

# Machine learning-based surrogate model construction for optics matching at European XFEL injector

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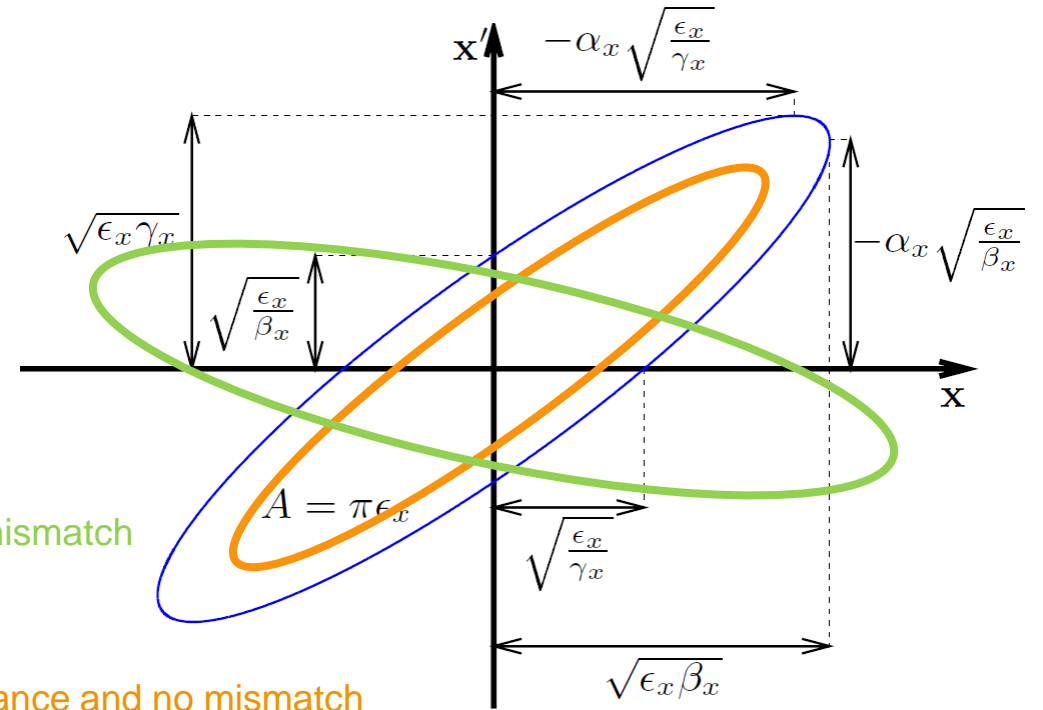
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# Emittance and twiss parameters

- Emittance is the area in phase space covered by particle distribution
- Ellipse is introduced to describe the distribution
- Emittance is a conserved quantity if beam only subjected to conservative forces (Liouville's Theorem)
- The beam orientation can be described using the Twiss parameters  $\alpha, \beta, \gamma$
- Mismatch: beam distribution does not have the same orientation as the design optics

same emittance with mismatch

smaller emittance and no mismatch



# Motivations

Optics matching in the injector section is essential to the downstream beam transport and final FEL lasing performance.

Linear optics model

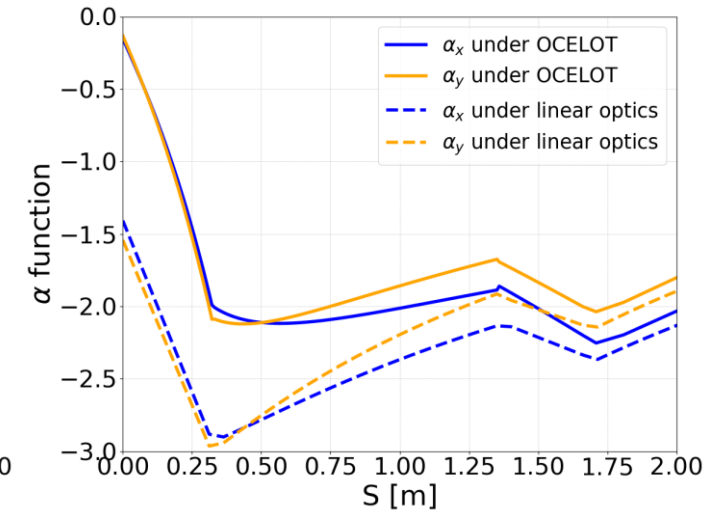
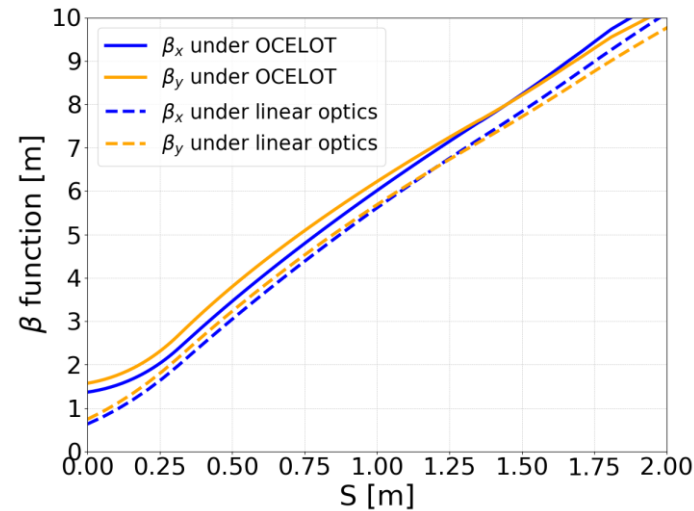
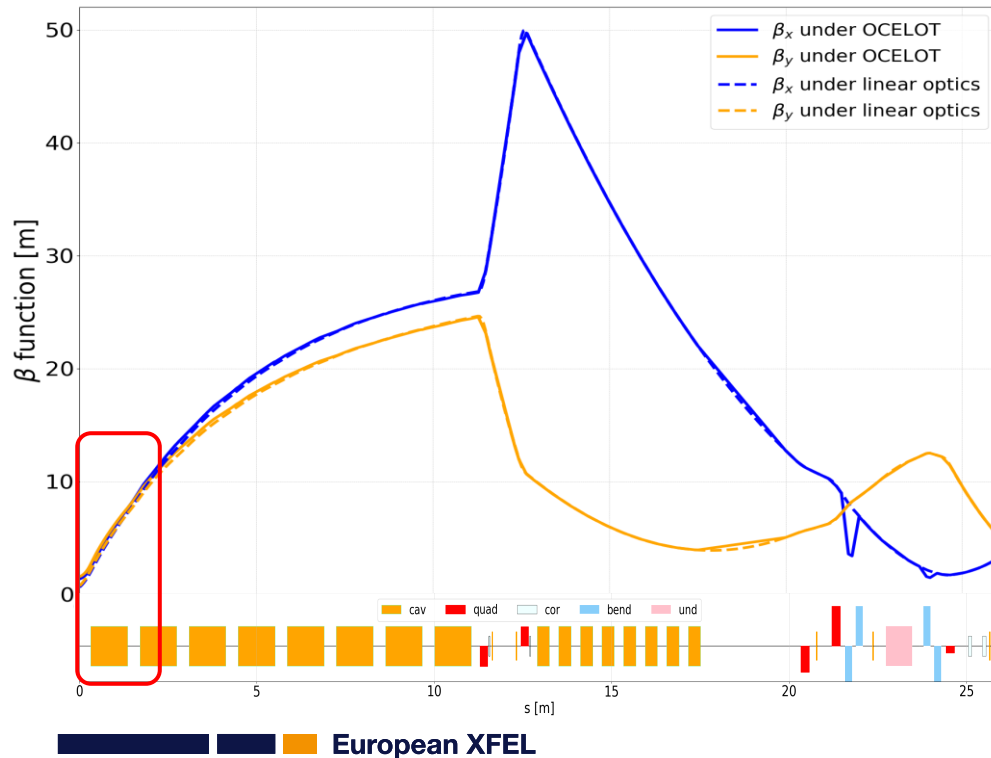
No access to get beam optics properties at gun cavity exit

Samples from OCELOT simulation

Start from 6 MeV

Collective effects (space charge, wakefields)

Second-order optics

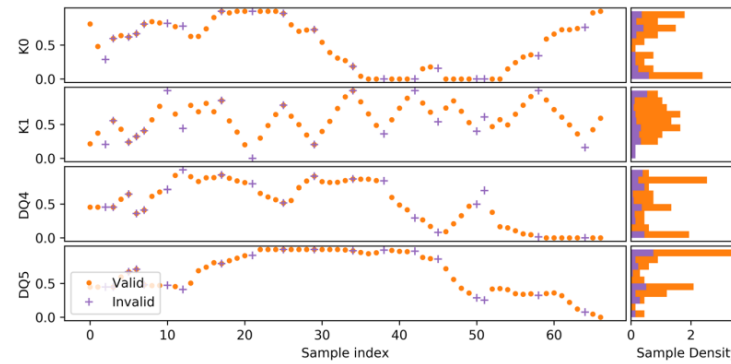


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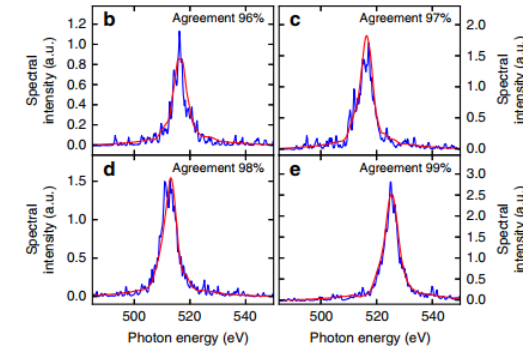


# Motivations

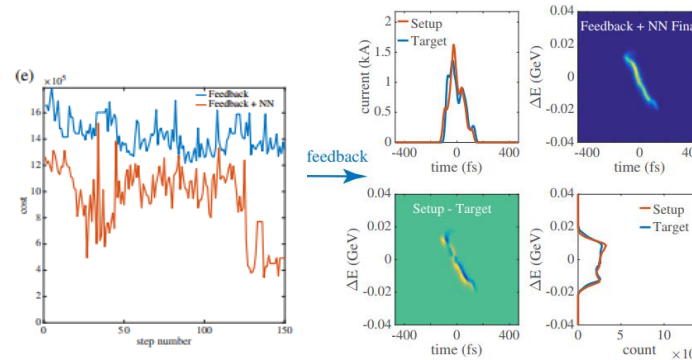
- Machine learning techniques are mature enough to be applied to accelerators
- Anomaly detection and machine protection
- System modeling**
- Virtual diagnostics
- Online tuning and control
- Advanced data analysis
- Introducing the ML-based technique to construct a surrogate model
- an alternative to the existing linear optics toolkit
- A virtual diagnostic tool for beam transverse properties at gun exit



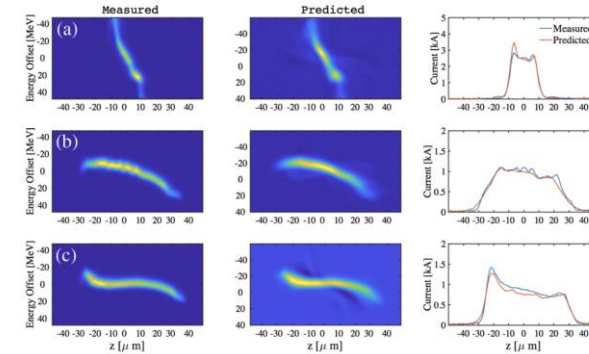
Roussel R, et al. Nature Communications, 2021, 12(1): 1-7.



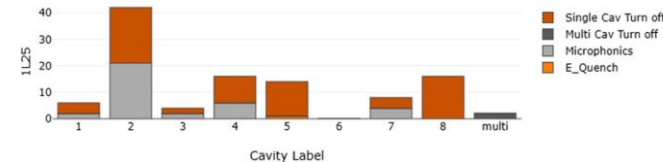
Sanchez-Gonzalez A, et al. Nature Communications, 2017, 8(1): 1-9.



Scheinker A, et al. Physical Review Letters, 2018, 121(4): 044801.



Emma C, et al. Physical Review Accelerators and Beams, 2018, 21(11): 112802.

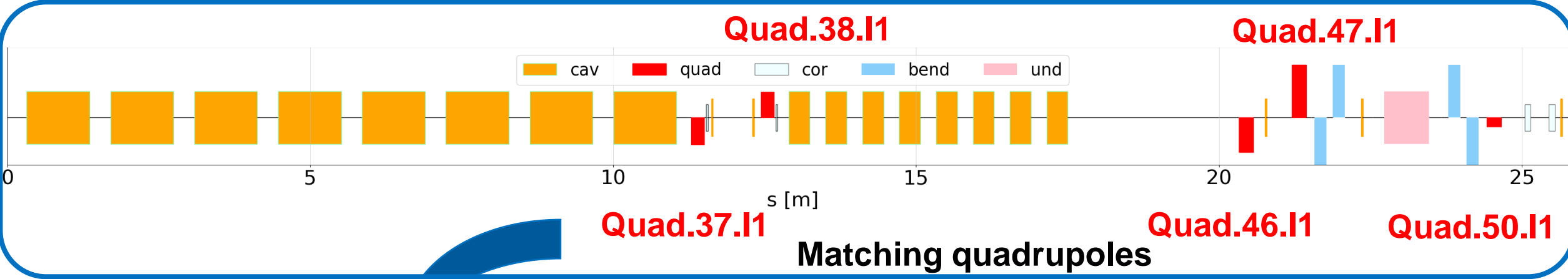


Tennant C et al. Physical Review Accelerators and Beams, 2020, 23(11): 114601.



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**Initial point**  
**Twiss parameter:**  
 $\alpha_{xi}, \alpha_{yi}, \beta_{xi}, \beta_{yi}$

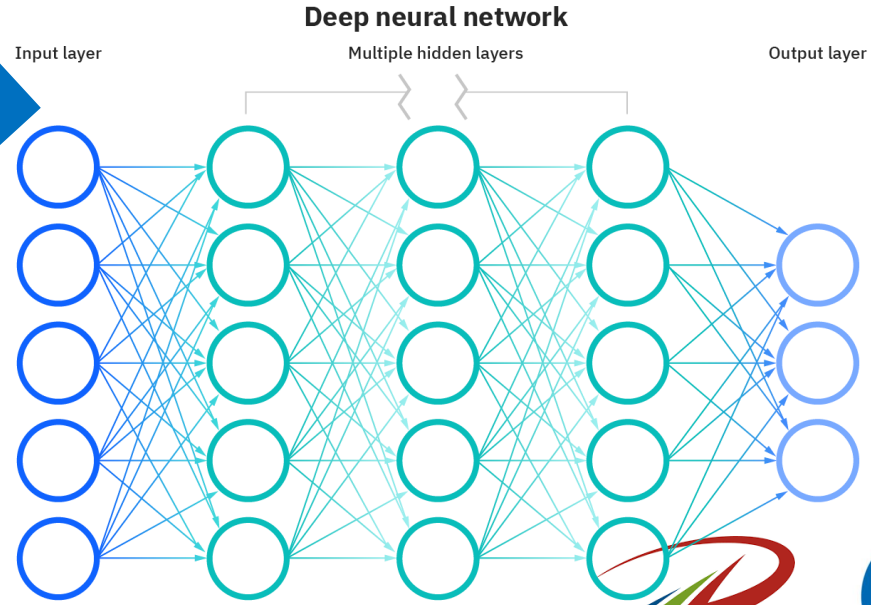
**Reference point**  
**Twiss parameter:**  
 $\alpha_{xr}, \alpha_{yr}, \beta_{xr}, \beta_{yr}$

Input parameters:

- Initial twiss parameters:  $\alpha_{xi}, \alpha_{yi}, \beta_{xi}, \beta_{yi}$
- Matching quads settings: Q37.k1, Q38.k1, Q46.k1, Q47.k1, Q50.k1

Output parameters:

- Final twiss parameters:  $\alpha_{xr}, \alpha_{yr}, \beta_{xr}, \beta_{yr}$

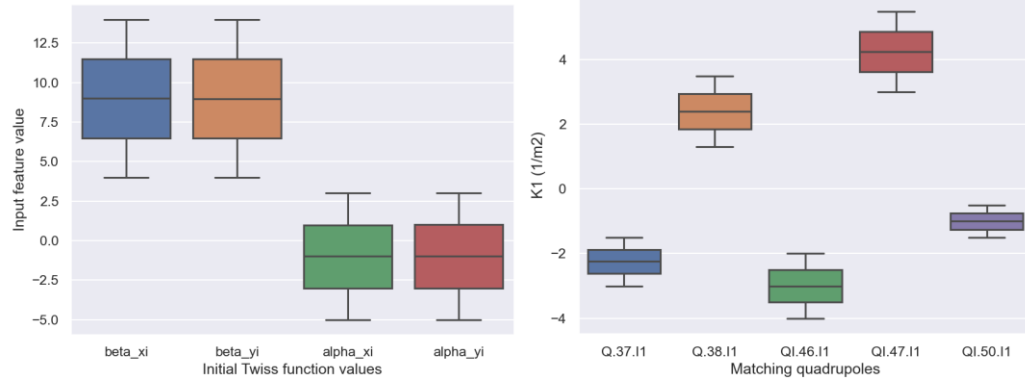


# Sample generation for training

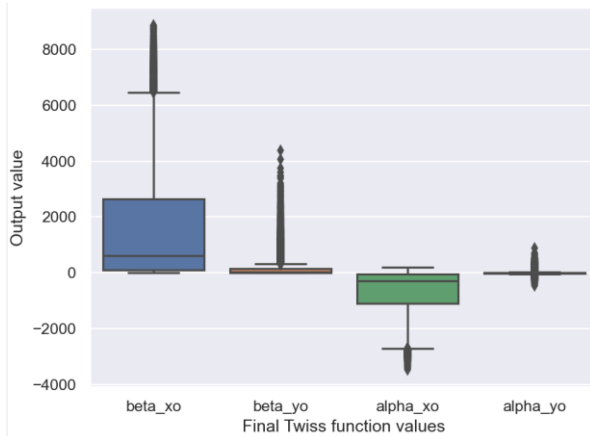
## Difficulty in the sample generation

Most time spent on generating those useless samples 😞

### Input features



### Output parameters

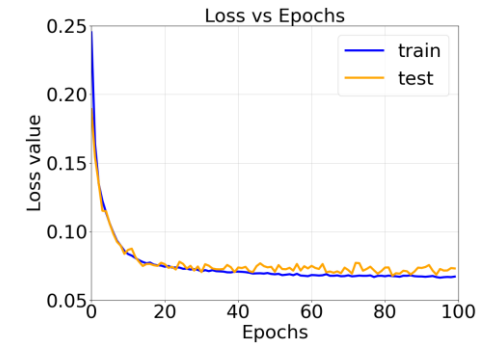
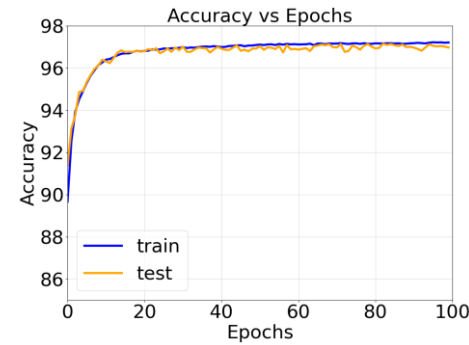


$\alpha_{Xr}: [-10,10]$   
 $\alpha_{Yr}: [-10,10]$   
 $\beta_{Xr}: [0,20]$   
 $\beta_{Yr}: [0,20]$

## Binary classification NN

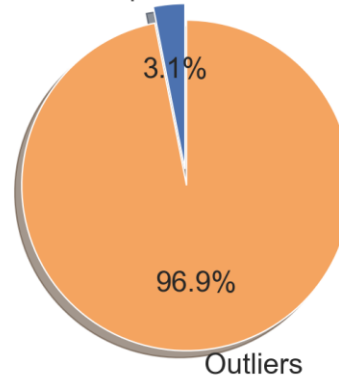
This model acts as a filter for efficient samples generation

- Label assignment for the whole sample set
- Model training and validation



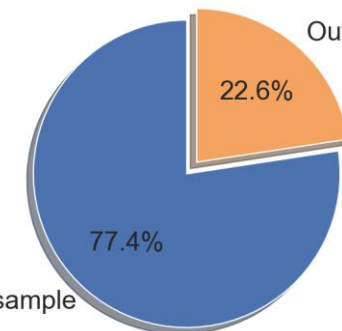
## Model application for sample generation

Normal sample



Outliers

Outliers

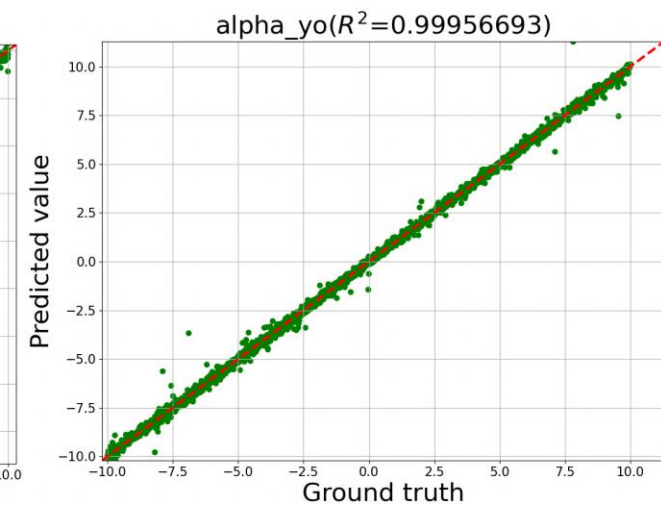
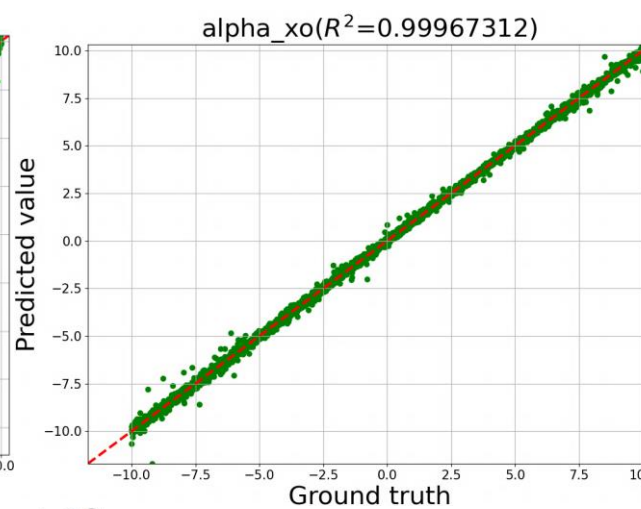
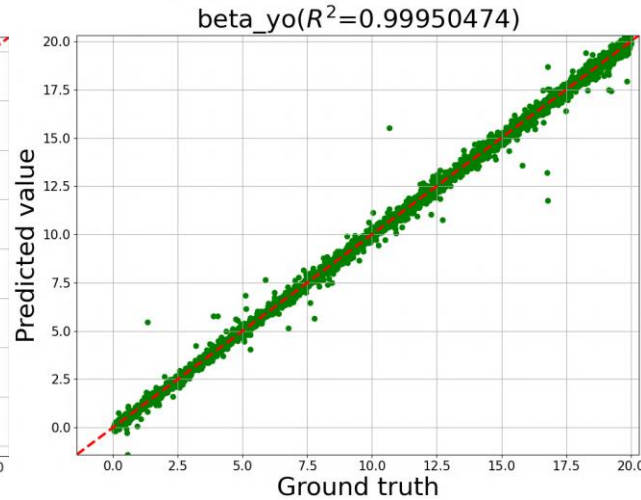
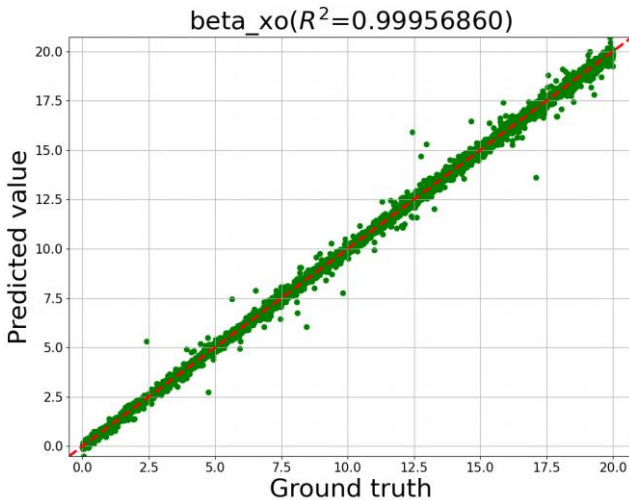


Normal sample

# Surrogate model construction

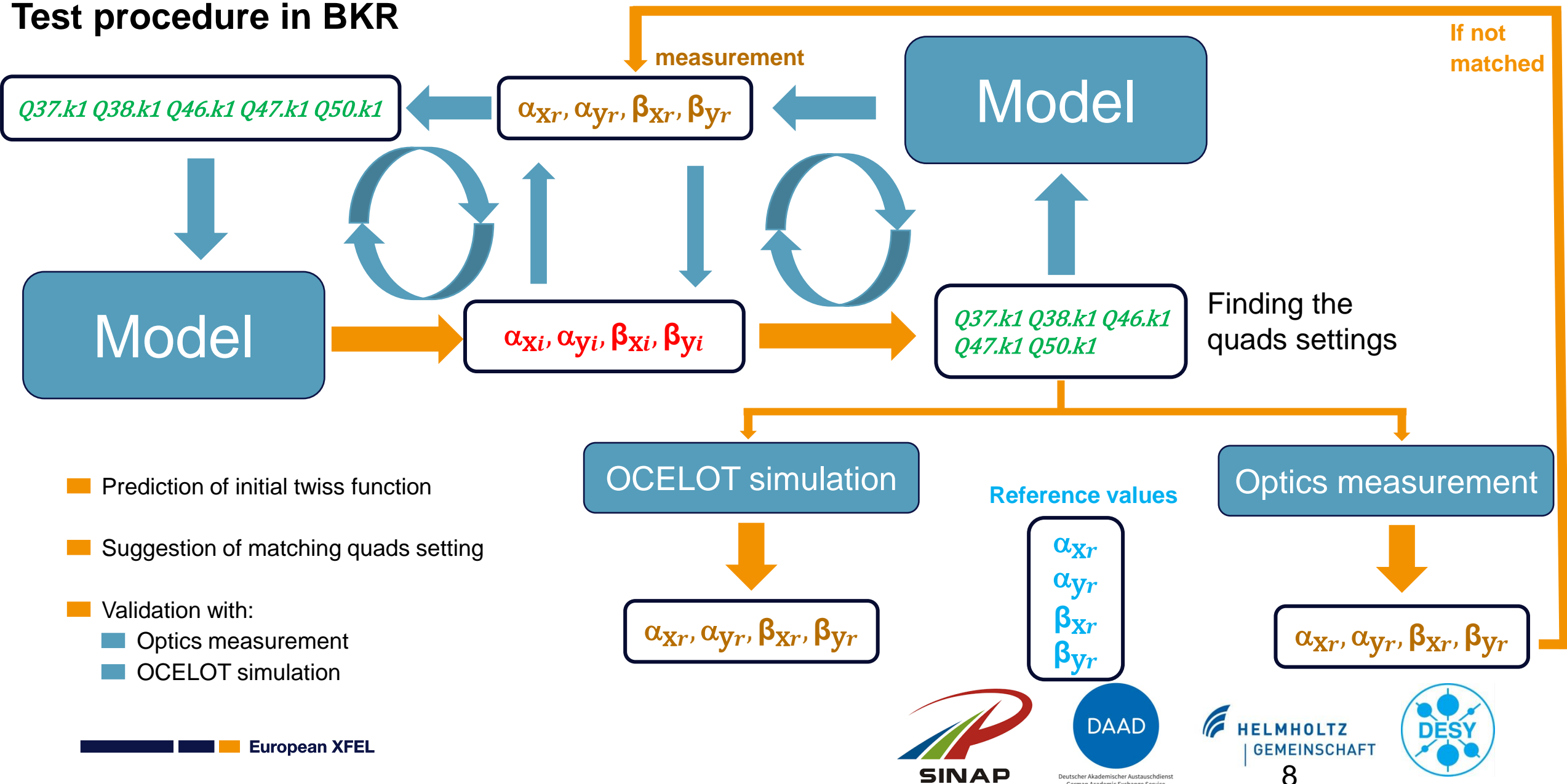
- Samples under OCELOT simulation in Maxwell
- simulation in parallel in 20 nodes (40 CPUs in each)
- Spent ~ 13 h to get sample sets of 40k

- Neural network architecture and hyperparameters
- 200k samples in total (20% for testing)
- 3 hidden layers with 128 neurons in each
- Loss function: Mean square error
- Activation function: Tanh
- Optimizer: Adam



$$R^2 = 1 - \frac{SS_{RES}}{SS_{TOT}} = 1 - \frac{\sum_i (y_i - \hat{y}_i)^2}{\sum_i (y_i - \bar{y})^2}$$

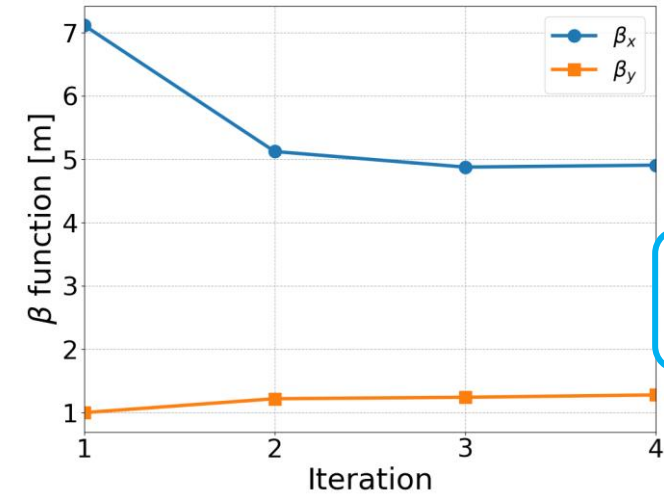
# Test procedure in BKR



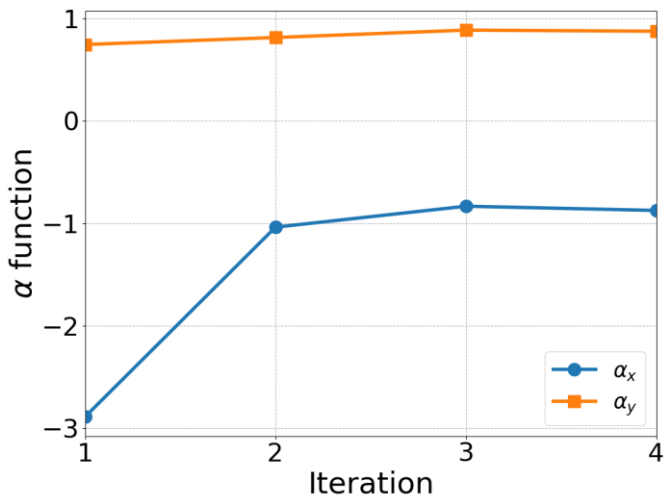


# CASE 1: Solenoid current of 344A (nominated strength)

Optics function values evolution during the optimization



Asymmetry between the transverse planes

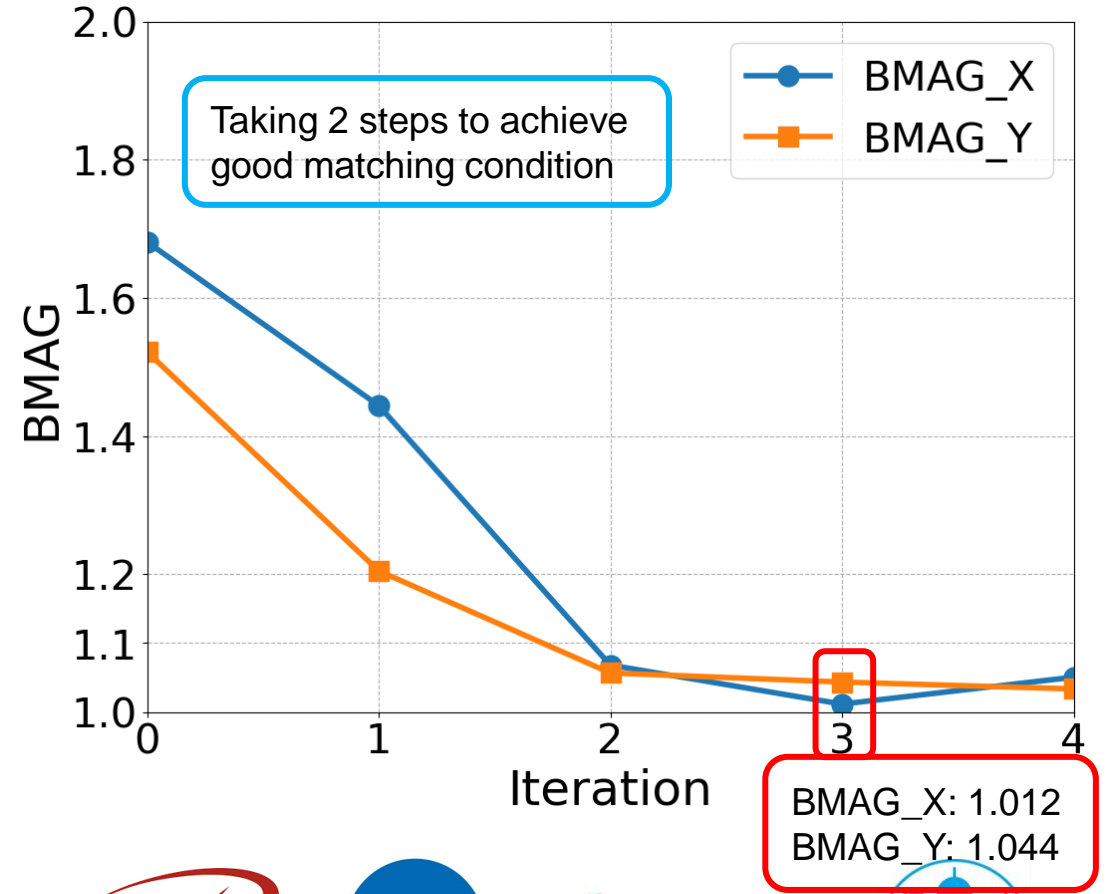


$$\xi = \frac{1}{2} (\gamma\beta_0 - 2\alpha\alpha_0 + \beta\gamma_0)$$

$$BMAG = \xi + \sqrt{\xi^2 - 1}$$

$$\frac{\beta_0}{BMAG} < \beta < BMAG \cdot \beta_0$$

Mismatch parameters evolution during the optimization

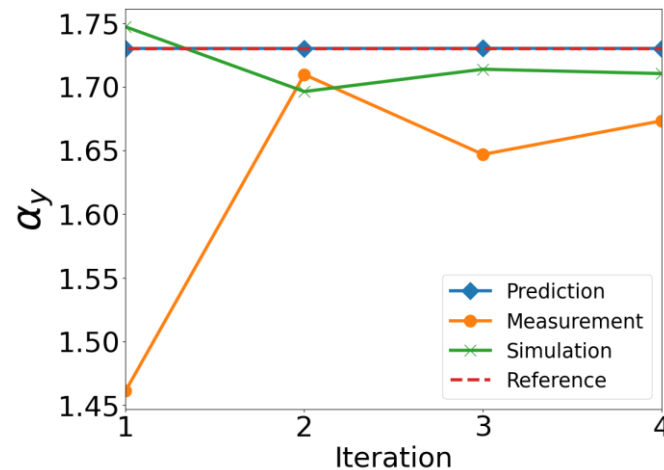
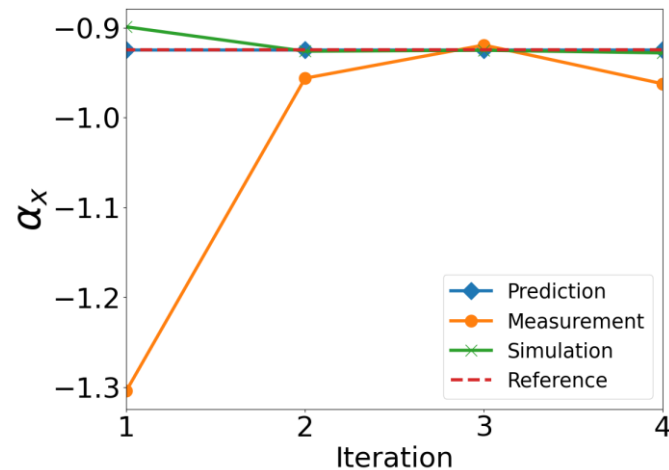
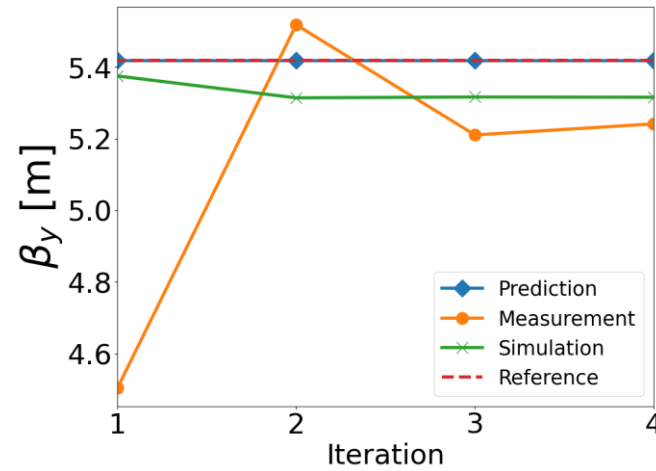
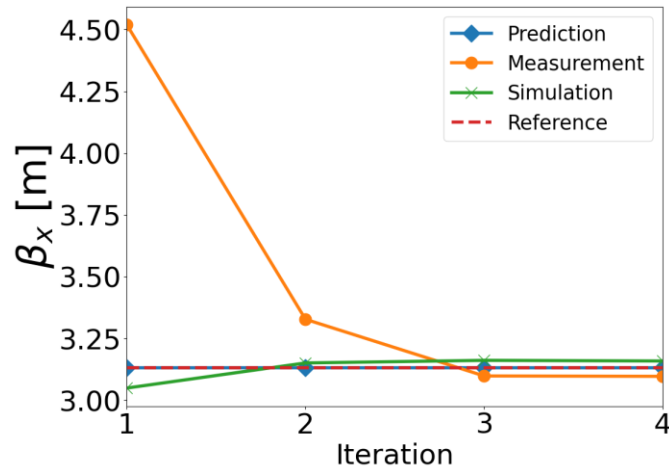


Taking 2 steps to achieve good matching condition

BMAG\_X: 1.012  
BMAG\_Y: 1.044

# CASE 1: Solenoid current of 344A (nominated strength)

Optics function values evolution during the optimization



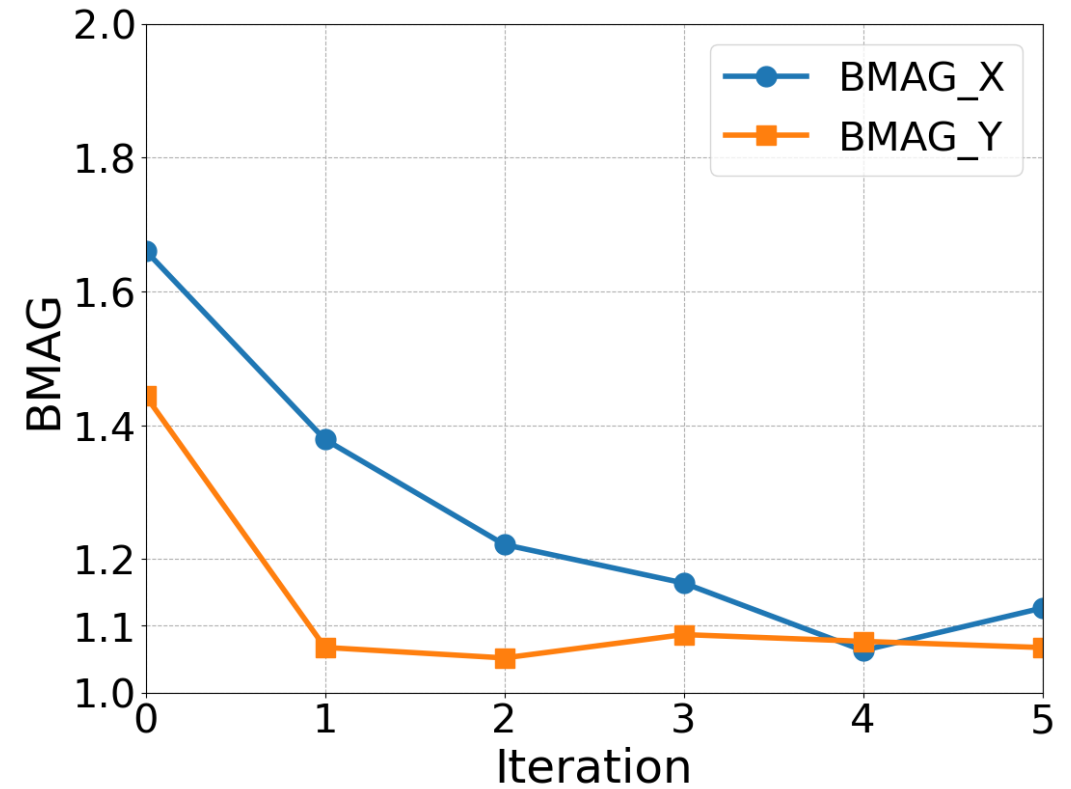
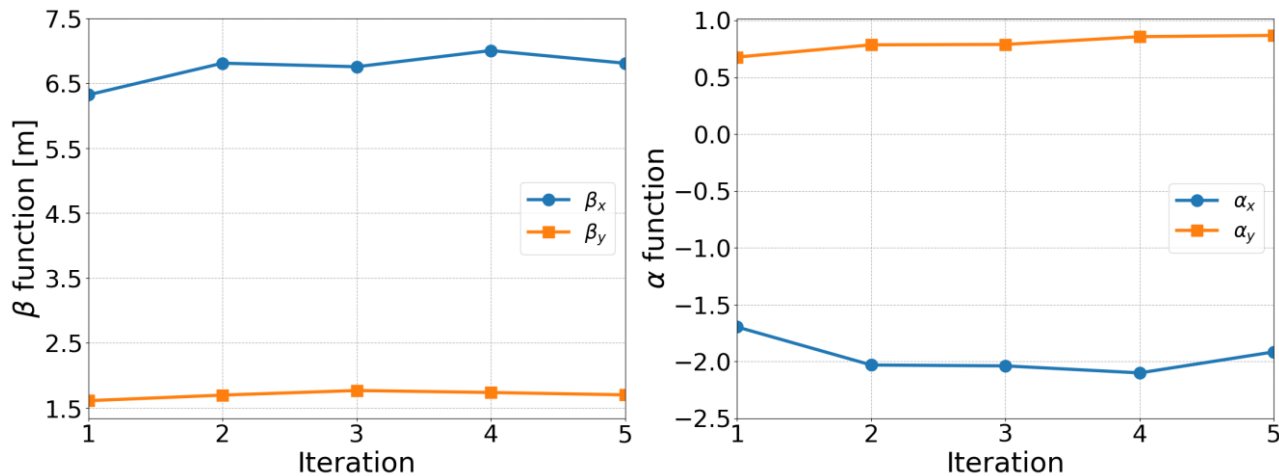
- Model can find the solution based on its knowledge
- Prediction from the surrogate model agrees well with the simulation

# CASE 3: Solenoid current of 343.5 A

■ Optics function values evolution during the optimization

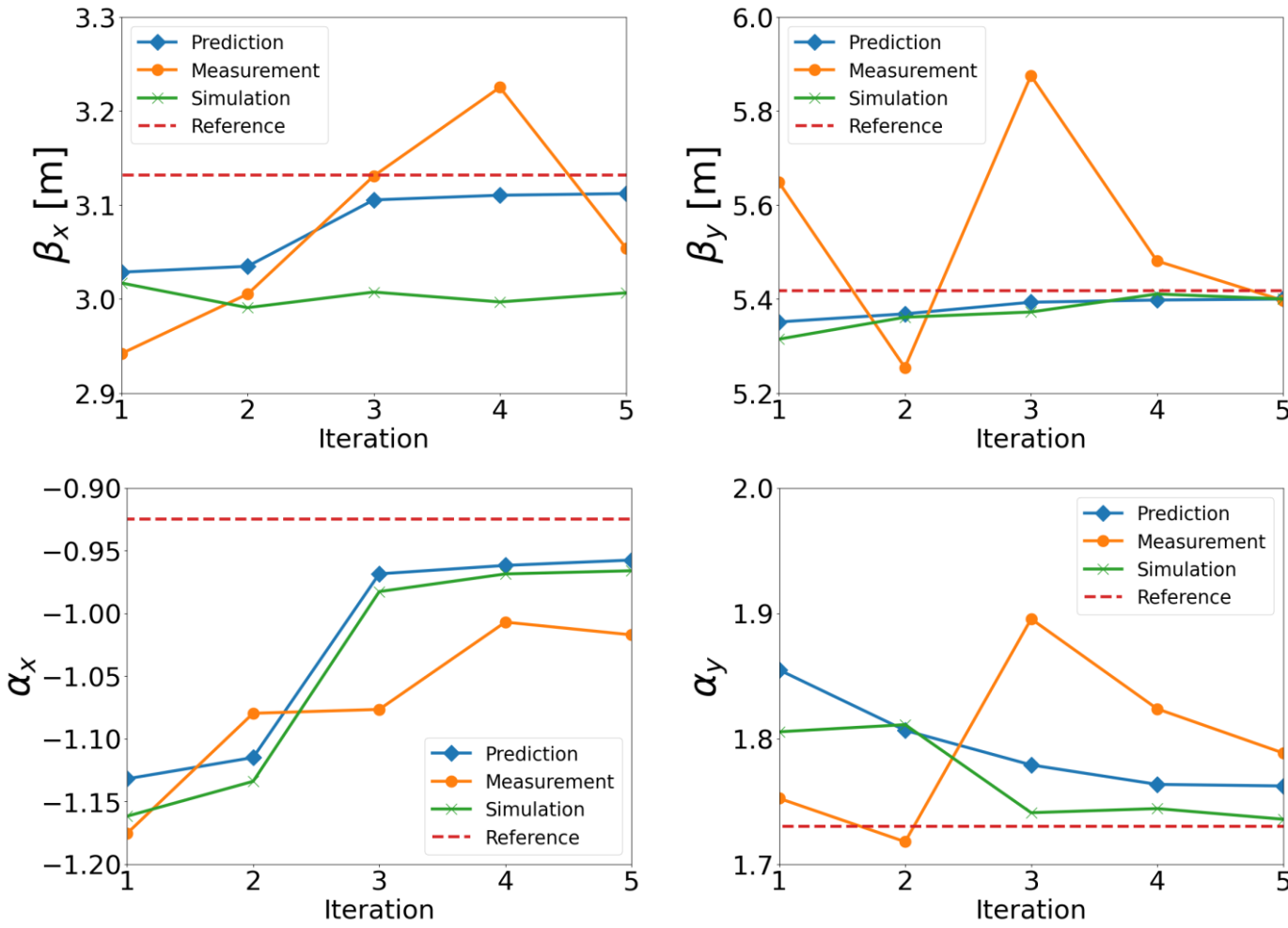
■ Mismatch parameters evolution during the optimization

- Beam initial condition changed after solenoid tuning
- Keep the Q.50 strength fixed, adjust the other 4 quads
- Matching achieved, but see some limitations



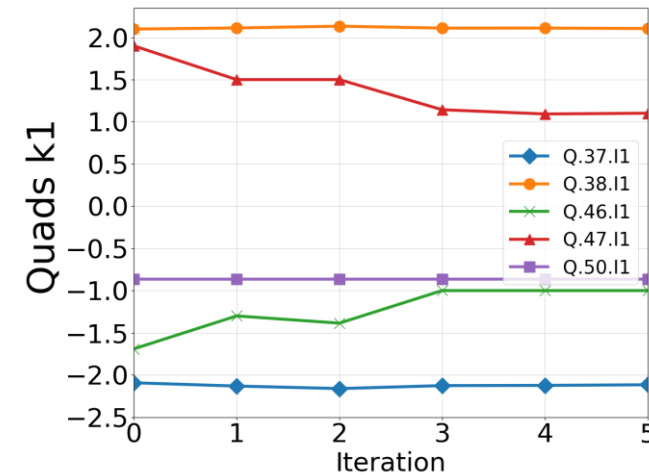
# CASE 3: Solenoid current of 343.5 A

Optics function values evolution during the optimization



There is always gap between the predicted values and reference value

Good matching condition achieved in 4<sup>th</sup> step



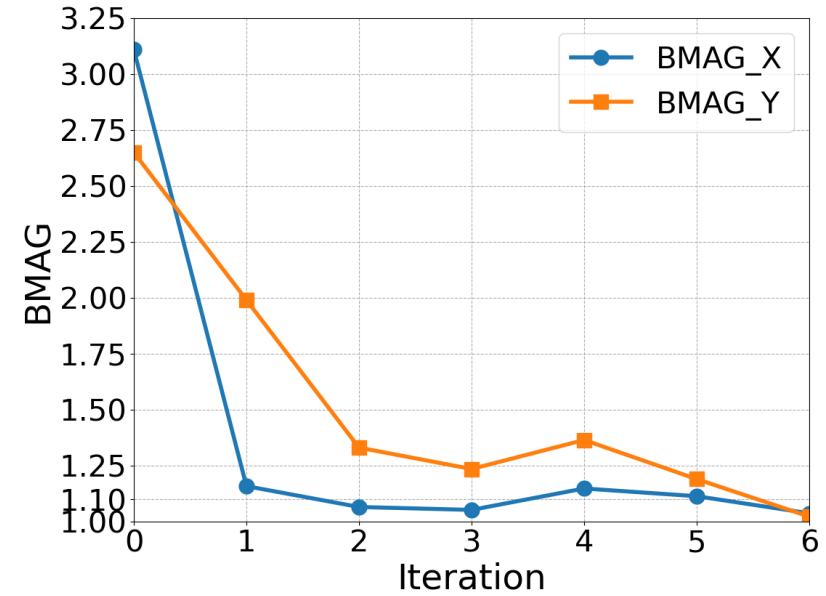
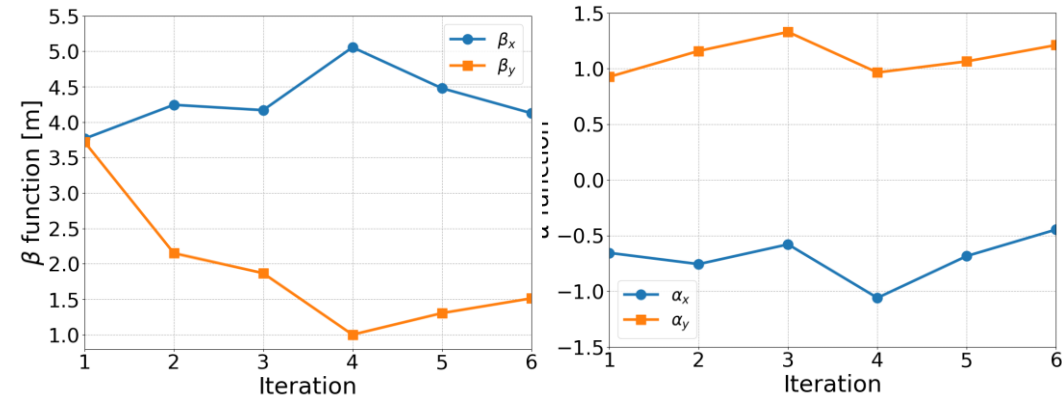
Q47: Adjust the lower bound from 1.5 to 1.0

Q46: hit the upper bound (-1.0)

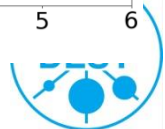
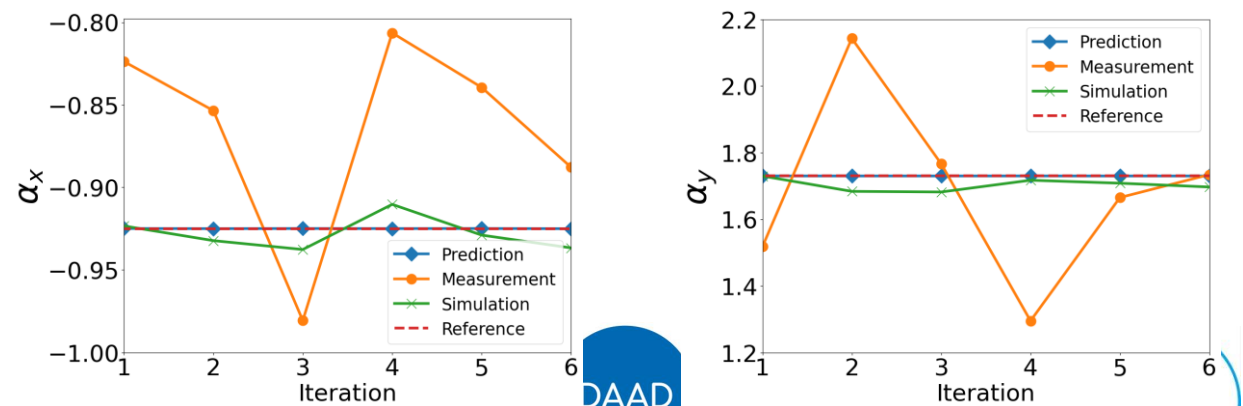
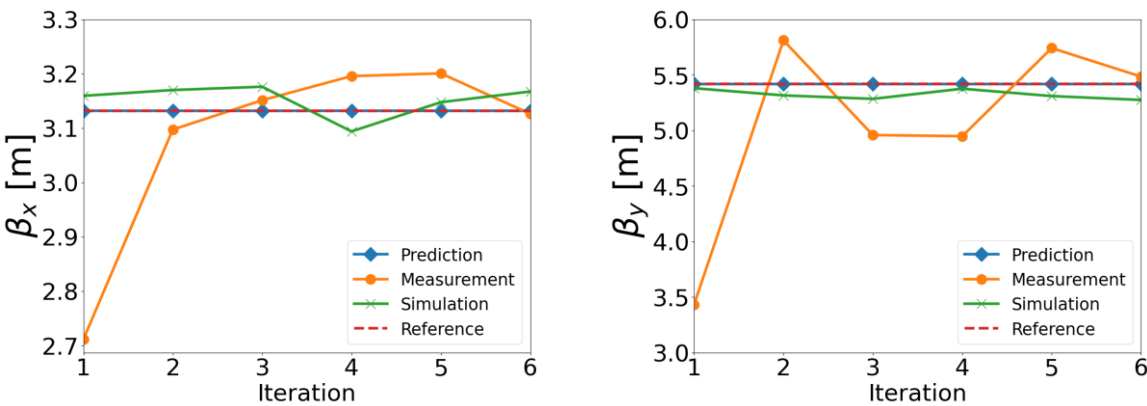
# CASE 2: Solenoid current of 344.5 A

■ Optics function values evolution during the optimization

■ Mismatch parameters evolution during the optimization



■ Optics function values evolution during the optimization



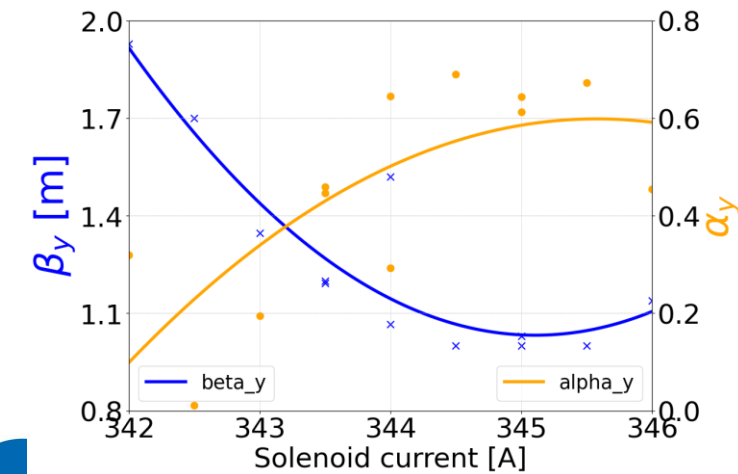
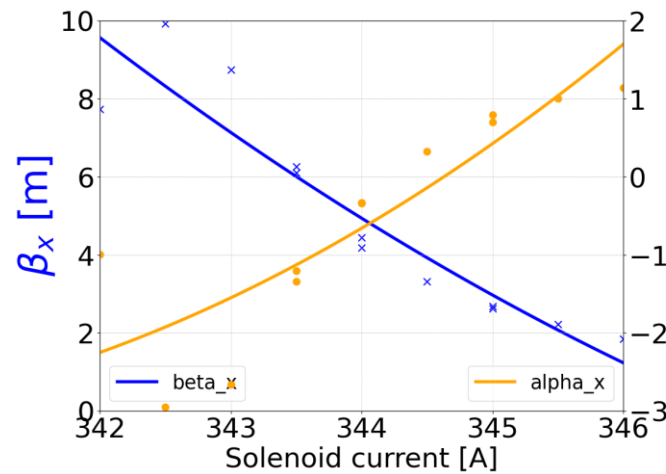
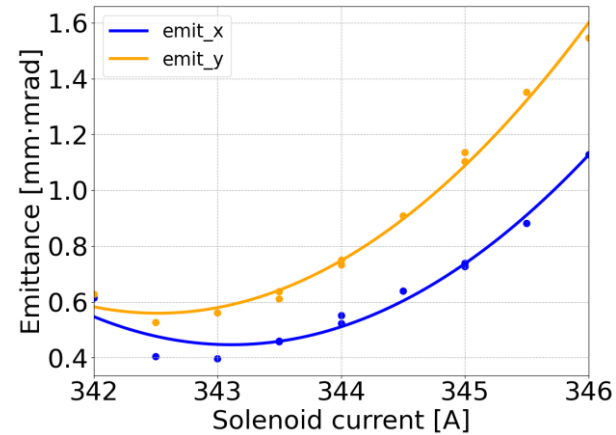
# Solenoid strength scan

■ Transverse emittance at the matching position dependent on solenoid strength

■  $\pm 2$  A around the nominated strength (344 A)

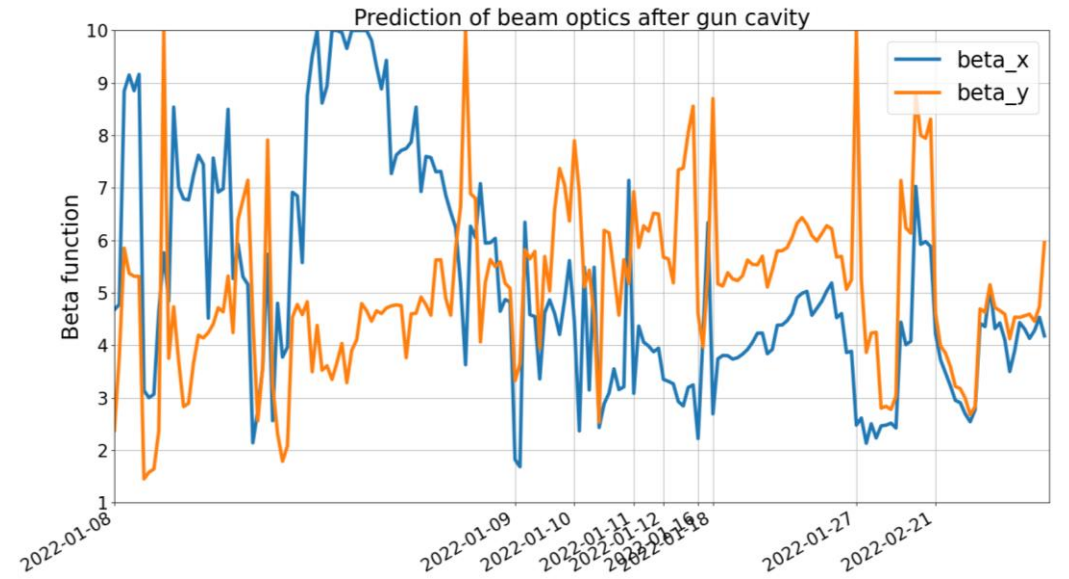
■ Keep the machine lattice the same during the scan

■ Prediction of initial twiss function values

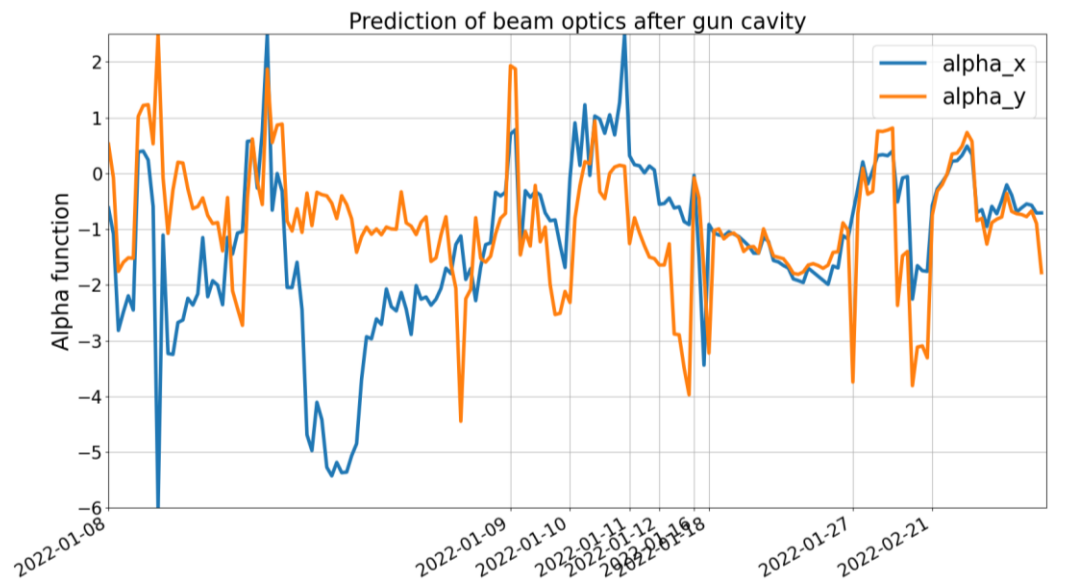


# Application to beam diagnostics at gun cavity

- Collection of all the optics measurement data in BKR from 08/01/2022 to 21/02/2022 (189 samples in total)
- Prediction of initial twiss function values from each samples
- Potential further application for research on:
  - Beam transverse properties optimization at gun cavity
  - Laser stability on the cathode
  - Emittance compensation with solenoid



Set-up the machine after shutdown | Beam delivery, SASE2 tuning | Gun study



# Summary

## Achievements

- The prediction of the constructed model agrees well with the result from OCELOT simulation
- The surrogate model can achieve good matching scenario ( $\text{BMAG} < 1.1$  in two planes) under multiple solenoid settings
- The model can be applied as a diagnostics tool towards beam transverse properties at the gun cavity exit.

## Outlook

- Improvement of model capability (wider parameters search space)
- Involving knobs in the gun cavity to the surrogate model
- Exploration of more machine learning based online optimization and virtual diagnostic application.