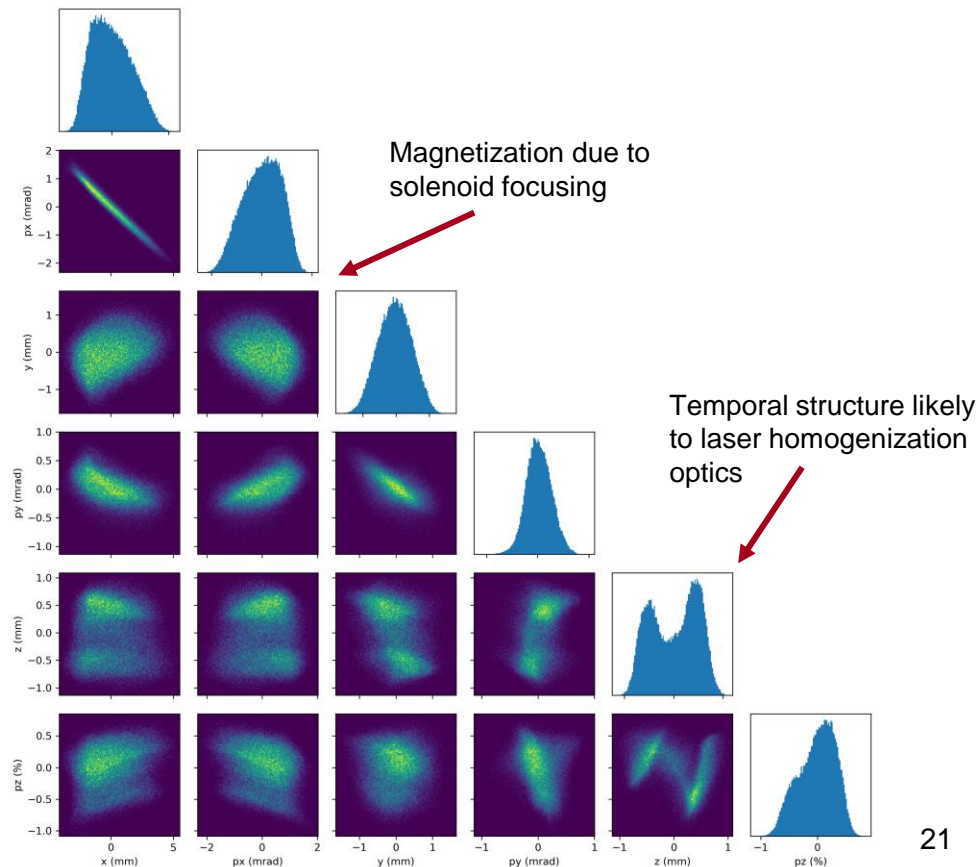
Reconstructed 6D distribution

Measurement: ~8 mins

Training: ~17 mins

Roughly **75x faster** than previous demonstrations

Roussel, R, et al. <https://arxiv.org/abs/2404.10853>  
(Submitted to PRAB)



- **Generative modeling + differentiable physics simulations (GPSR)** → substantial improvements in analysis capabilities
- Experimental demonstration at AWA → reconstructions + analysis **in ~ 17 mins** → can be further improved
- **Enables better understanding of beam distributions in experiment** → better wakefield accelerators, beams for colliders/cooling, better understanding of beam physics (CSR)



# Questions?

## SLAC

- Auralee Edelen
- Chris Mayes
- Daniel Ratner

## U. Chicago

- Juan Pablo Gonzalez-Aguilera

## Argonne Wakefield Accelerator

- Seongyeol Kim
- John Power
- Eric Wisniewski
- Wanming Liu
- Alexander Ody

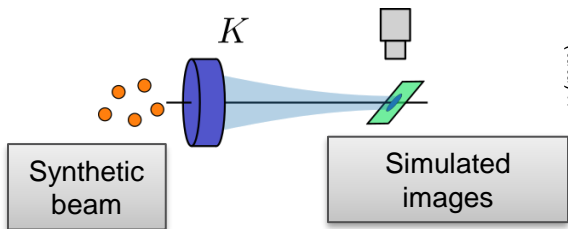
**Thanks to the team!**

# Backup

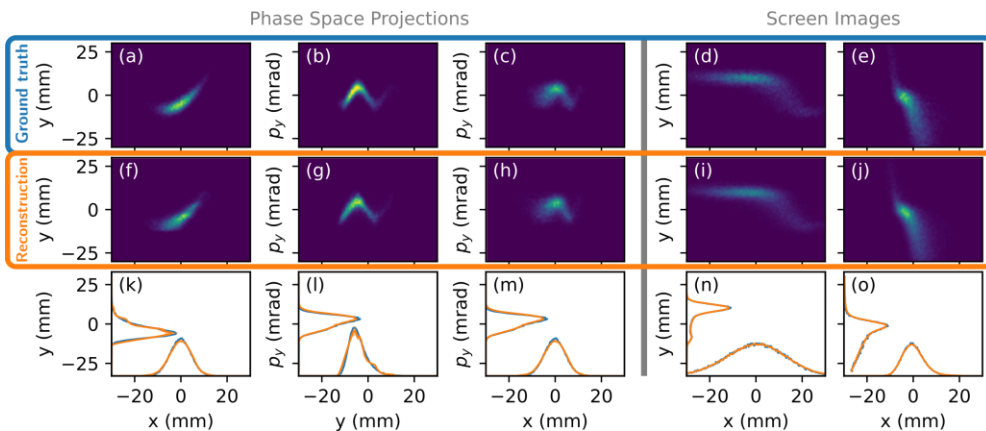
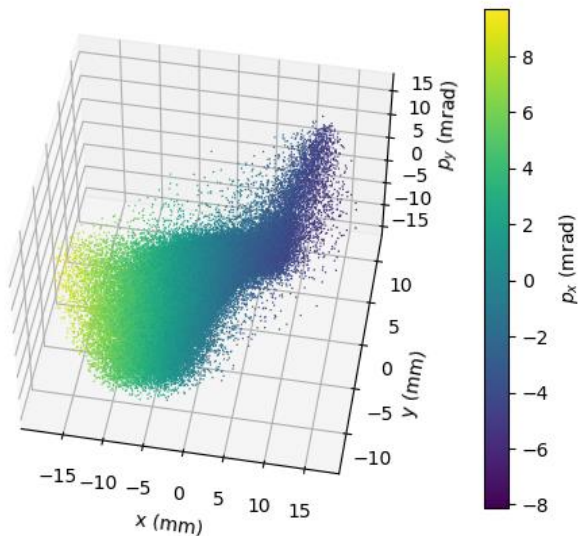
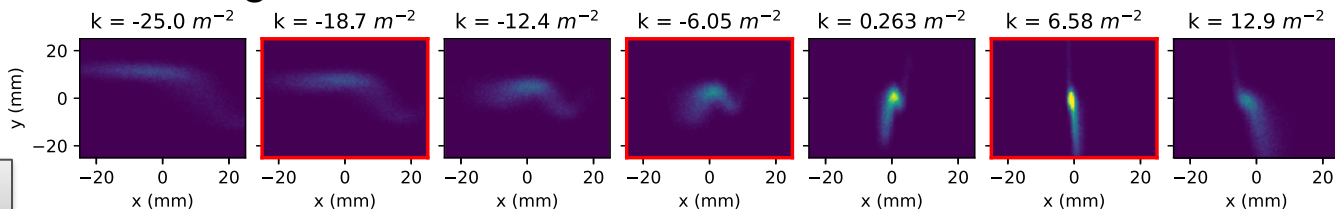
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# GPSR Synthetic Example

## Beamline

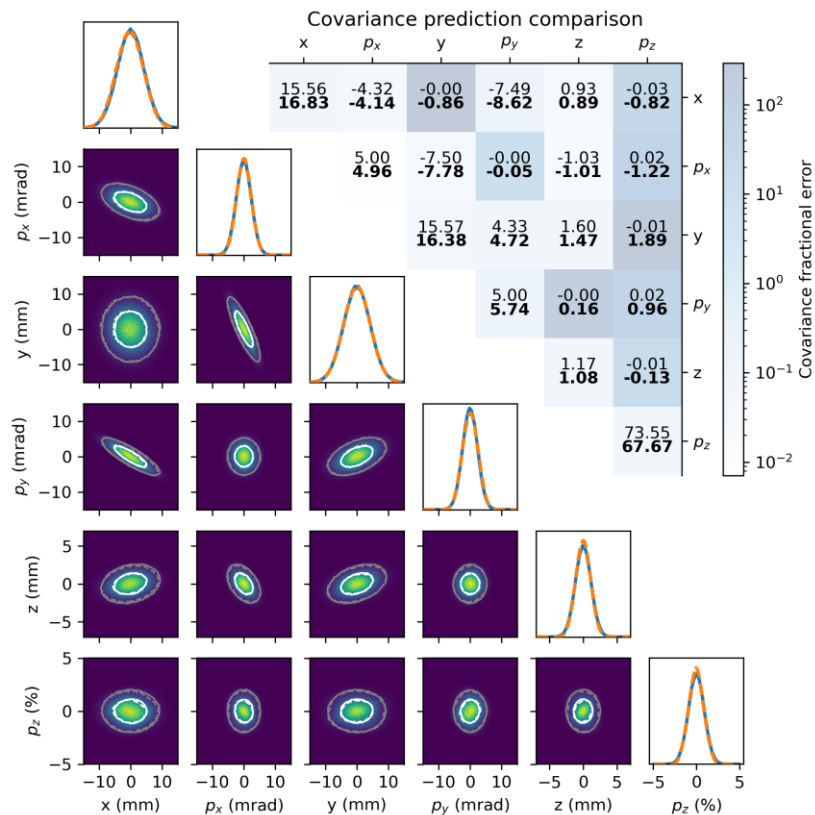


## Screen images

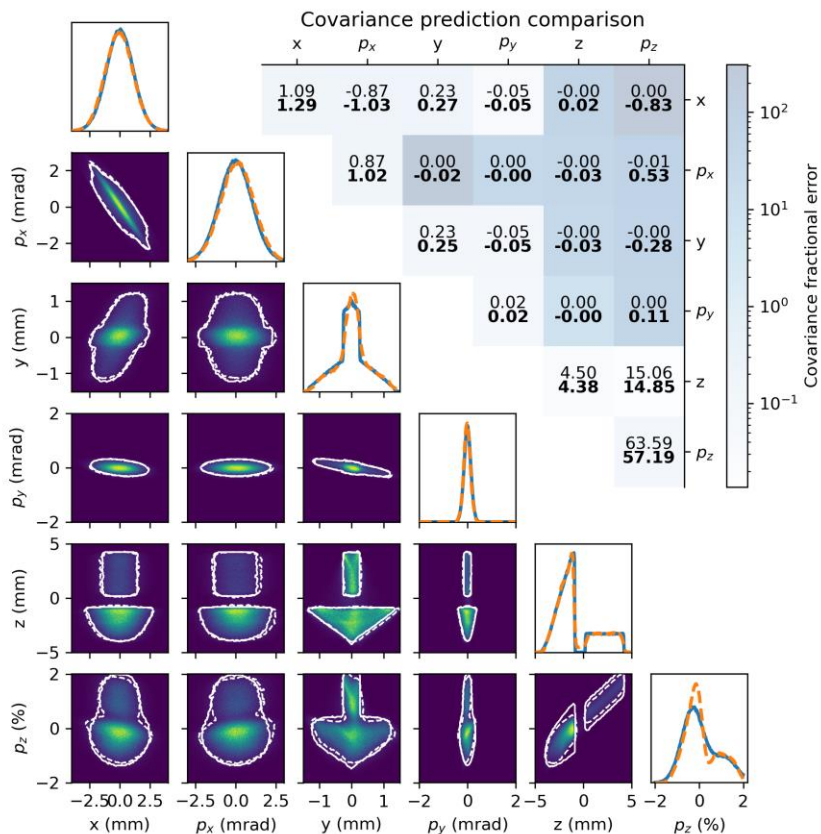


Accurate **4D reconstructions** using **10 measurements**  
Reconstruction (training) time **< 5 mins**  
**No prior training data needed**

# Reconstruction Results (Gaussian Beam)

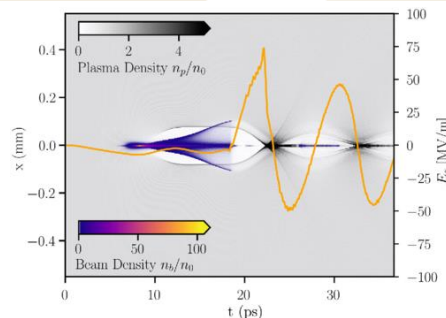


# Reconstruction Results (EEX Beam)



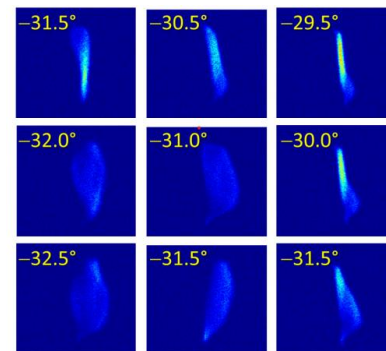
# Needs From the Accelerator Community

- There is a need for **fast, sample efficient** methods to reconstruct **detailed 6D beam distributions**
- This would enable:
  - **Practical optimization of complex beam distributions** for wakefield acceleration, magnetized beams, collective effect mitigation, free electron accelerators
  - **Improvements in understanding complex physical phenomena** (plasma wakes, coherent synchrotron radiation (CSR) on beam distributions)



High transformer ratio wakefield acceleration using triangular beams

Roussel, R, et al. PRL 124.4 (2020): 044802.



CSR measurements @ LCLS

Emma, P., et al. FEL2012 (2012).