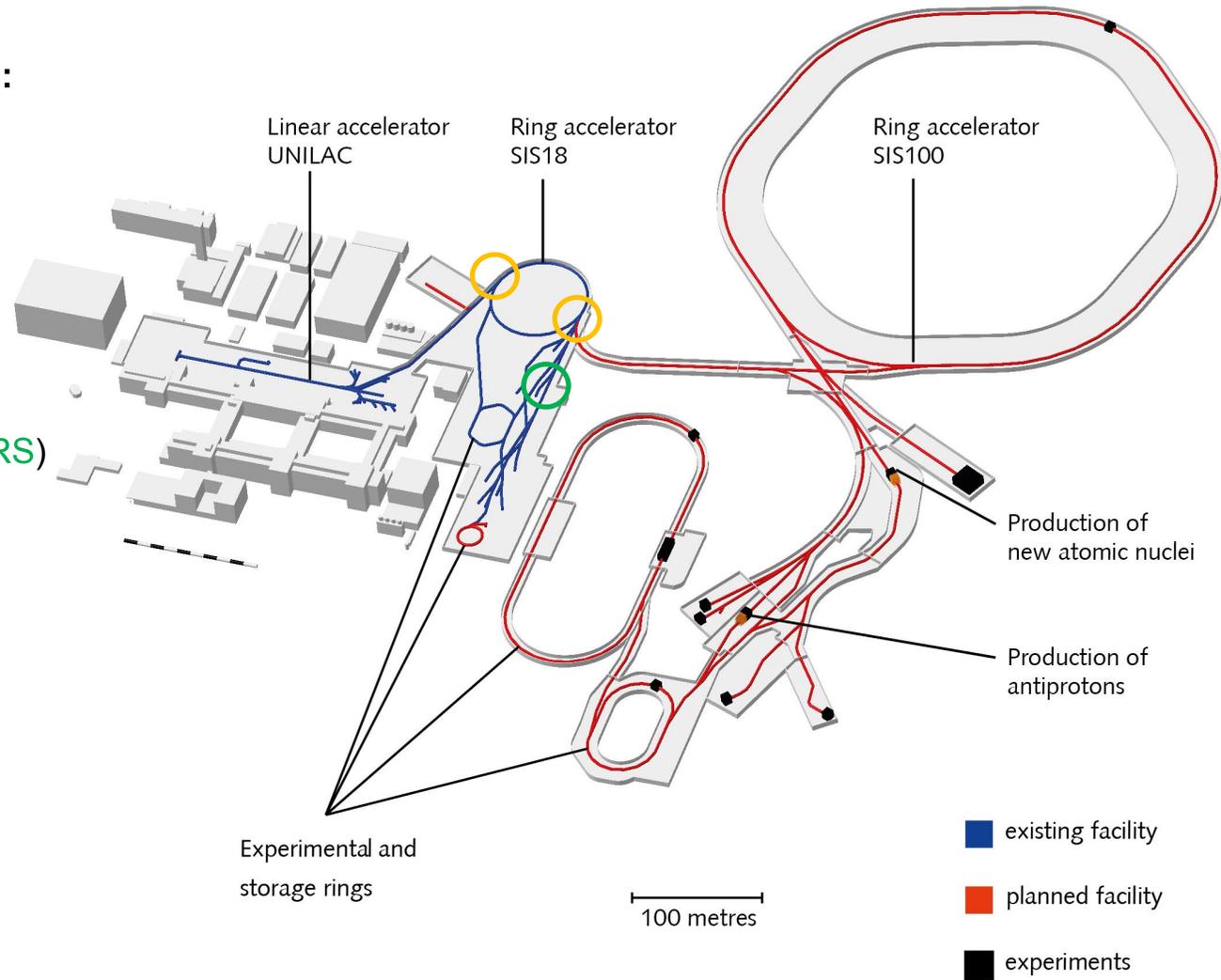
A detailed wireframe 3D model of an accelerator facility. The central feature is a large, horizontally-oriented oval ring structure, likely representing a synchrotron or storage ring. In the background, there are several smaller, more complex structures, including what appears to be a building with a grid-like facade and other smaller rings or components, all rendered in a similar wireframe style.

# Optimization at the GSI/FAIR accelerator facility

S. Appel

## Optimization with numerical optimizers:

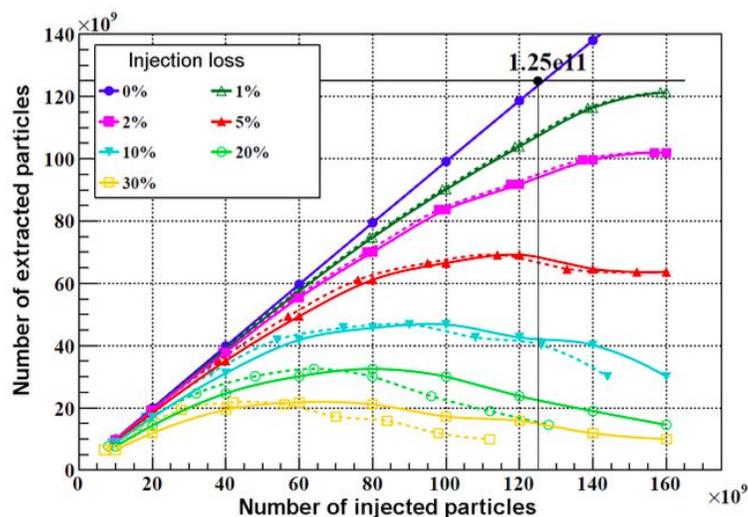
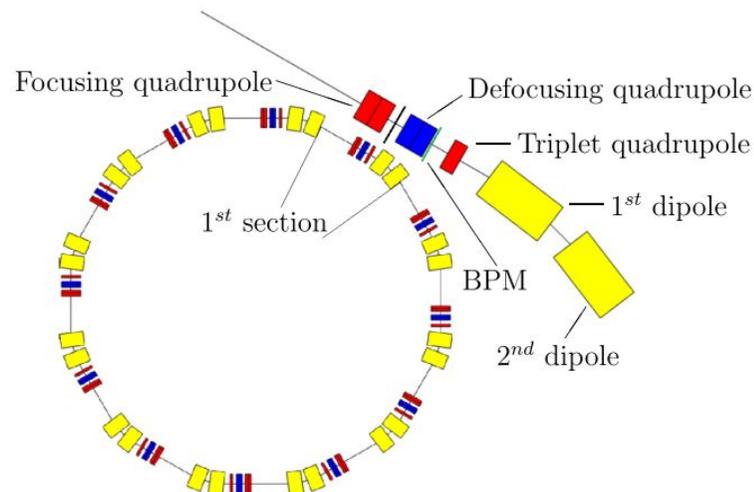
- Multi-turn injection (SIS18)
- Closed orbit correction for challenging optics (SIS18)
- Slow extraction (SIS18)
- Beam steering to production target (FRS)
- ...



# FAIR construction site: October 2023



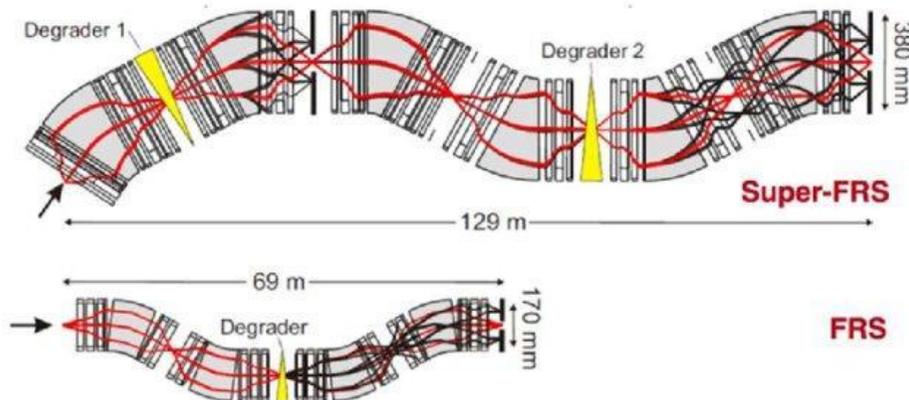
- Booster for FAIR SIS100 Synchrotron.
- Multi-turn injection into SIS18 is one bottleneck to reach intense beams for FAIR.
- The (incoherent) transverse space charge force is the main intensity limiting effect in the FAIR synchrotrons.
- For intermediate charge state ions, the loss-induced vacuum degradation is another important key intensity-limiting factor.



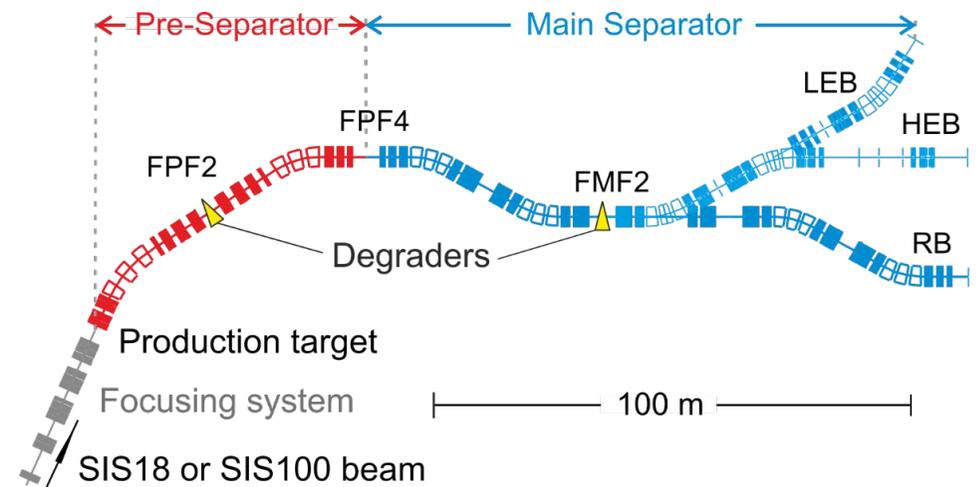
Beam Energy	U : 50-1000 MeV/u Ne: 50-2000 MeV/u p : 4,5 GeV
Particle Number Per Cycle	$10^{10}$ (U <sup>73</sup> ) – $10^{12}$ (p)
Typical Cycle	3 s (FAIR 2.7 Hz)
Extraction	fast: 1 $\mu$ s with compression: 200 ns slow: 10-8000 ms

# FRS (FRagment Separator) and Super-FRS

- Production and investigation of nuclear structure of exotic nuclei.
- characteristics of the high-resolution magnetic spectrometer FRS, exotic nuclei can be produced, separated, identified and eventually stored in a storage ring
- Super-FRS increases acceptance and complexity (about 4 times more magnets), Gain factors of 1000 ( $^{12}\text{C}$ ) and 7500 ( $^{132}\text{Sn}$ ) can be reached<sup>1</sup>.
- Automation of time-consuming tasks will therefore be essential



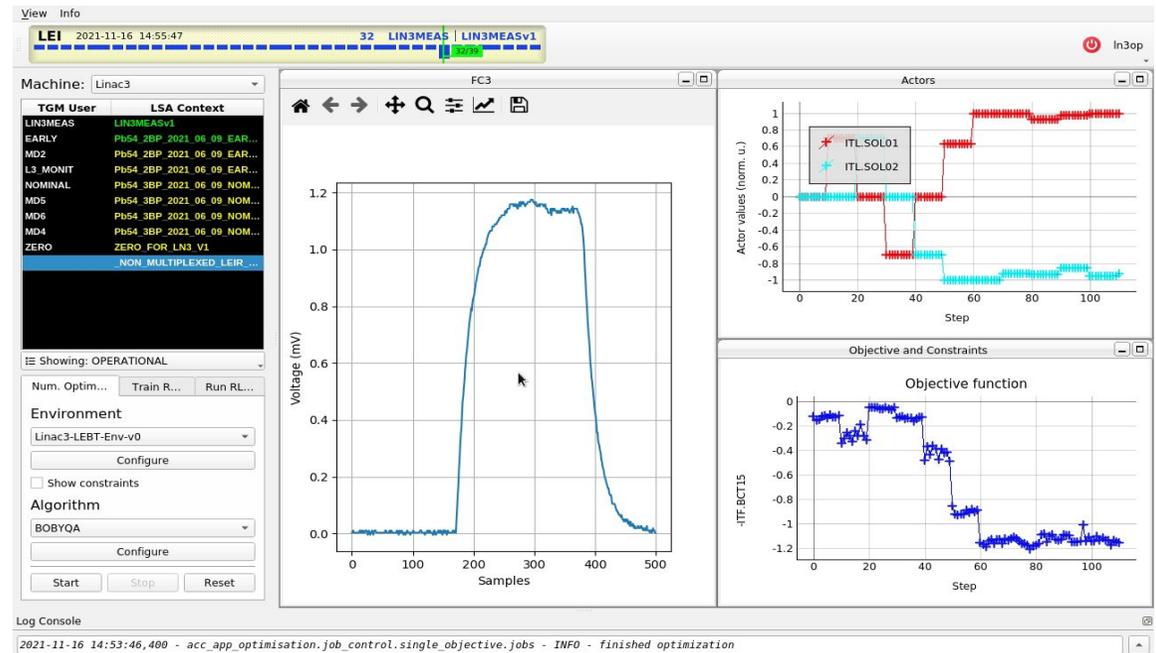
## Super-FRS layout:



<sup>1</sup>H. Simon et al., EMIS XVIII workshop, 2018.

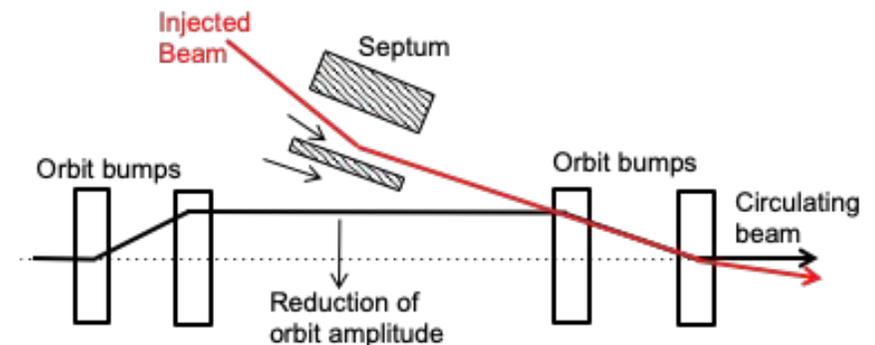
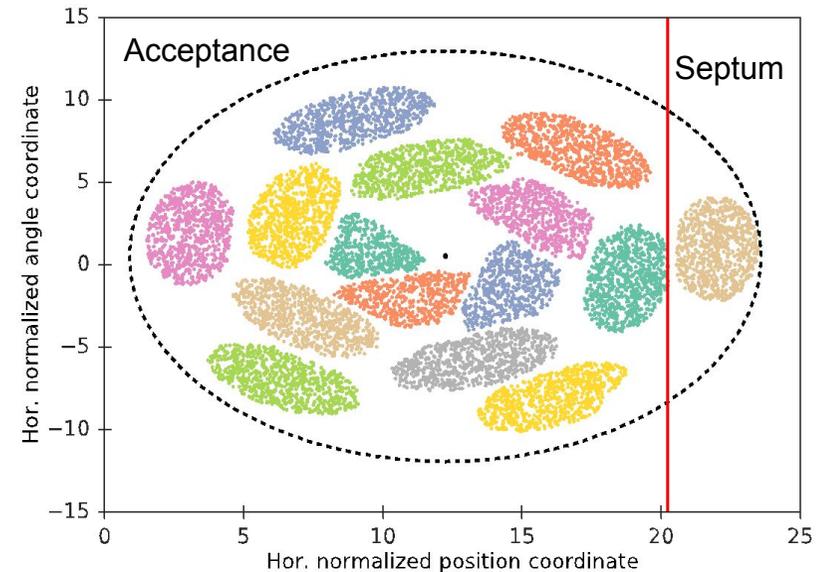
**Generic Optimization Frontend & Framework (GeOFF)** is a widely used framework for testing automation at CERN

- Python-based framework with use of `pylsa`
- lists, configures and runs optimization problems
- standardized interfaces and adapters for various packages via Common Optimization Interfaces
- Optimization problem formulated in standard interfaces
- Class contains logic for live plotting, data logging, and communication with LSA, FESA and the Device Access system



- Adaption of code quickly and on-the-fly during shift: **Flexibility** of framework made this **easy**

- MTI has to respect Liouville's theorem: Injected beams only in free space
- Gain factor should be as high as possible to reach the space charge limit
- Injection loss should be as low as possible to avoid loss-induced vacuum degradation
- Objective: Minimize beam loss for given number of injection with five parameters
- MTI model has been implemented in Xsuite<sup>1</sup> for fast tracking
- Xsuite support CPUs and GPUs
- Testing RL optimization with SCL + MPC, see poster

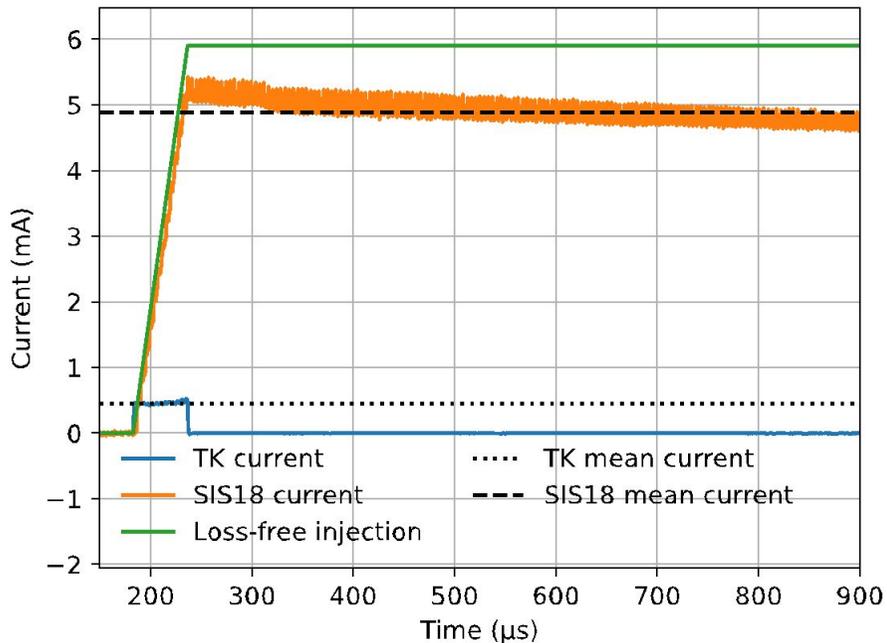


<sup>1</sup><https://xsuite.readthedocs.io/en/latest/>

# Automation of Multi-Turn Injection

- Usually, MTI is optimized manually with varying success
- Idea: Automated optimization of MTI
  - with numerical optimizer (Py-BOBYQA)
  - by using GeOFF and Python bridge
  - Goal: low loss injection over many turns

injection variables to be modified by numerical optimizer



Multiturn Injection	
Bumper ramp down time	110 μs
Bumper amplitude	9610443115234 mm
Unilac Offset	100 μs
Chopper delay	50 μs
Chopper window	60.0 μs
Chopper correction angle	0.0 mrad
GTK7MU5 correction angle	0.0 mrad
GS12MU3I correction angle	-0.07552774331 mrad
I-Septum correction angle	-0.44575636275 mrad

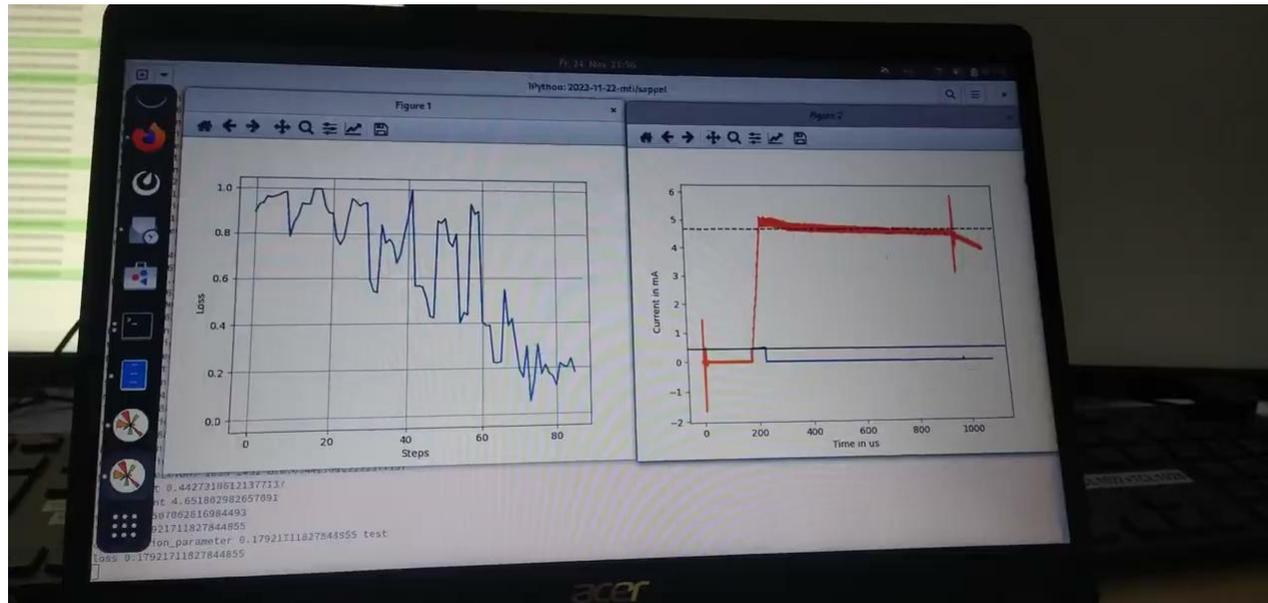
## Loss definition:

- $Loss = (ideal\ current - SIS18\ mean\ current) \div ideal\ current$
- $ideal\ current = mean\ TK\ current \times expected\ number\ of\ injections$
- $Expected\ number\ of\ injections = Chopper\ window \div SIS18\ rel.\ time$

Sabrina Appel (GSI), Nico Madysa (GSI)

# Automation of Multi-Turn Injection

- With the Python bridge fully automated multi-turn injection optimization can be performed in about 15–20 min.
- The loss could be reduced from 40 % to 15 % using five optimization parameter.



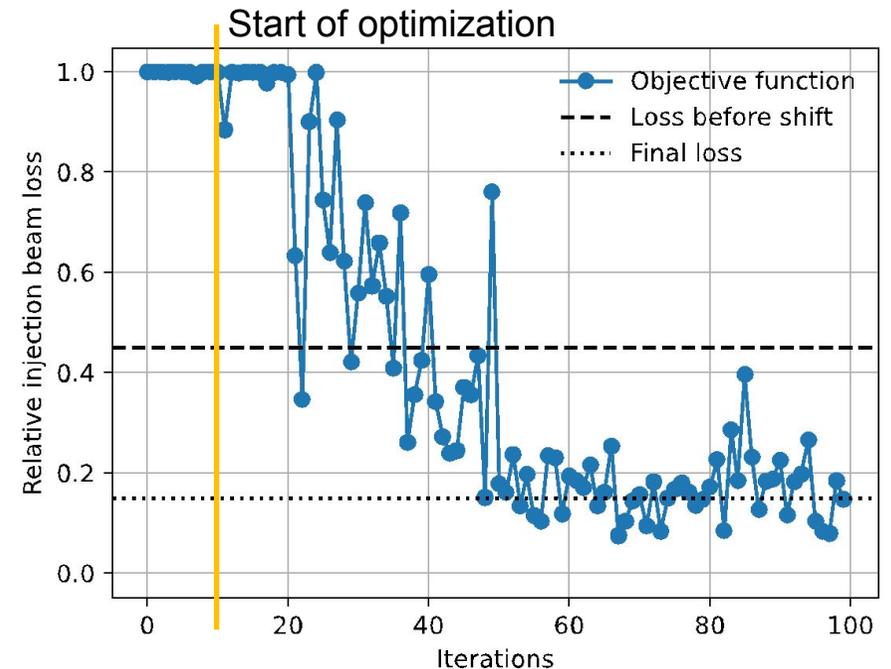
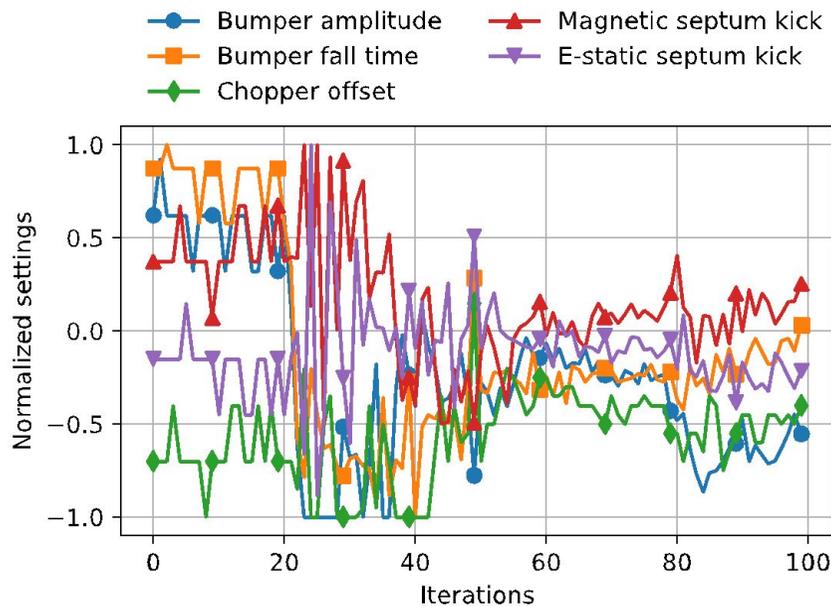
Usage of GeOFF in a scripting context, with the terminal window in the background, and continuously updating figures for monitoring in the foreground.

Sabrina Appel (GSI), Nico Madysa (GSI)

# Automation of Multi-Turn Injection

- Using five optimization parameters.
- It ran for **100** iterations, which took about **19 minutes**.
- Loss could be reduced from 40 % to 15 %.**
- Optimum was found, even though beam was fully lost in the beginning.

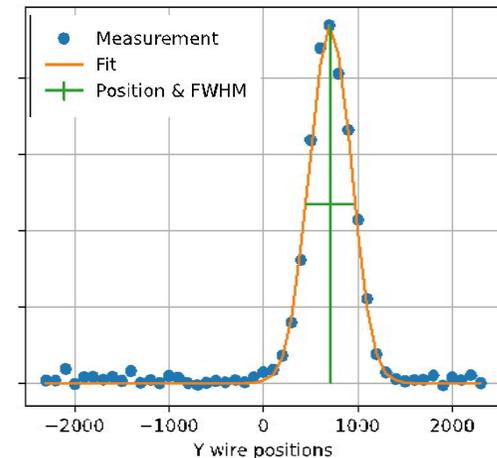
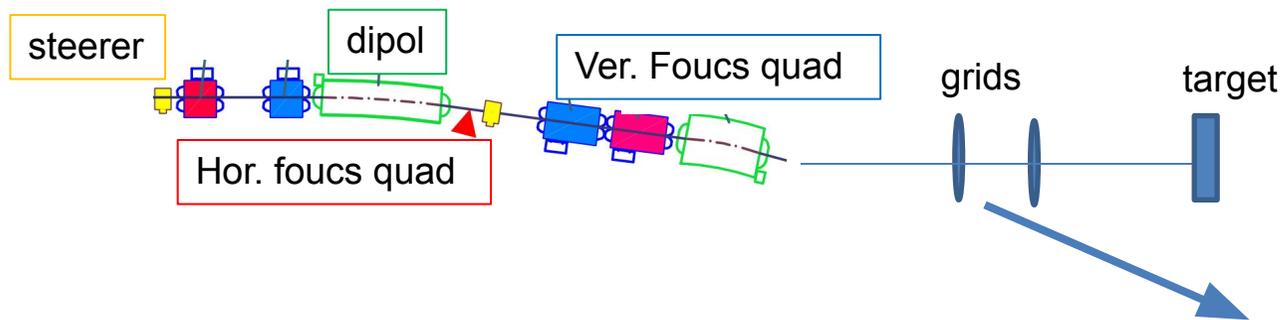
Multiturn Injection			
Bumper ramp down time	1	<input type="text" value="110"/>	$\mu\text{s}$
Bumper amplitude	2	<input type="text" value="0.610443115234"/>	mm
Unilac Offset		<input type="text" value="100"/>	$\mu\text{s}$
Chopper delay	3	<input type="text" value="50"/>	$\mu\text{s}$
Chopper window		<input type="text" value="60.0"/>	$\mu\text{s}$
Chopper correction angle		<input type="text" value="0.0"/>	mrad
GTK7MU5 correction angle		<input type="text" value="0.0"/>	mrad
GS12MU3I correction angle	4	<input type="text" value="-0.07552774331"/>	mrad
I-Septum correction angle	5	<input type="text" value="-0.44575636275"/>	mrad



Sabrina Appel (GSI), Nico Madysa (GSI)

## Objective Functions:

Misplacement or distance to the beam target using two grids in front of the target



## Limitation/time condition:

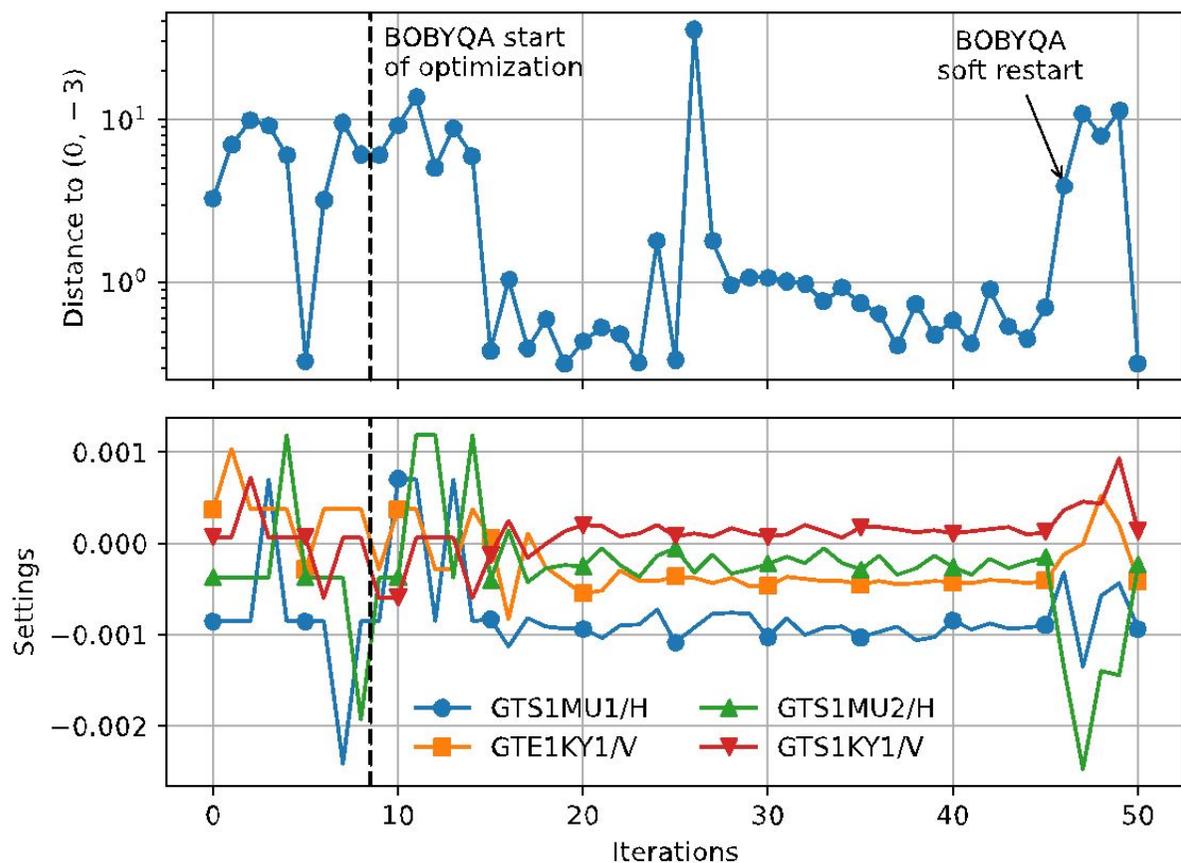
- Bandwidth multiplexer: multiple shots were necessary to acquire X/Y-grids
- “Bad shots”: analysis class cloud not perform reliable analysis of grids
- Slow extraction: SIS18 cycle length

S. Appel, E. Kazantseva, N. Madysa and S. Pietri (all GSI)

# FRS: Automatic online steering

## Proof of principle:

Online beam steering in **50 iteration** and took **18 minutes**.



S. Appel, E. Kazantseva, N. Madysa and S. Pietri (all GSI)

# Idea of Bayesian Optimization (BO) based closed orbit correction

## BO-based closed orbit correction:

Probability distribution (Gaussian Process) acts as a surrogate model

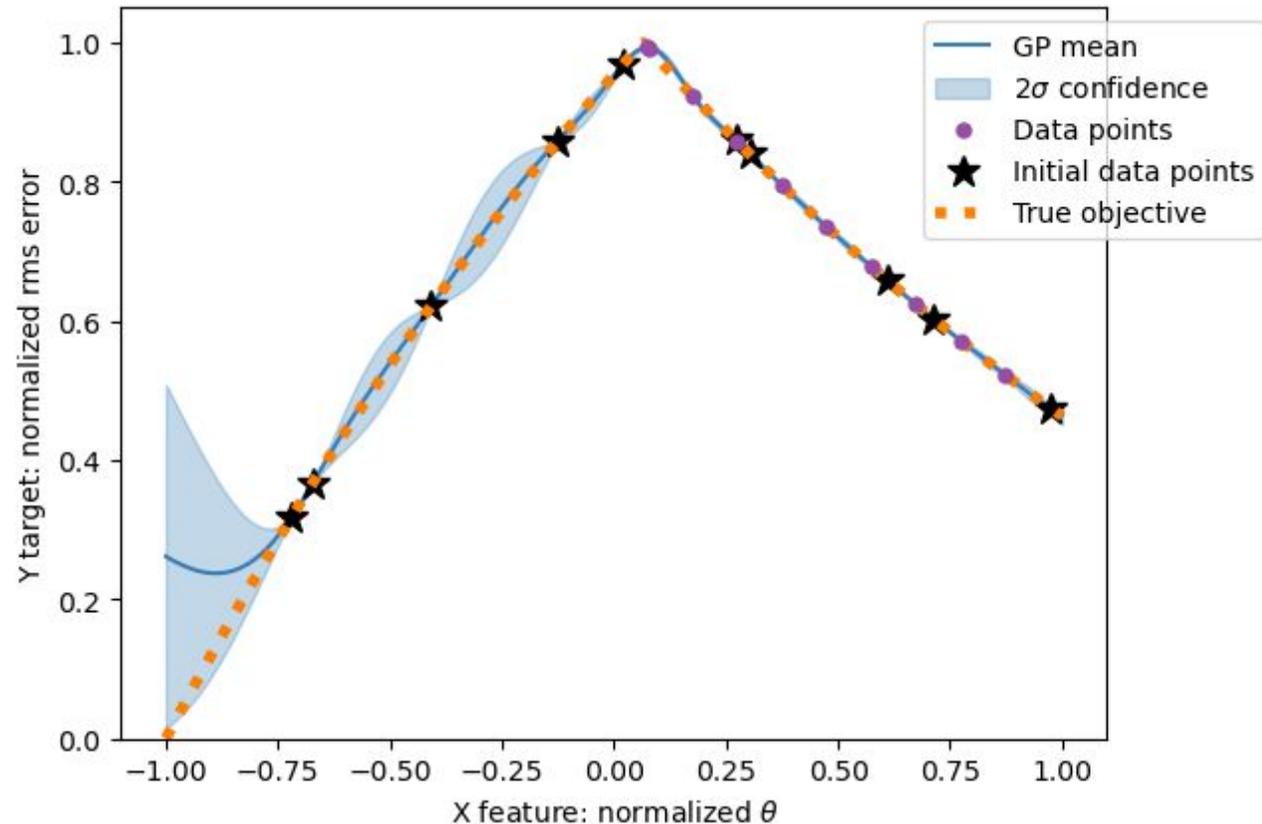
Described by mean function and the covariance function or kernel (currently using a Matern kernel, next step: physics-informed kernel)

Objective function: RMS deviation at BPMs

Next most promising evaluation point: Optimizing acquisition function (Upper Confidence Bound)

V. Isensee, MSc thesis 2023, Methods of Closed-Orbit Correction in Synchrotrons"

Simulation of the correction of a FODO cell (noise free):



V. Isensee (TU Darmstadt), C. Caliri (TU Darmstadt), A. Oeftiger (GSI)

# Challenging scenario at the experiment

## Challenges due to:

Nonlinearities introduced by the chromaticity correction with sextupoles

BPM Noise

Asymmetric optics:

Optical setting to shift the transition energy, called sigma optics

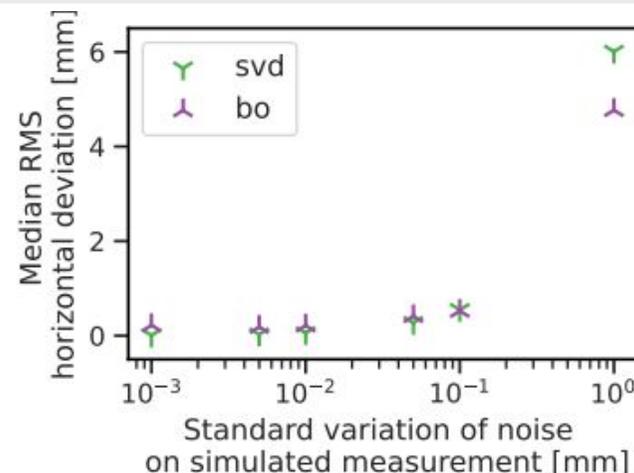
Superperiodicity reduced from S=12 to S=6

Can cause standard closed orbit correction methods to fail

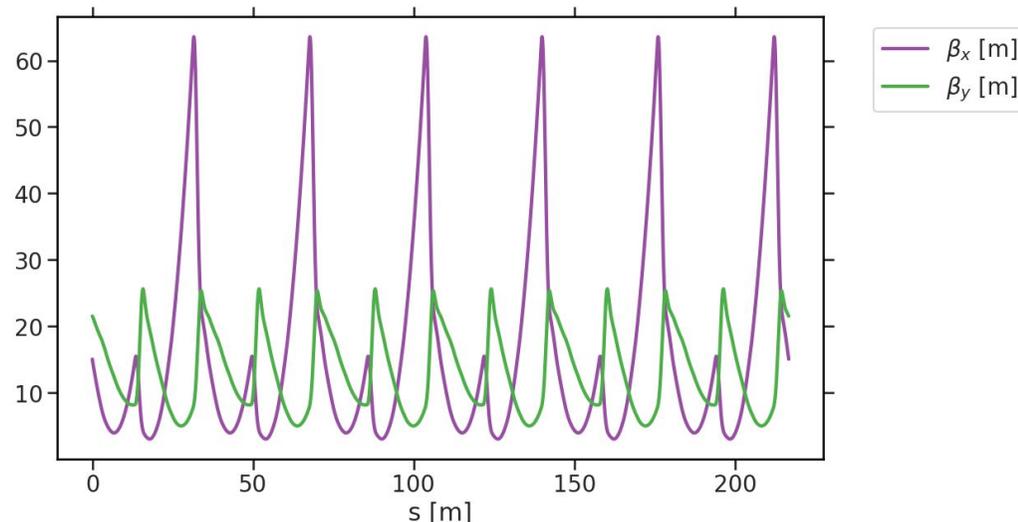
(see:

B. Franczak, K. Blasche and K. H. Reich, "A Variable Transition Energy Lattice for SIS 12/18," in IEEE Transactions on Nuclear Science, vol. 30, no. 4, pp. 2120-2122, Aug. 1983)

Simulated influence of Noise on BPMS:  
For higher noise level  
BO achieves better  
results than SVD



Simulated Betafunctions in sigma optics scenario:



V. Isensee (TU Darmstadt), C. Caliri (TU Darmstadt), A. Oeftiger (GSI)

# BO-based correction for challenging machine optics

## Results of BO-based correction:

Sigma optics with sigma value = 0.5

Random data needed for initialization

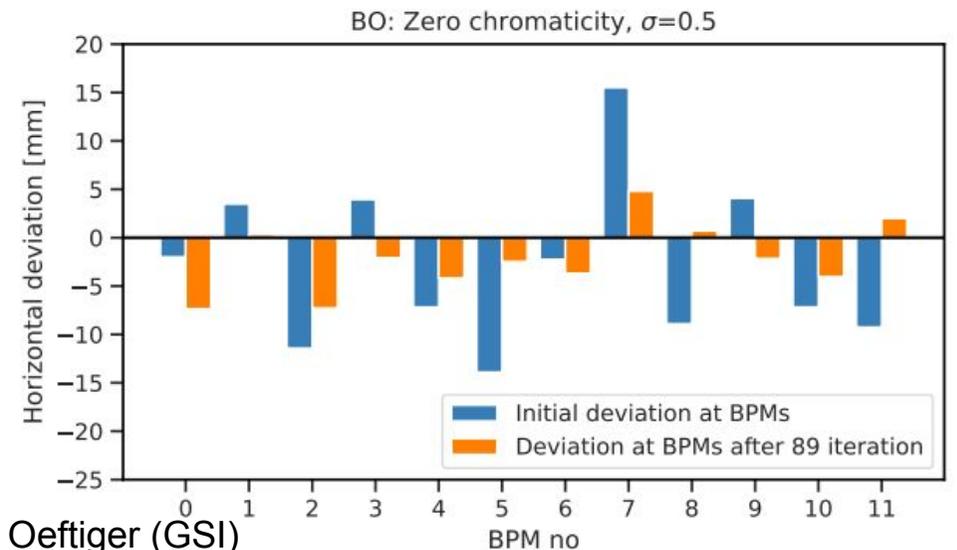
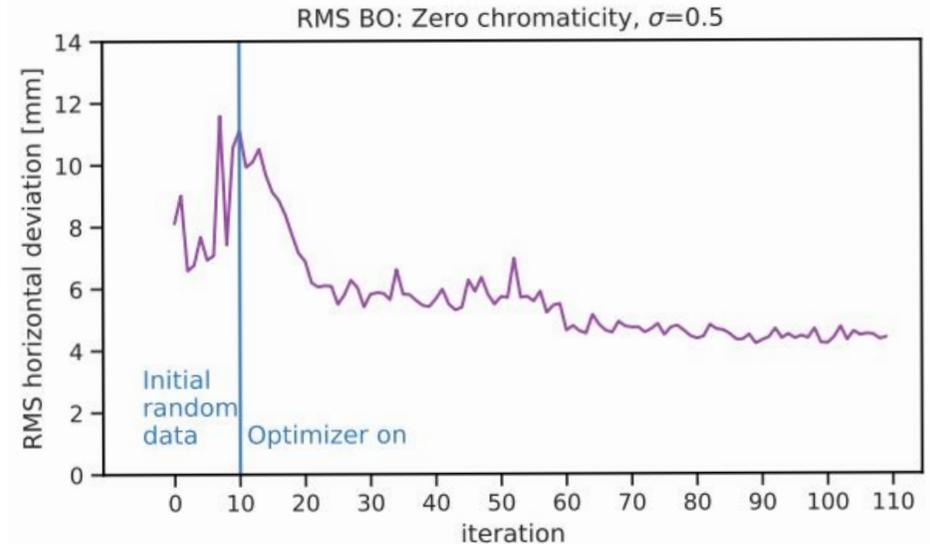
RMS of horizontal deviation converges

BO-based correction found result, where the standard methods (SVD) fails

## Best found closed orbit:

Trained surrogate model, including a hyperparameter for noise, which can be reused

→ Basis for physics-informed kernel: BO with orbit model in between BPMs



V. Isensee (TU Darmstadt), C. Caliari (TU Darmstadt), A. Oeftiger (GSI)

# BO-based correction for challenging machine optics

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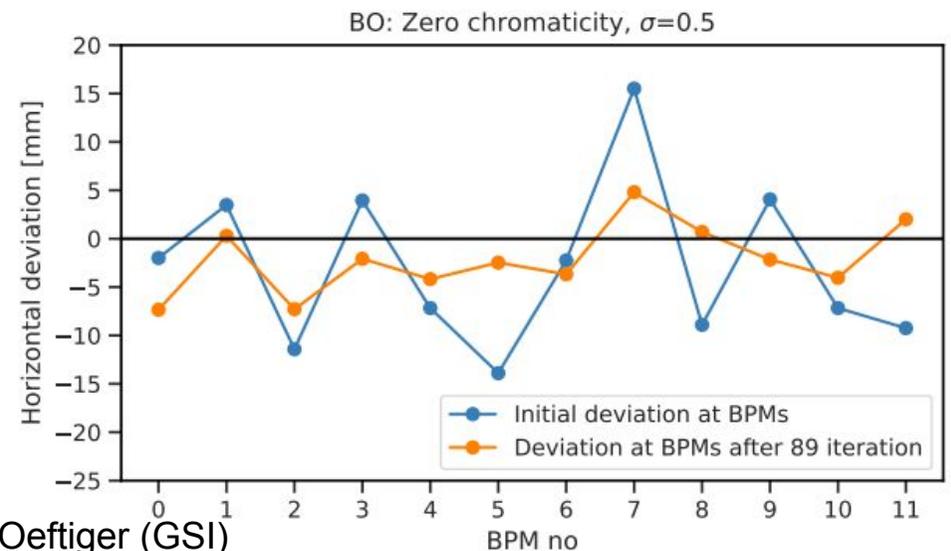
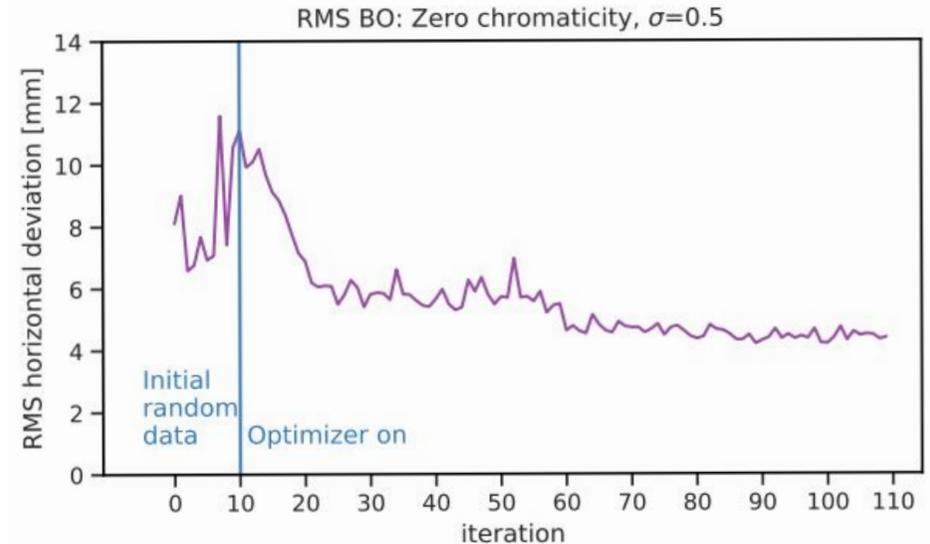
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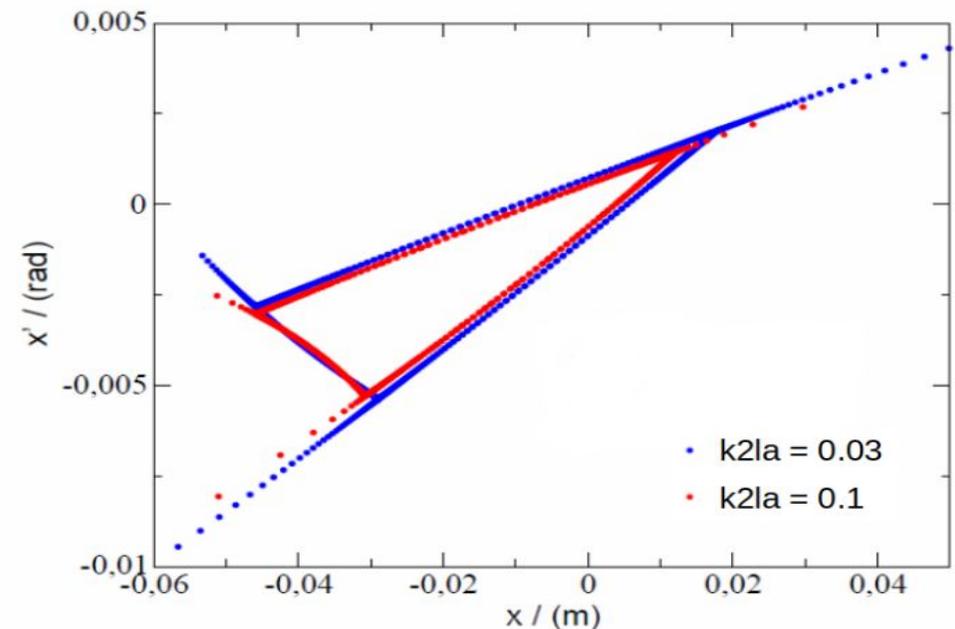


V. Isensee (TU Darmstadt), C. Caliri (TU Darmstadt), A. Oeftiger (GSI)

- Slow extraction (typ. 1 to 10 s) based on interplay of many parameters
- Very important for the fix target experiment (production of rare isotopes, compressed matter)

## Slow Extraction by Tune sweep:

- Excitation of 3rd integer resonance with sextupoles.
- Shrinkage of separatrix due to quadrupole ramp.
- Sextupole strengths are characterized by amplitude and phase.
- Large amplitude results in lower particle loss due to transit from beam extraction channel in large steps.
- Phase affects orientation of triangle, thus, determines particle loss.



Olha Kazinova, TU Darmstadt  
Stefan Sorge, GSI

## Maximizing extraction efficiency

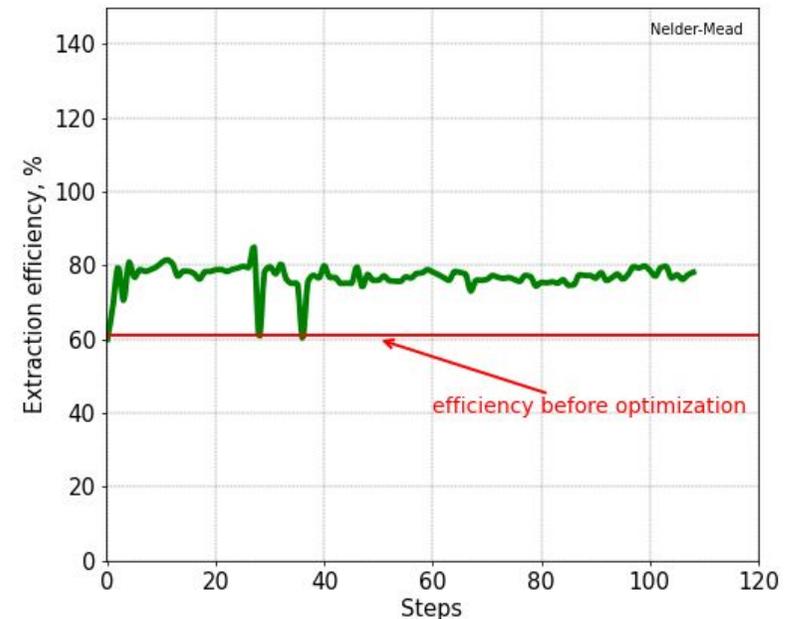
$$p_{\text{eff}} = \frac{N_{\text{ex}}}{N_{\text{circ}}}$$

Extracted particles  
Particles in circulating beam

Variables used for optimization:

sextupole amplitude  $(k_2L)_a$  and phase  $\phi$  which define strengths of the resonances sextupoles by

$$(k_2L)_{n_{\text{period}}} = (k_2L)_a \sin \left( 2\pi \frac{n_{\text{period}} - 1}{N_{\text{period}}} + \phi_{\text{sx}} \right)$$



Extraction efficiency could be increase from **from 60 % to 80 %**.

Olha Kazinova, TU Darmstadt  
Stefan Sorge, GSI

- Thank you for your attention.
- Questions?