



Beam qualification using Machine Learning Techniques applied to Libera Brilliance+

Libera Workshop 2025

dr Jacek Biernat, Michał Żurek, Maciej Mleczko, Mikołaj Wróbel

SOLARIS National Synchrotron Radiation Centre

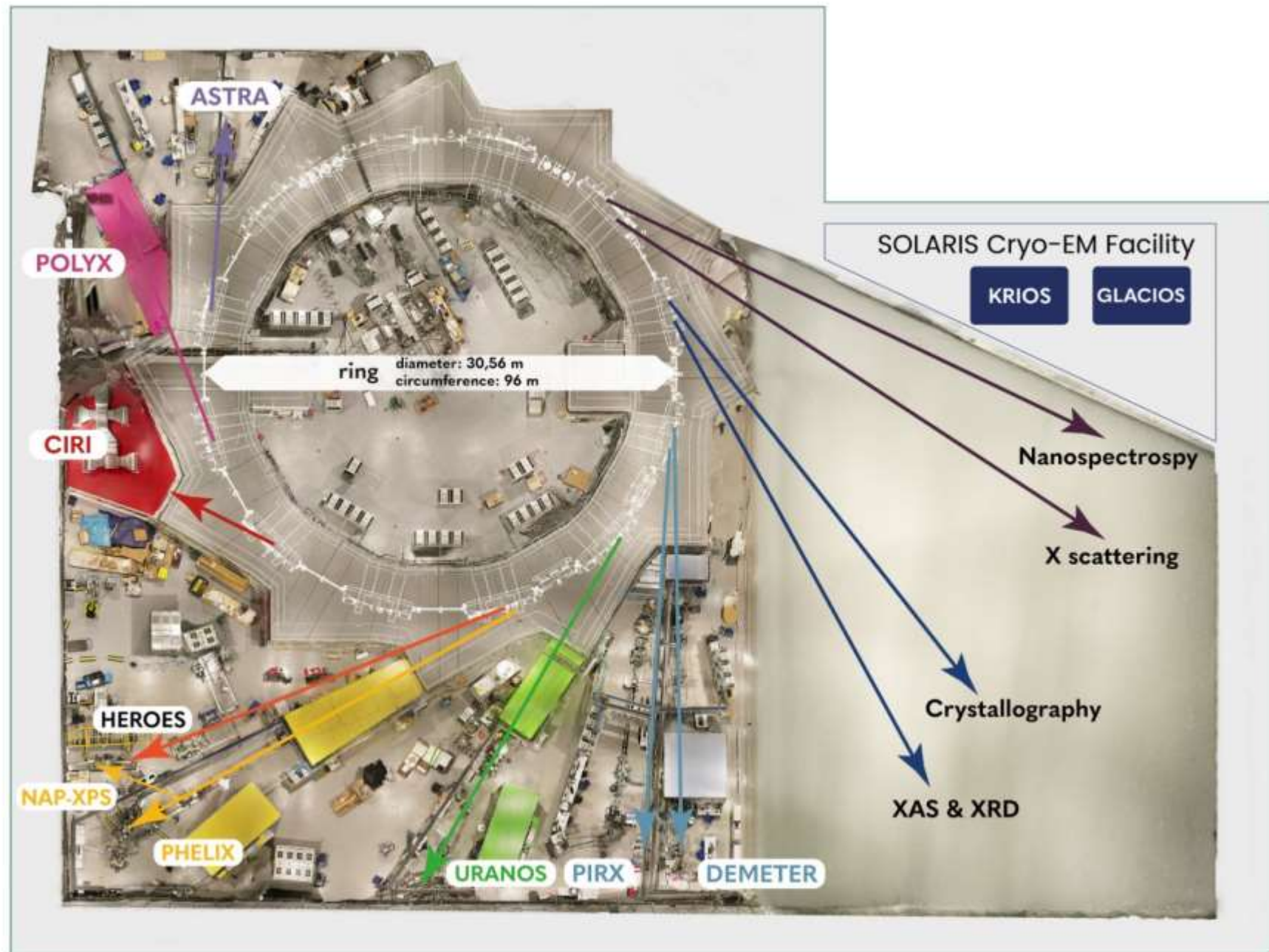
Agenda

- Introduction - SOLARIS facility
- Data streaming system
- Basic beam diagnostics using ML
- Plans for the future, sector wise diagnostic

SOLARIS facility

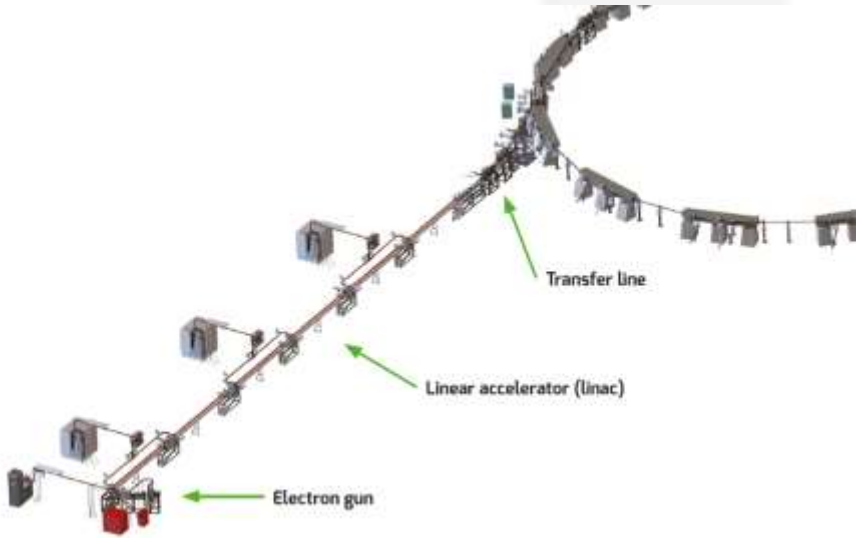
Synchrotron SOLARIS

- 3rd generation light source
- 7 fully operational beamlines
- Two more beamlines are under construction
- The only synchrotron in Central Eastern Europe located in Poland



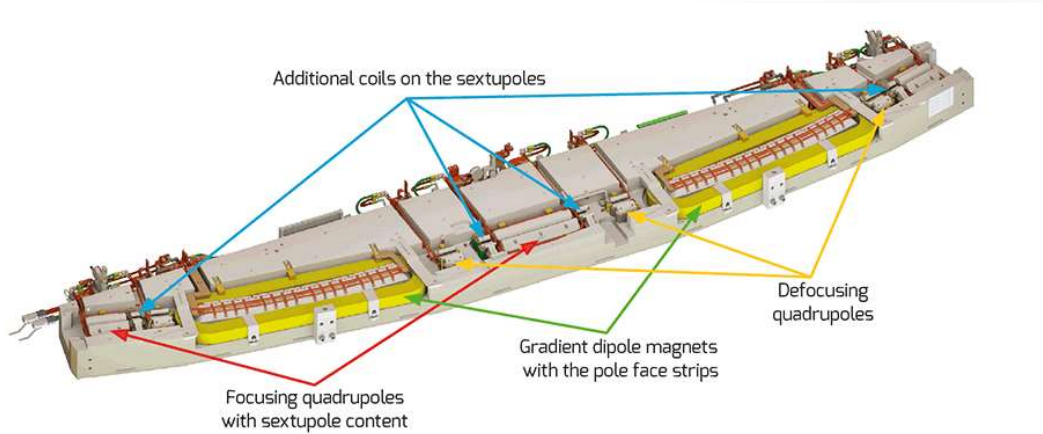
Linear accelerator

- 3 modulators,
- 6 accelerating structures,
- Delivered injection Energy - **536MeV**,
- Max beam current on output - **26 mA**.



Storage ring

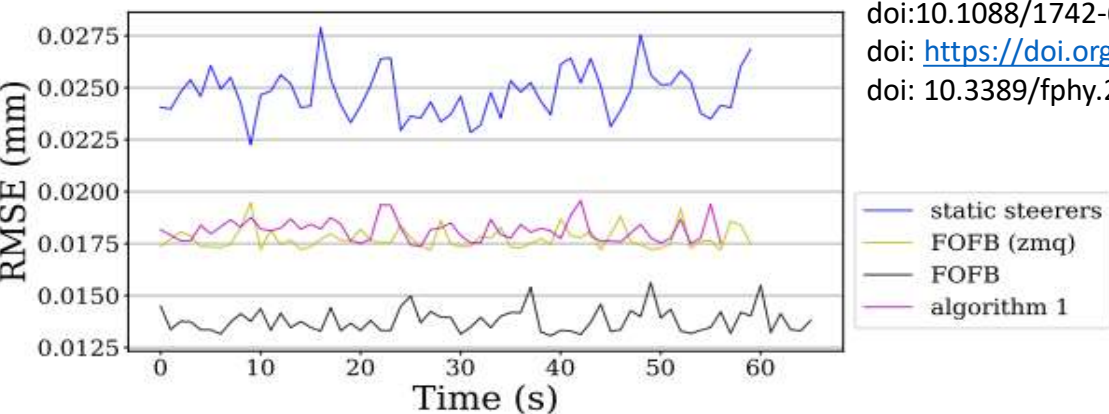
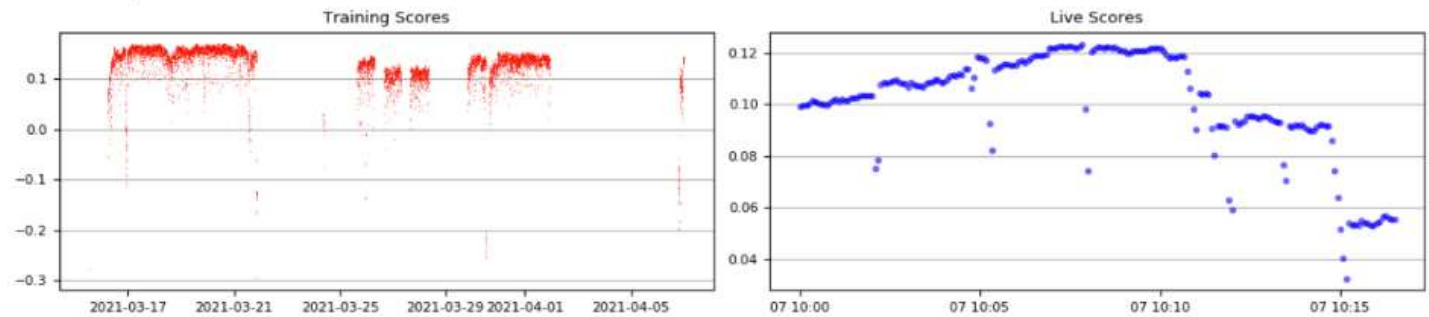
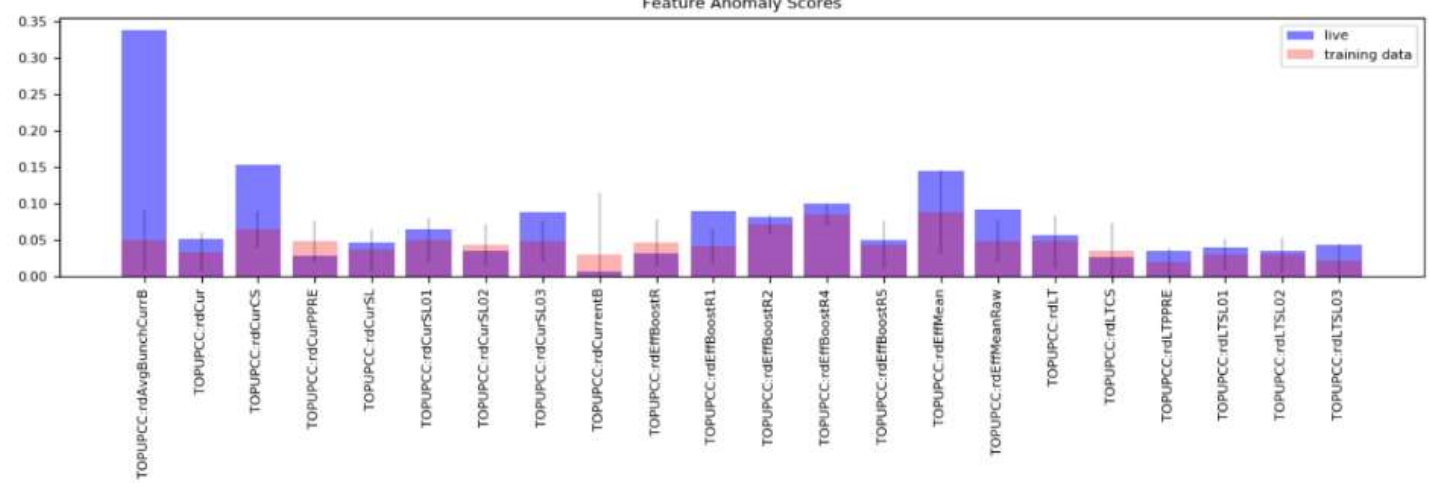
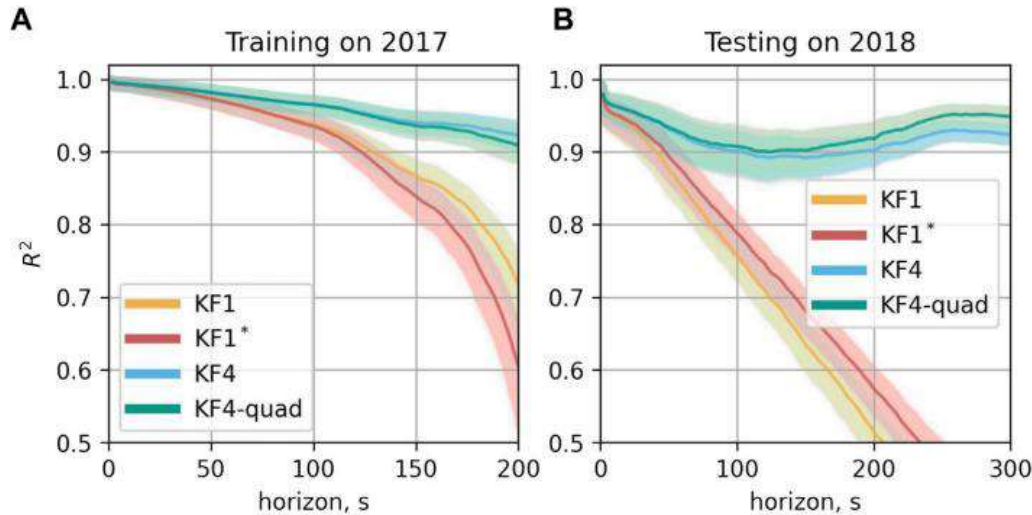
- Consists of **12 DBA cells**,
- Main RF frequency - **99,937 MHz**,
- Energy - **1,5 GeV**,
- Max beam current - **500 mA**,
- Circumference - **96 m**.



Machine learning in Synchrotron

What is done today

- 1. Optimization of injection efficiency
- 2. Beam loss forecasting and prevention
- 3. Beam stability enchantment
- 4. Anomaly detection
- 5. Accelerator lattice optimization



doi:10.1088/1742-6596/2687/6/062033
doi: <https://doi.org/10.32479/ijeep.7605>
doi: 10.3389/fphy.2022.960963

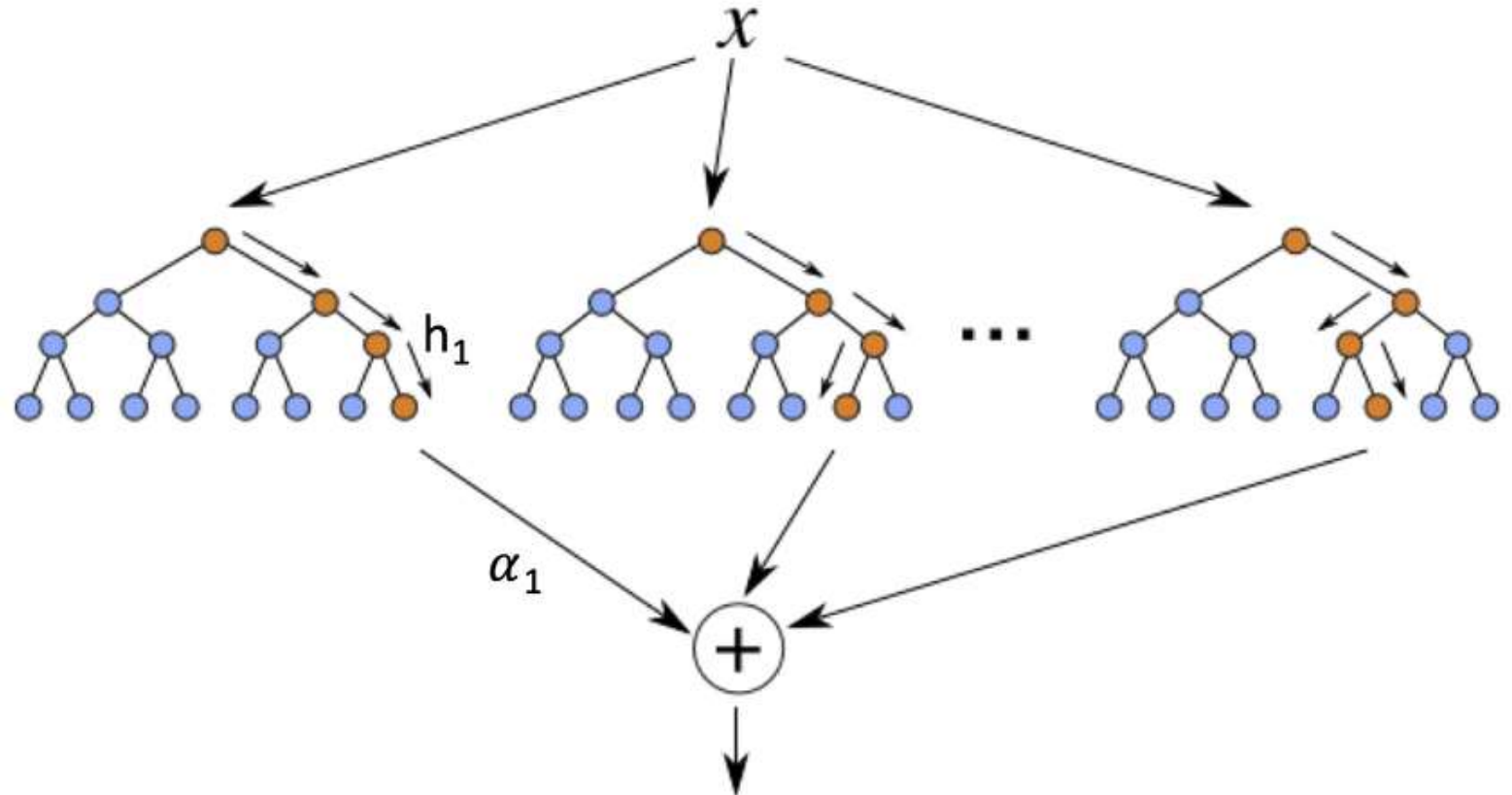
Beam anomaly detection - **Boosted Decision Tree**

XGBoost tree configuration

- Estimators numer: **600**
- Learning rate: **0.01**
- Max depth: **10**
- Objective: **binary:logistic**

Scenarios data collection

- RF active cavity power drop
- Magnet current fluctuation

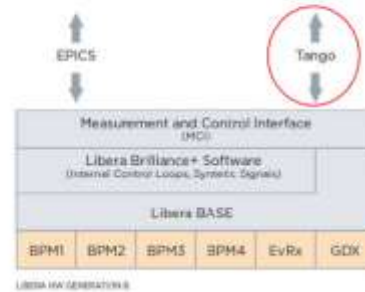


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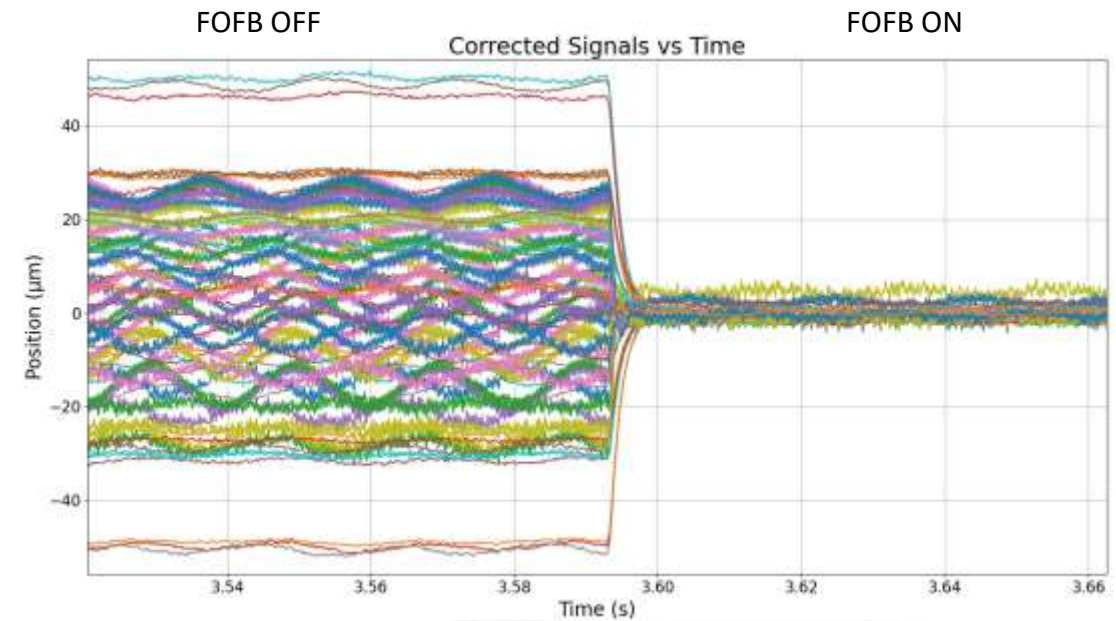
We currently operate:

- 36 Libera Brilliance + (connected in 12 chassis) connected to 36 BPMs (one BPM in X,Y plane)
- We use GDX module for fast orbit feedback

- 4 Libera Photon for beam light measurement



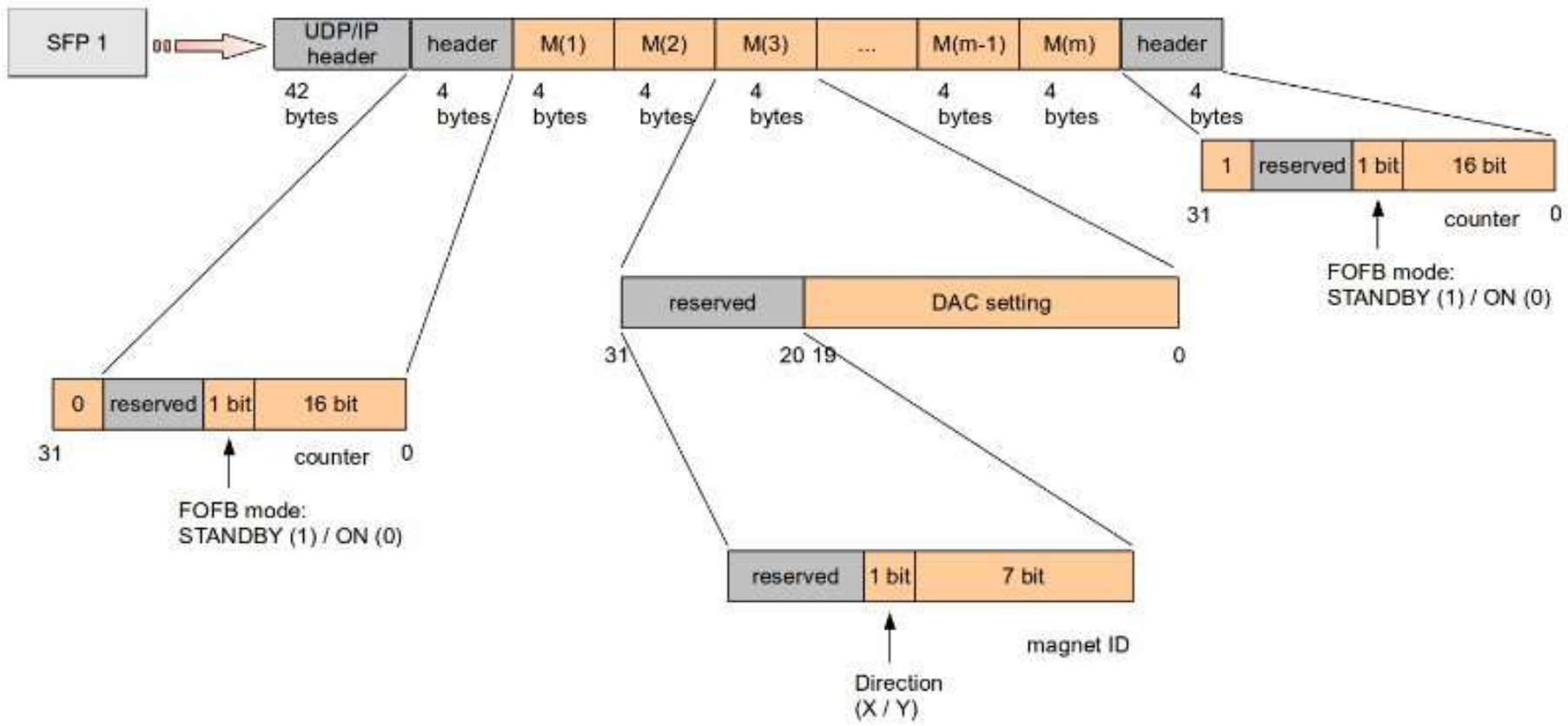
We classify data coming from all 36 BPMs, collected at 10 kHz, and converted to FFT spectrum



Streaming System

```
/* Magnet mode header */
struct libera_magnet_header_s {
    unsigned int counter : 16;
    unsigned int fofb_mode : 1;
    unsigned int reserved : 14;
    unsigned int end_of_packet : 1;
};

/* Magnet mode payload structure */
struct libera_magnet_s {
    unsigned int dac_setting : 20;
    unsigned int magnet_id : 7;
    unsigned int dir_x_y : 1;
    unsigned int reserved : 4;
};
```



*dac_setting is casted and sign extended to signed int 32-bit

First header, the payload structures with magnet position

Streaming System

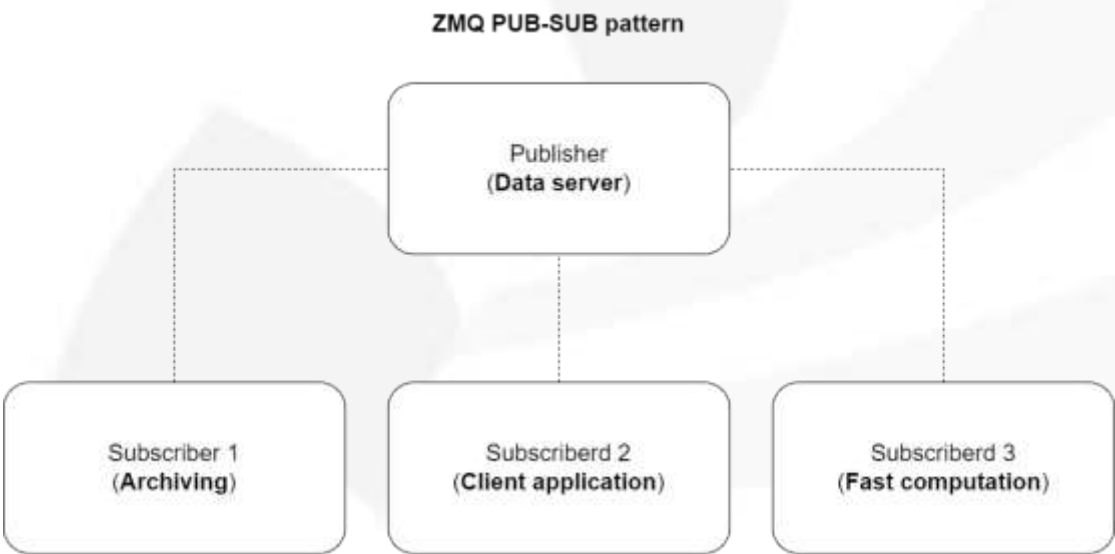
Packet ID (Libera counter)	BPM ID	Pos X	Pos Y	SUM
23365,0	152558	-127538	113087912	
23365,1	188894	-367270	145580692	
23365,2	-346918	-150797	110654149	
23365,3	-162470	48315	152579006	
23365,4	-398382	27917	193087428	
23365,5	168436	189321	151949427	
23365,6	234312	253477	112179604	
23365,7	-376408	-37319	142625063	
23365,8	-13211	-28653	109283688	
23365,9	-84731	276404	98039346	
23365,10	178814	-29177	125773017	
23365,11	-114600	154315	92900765	

- This program opens UDP socket on the working machine then captures all the data sent by Libera and stores in e.g. CSV format
- Parse bitfields, format the data to nice, ASCII form
- Store received data in CSV file
- Possibility to capture the BPM & magnets setpoints data

Streaming System

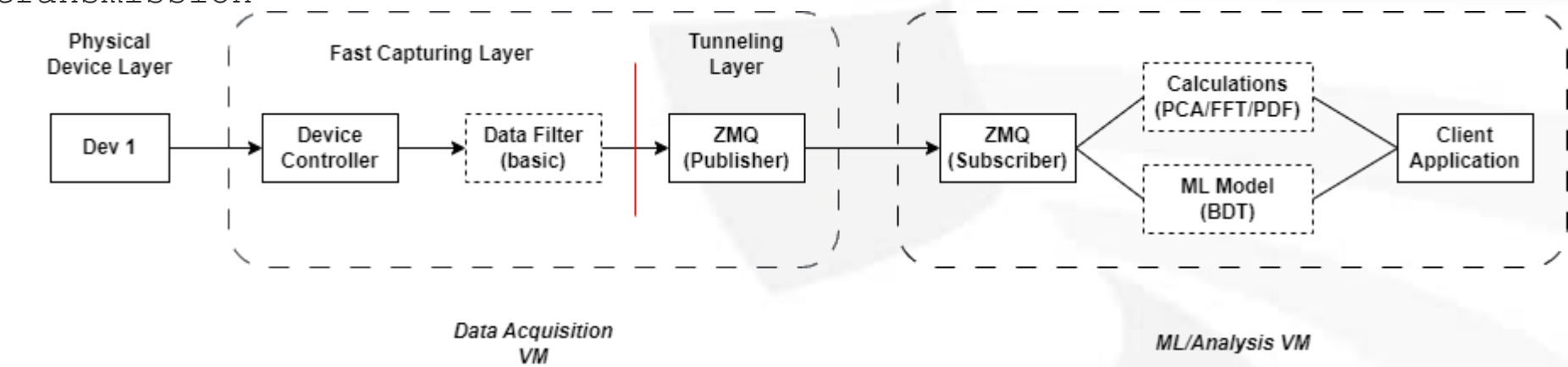
Technology:

- Written in Python 3.9
- ZMQ framework
- Build based on **Publisher-Subscriber** pattern



Abilities:

- Operating on 10 kHz signal
- 3 MB/s online transmission
- Scallable

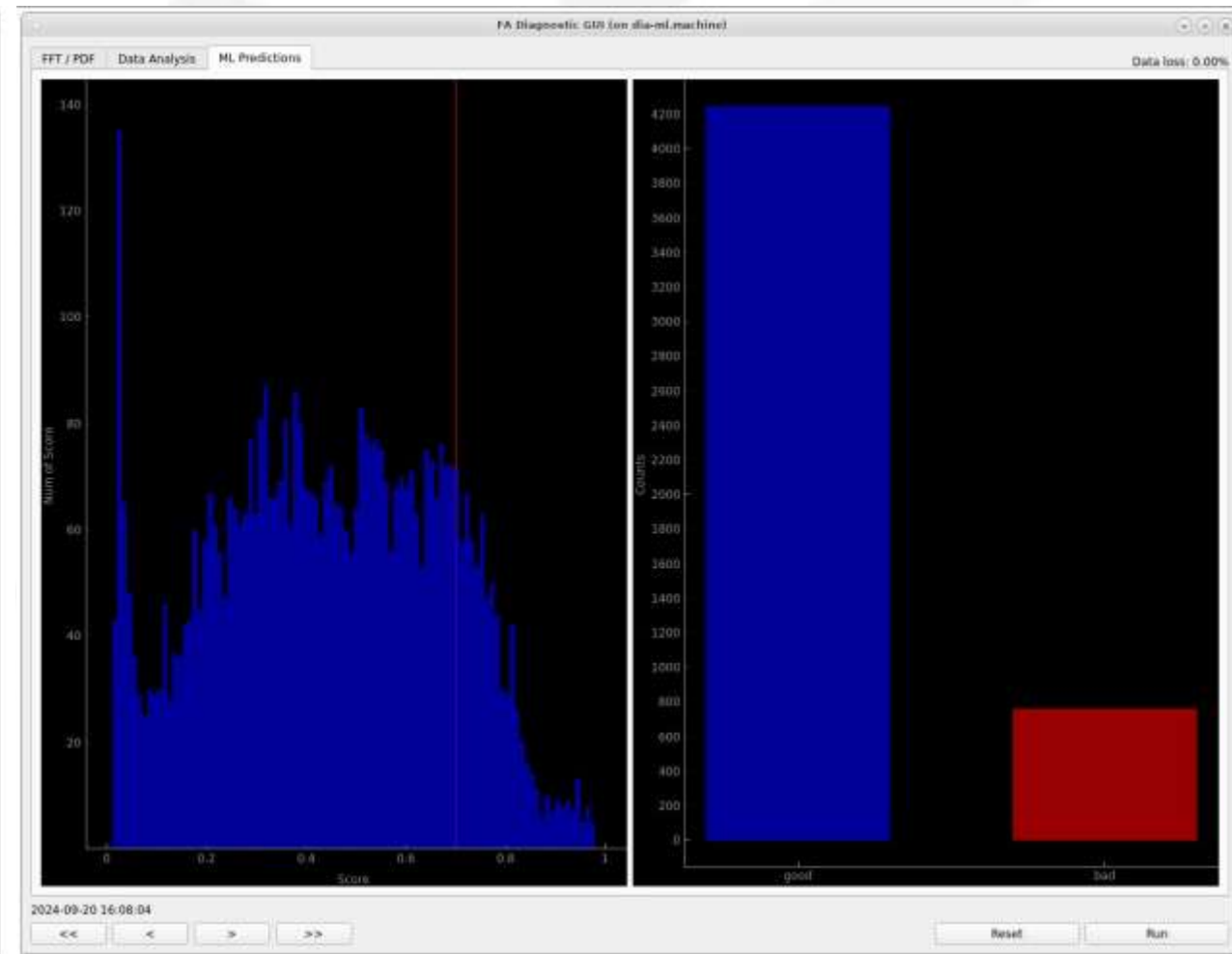
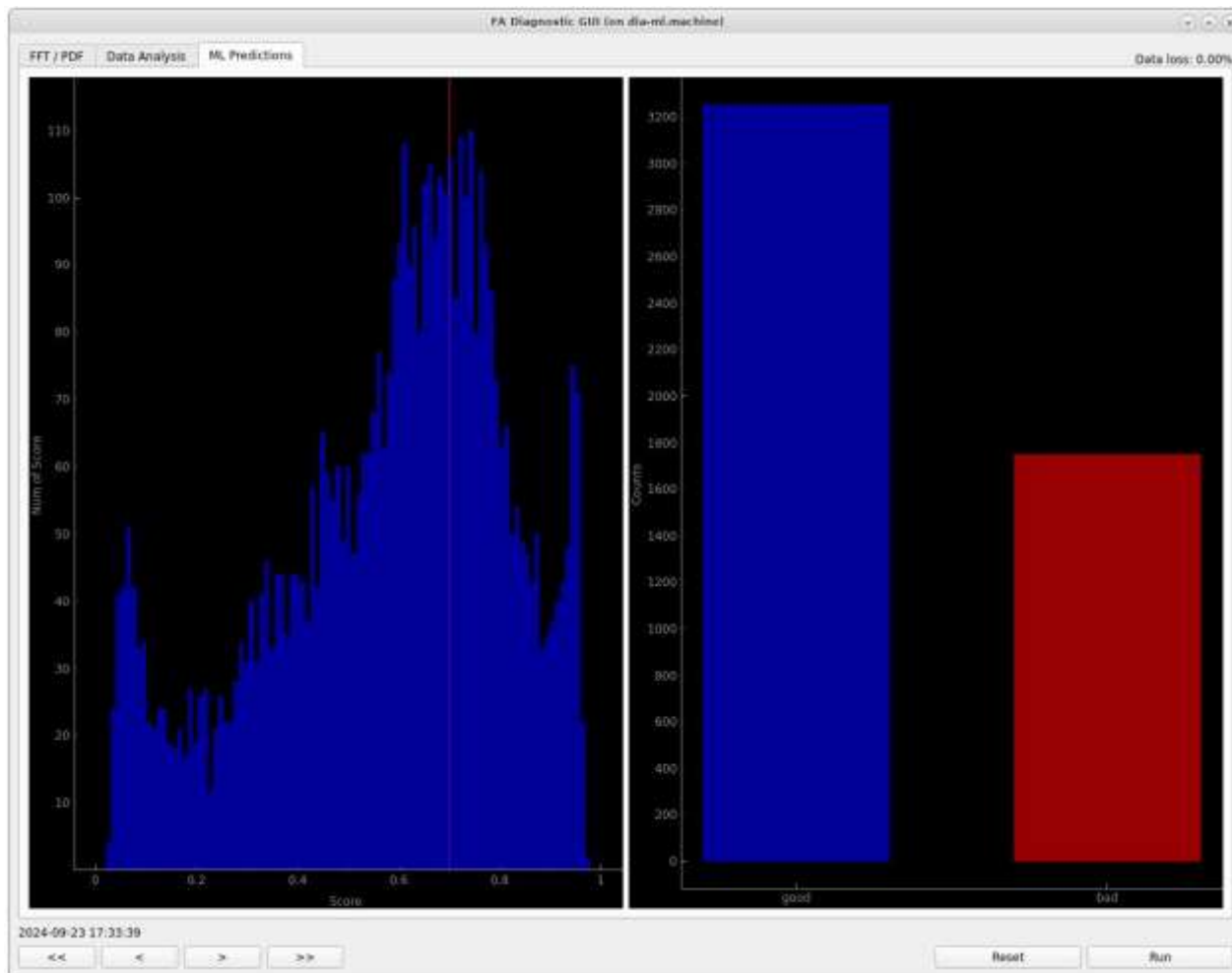


Machine learning in Synchrotron

Beam anomaly detection - **Online application**

Kicker, 3000 V, 120 mA

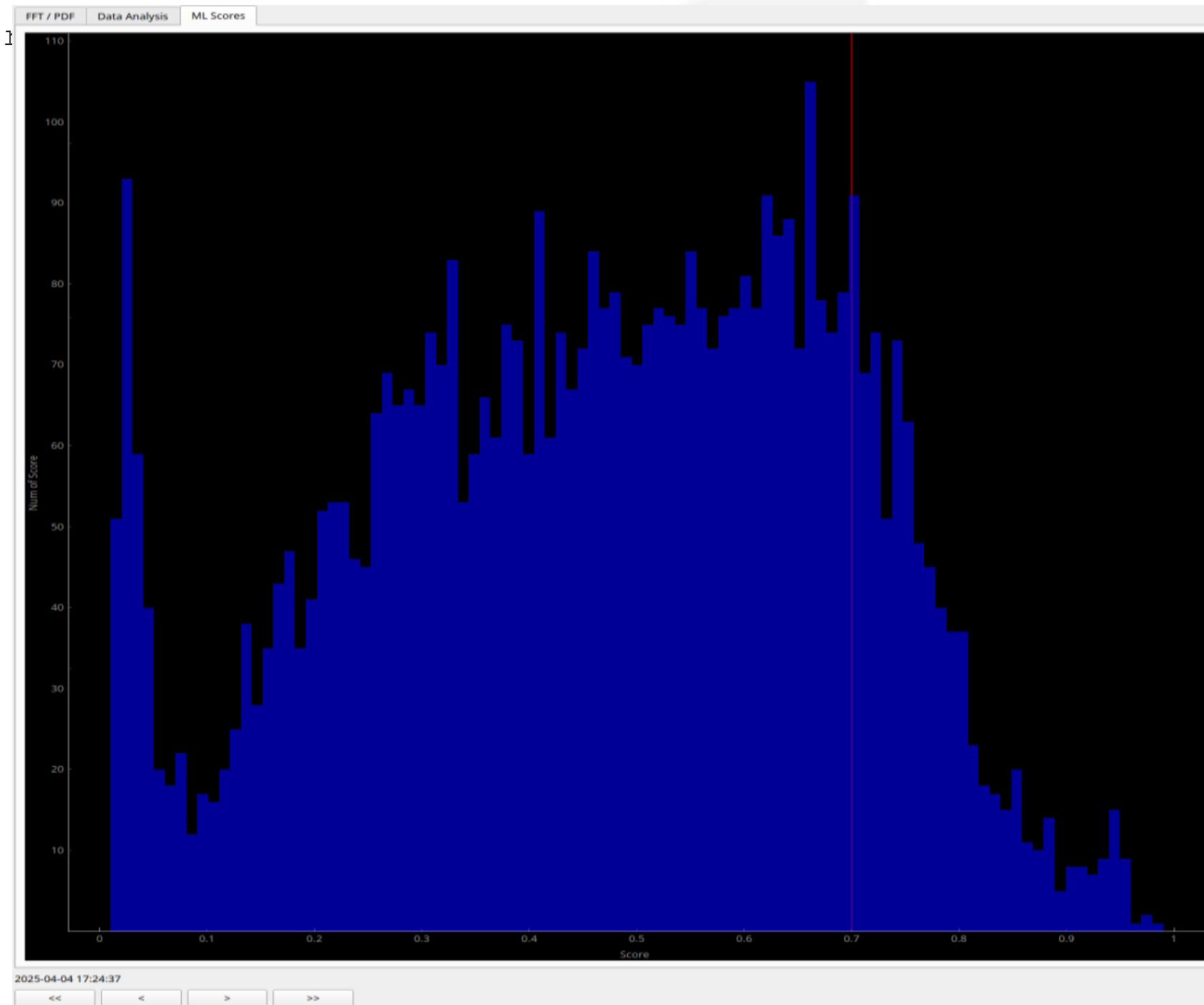
Typical operation



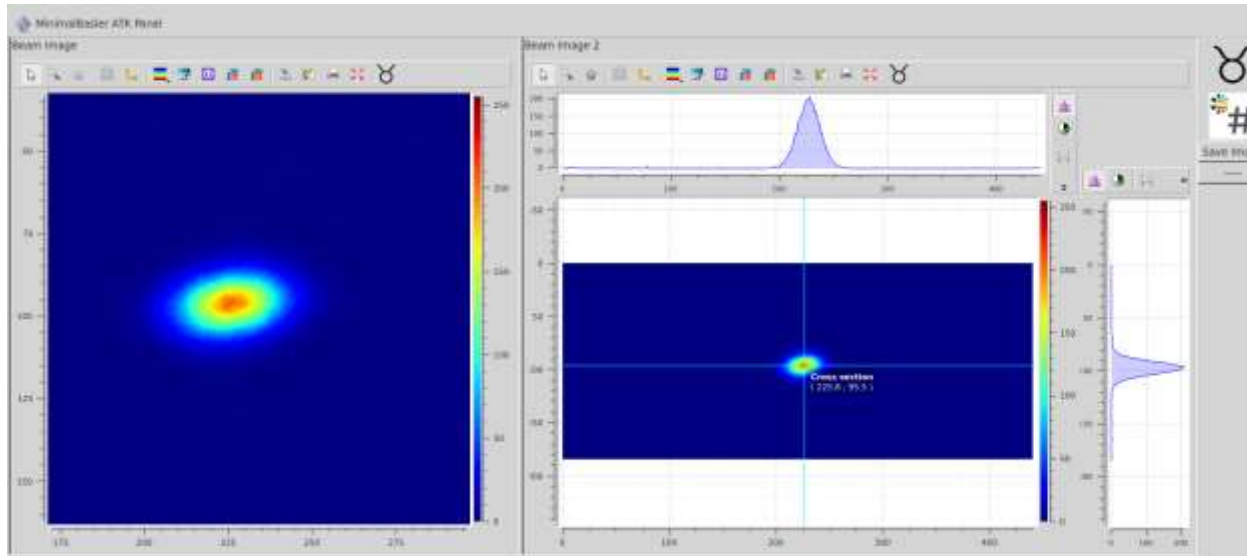
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Beam anomaly detection - **Online application**

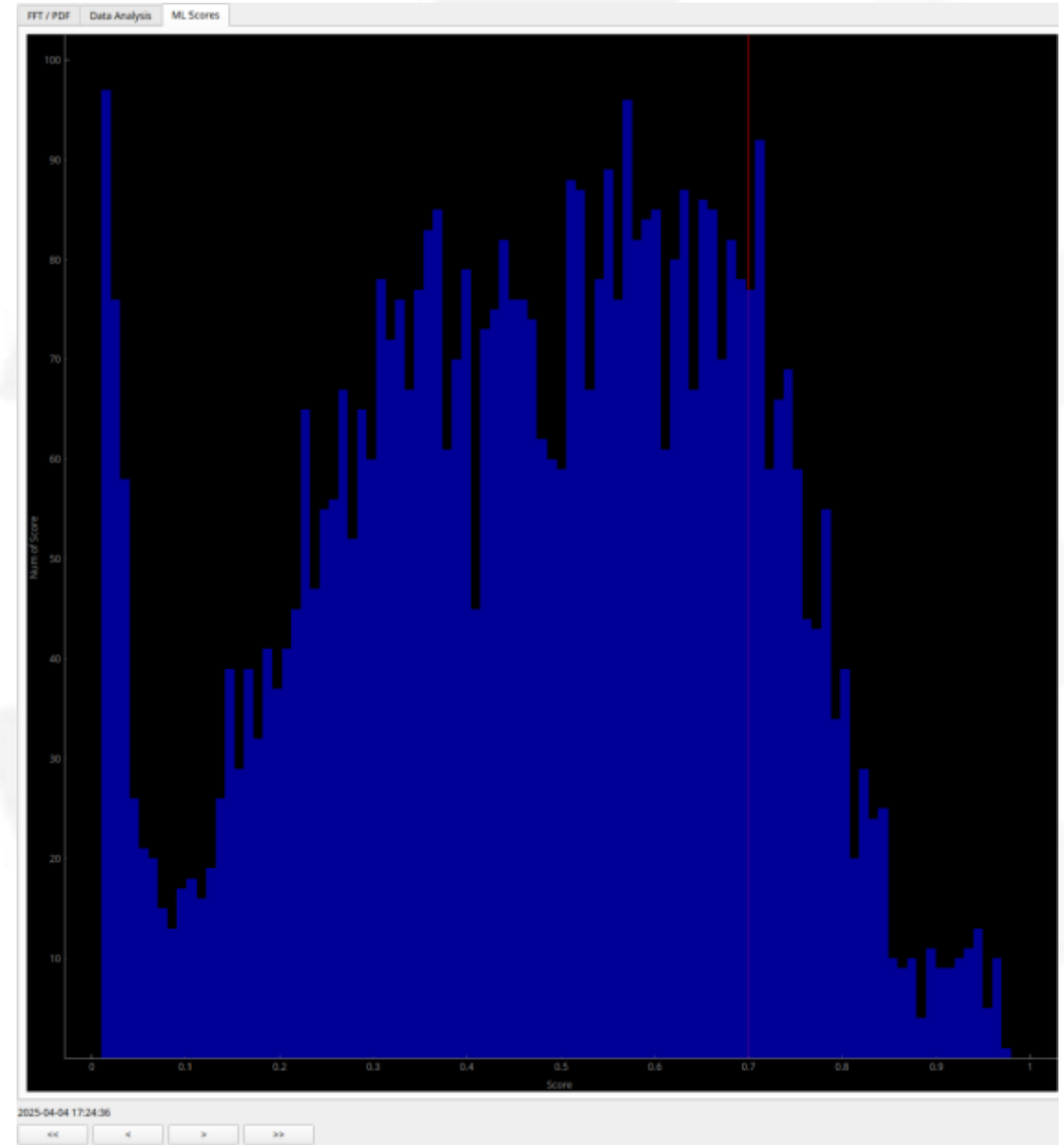
Unfortunately, some
sensitivity to intensity of
stored beam, seen below 250
mA



Beam anomaly detection - **Online application**

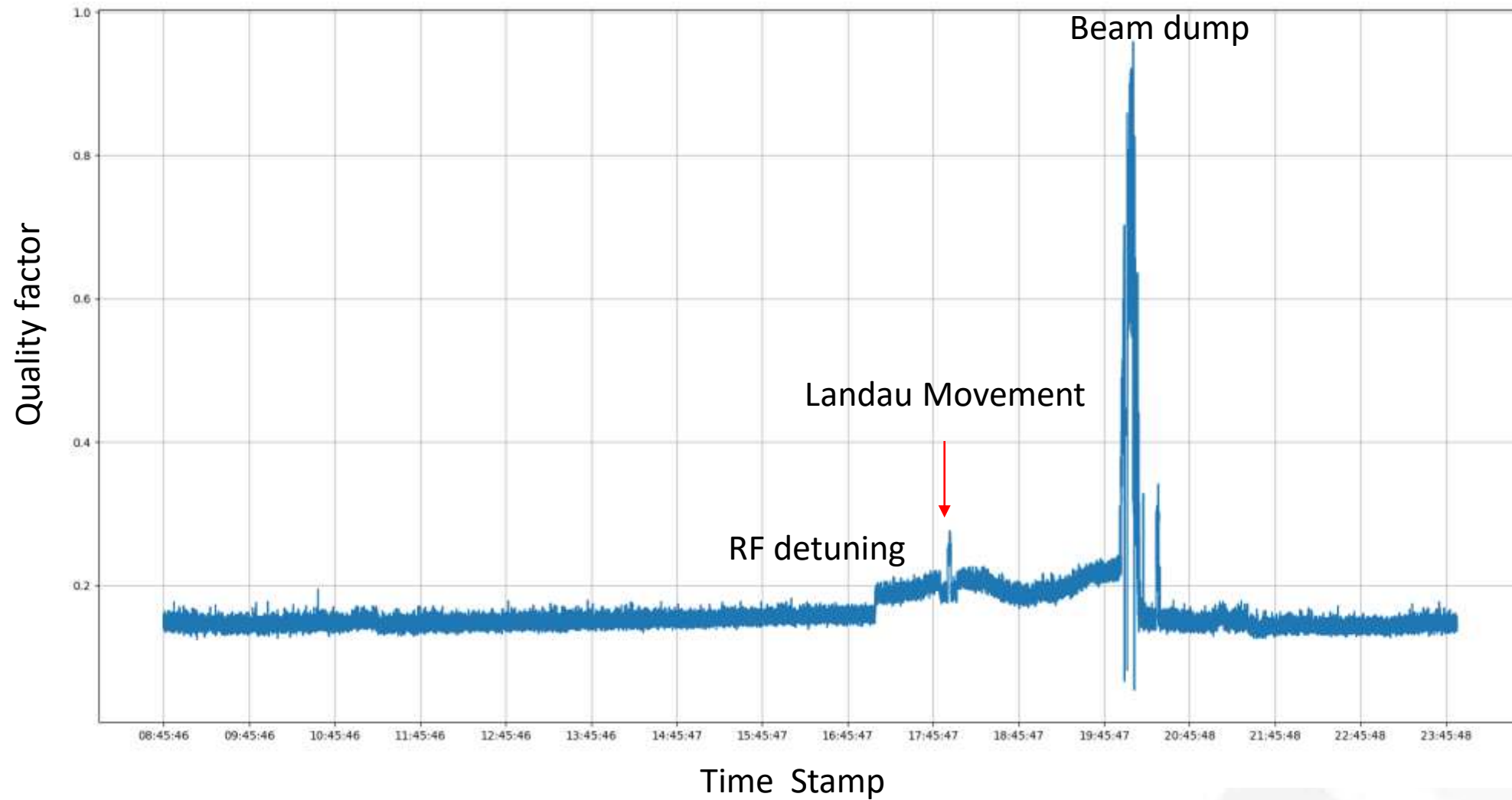


The ML score distribution follows the beam profile with a clear spike around 0, representing the electrons traveling at nominal momentum. The rest corresponds to beam spread (3.4 % RF momentum acceptance, halo etc..)



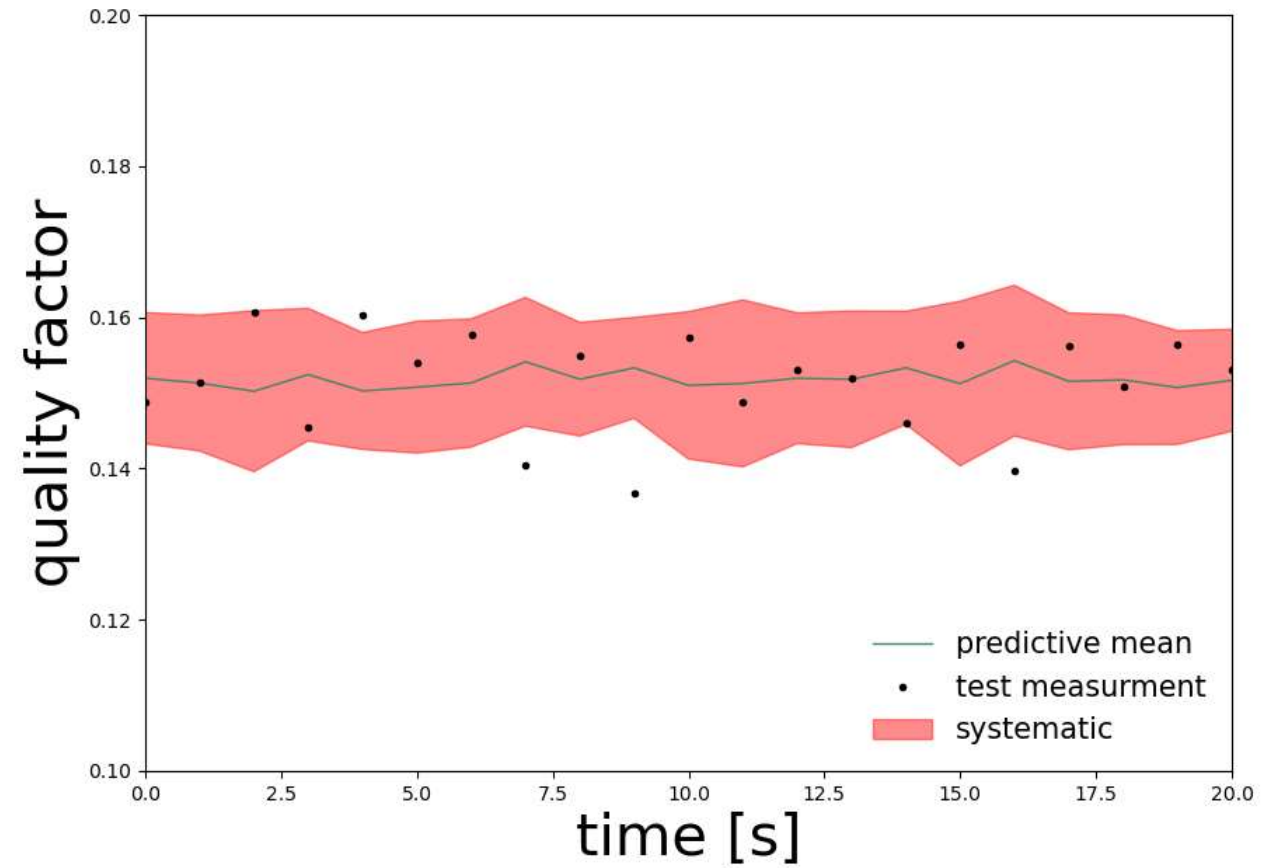
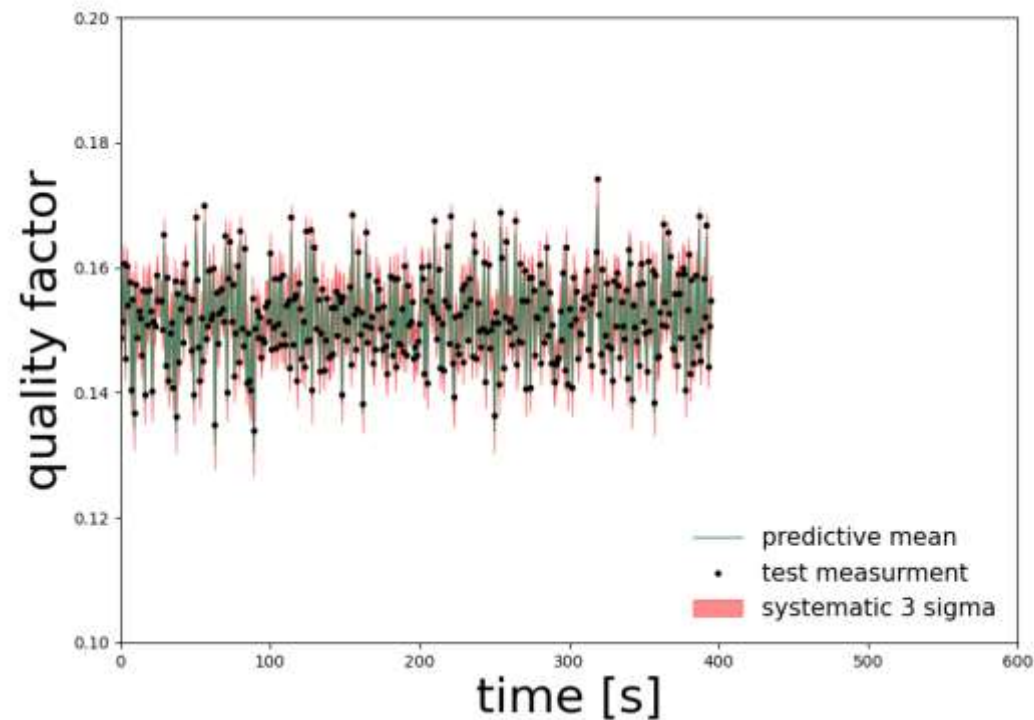
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Beam anomaly detection - **Online application**



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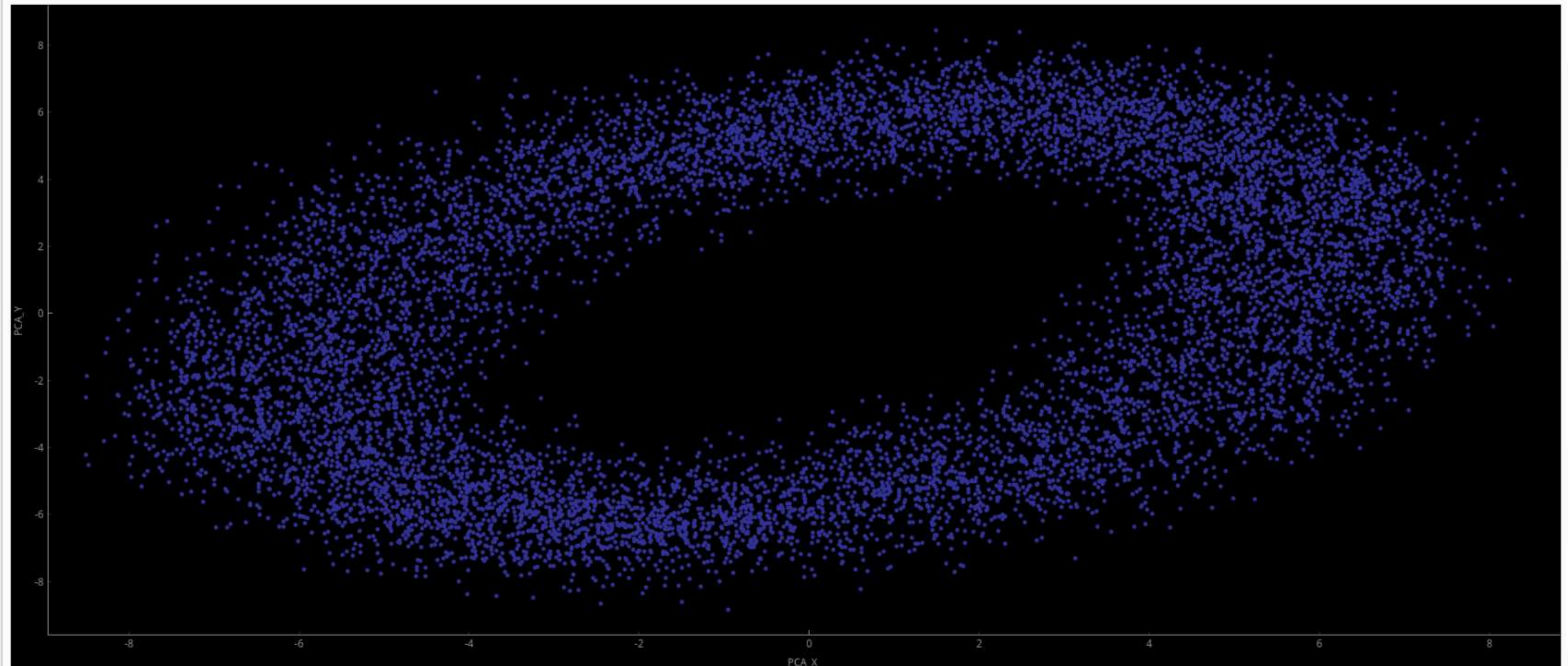
- Online beam quality prediction 400 seconds ahead
- Prediction done by a single layer fed forward net designed in pyTorch (based on Macarov chain Montecarlo)



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Beam anomaly detection - **Principal Component Analysis** (good case)

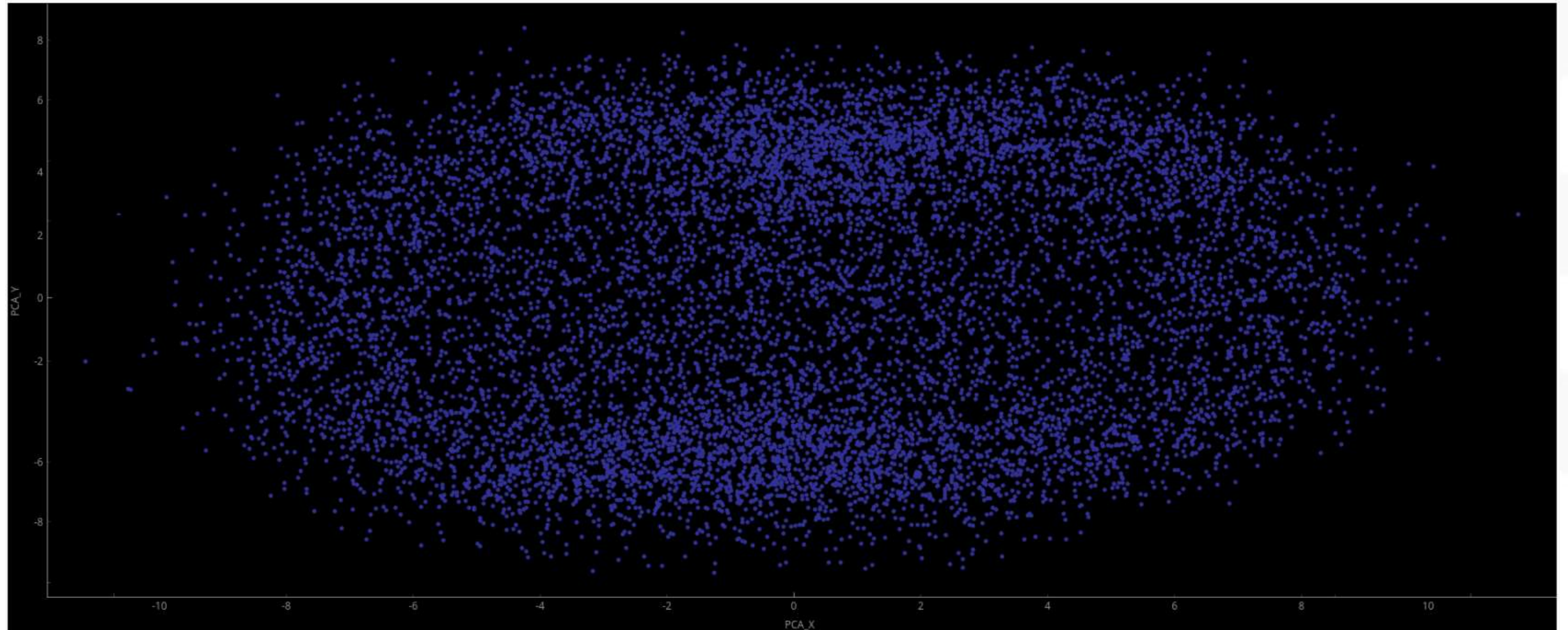
Tuned RF field, analysis done on first BPM after RF cavities



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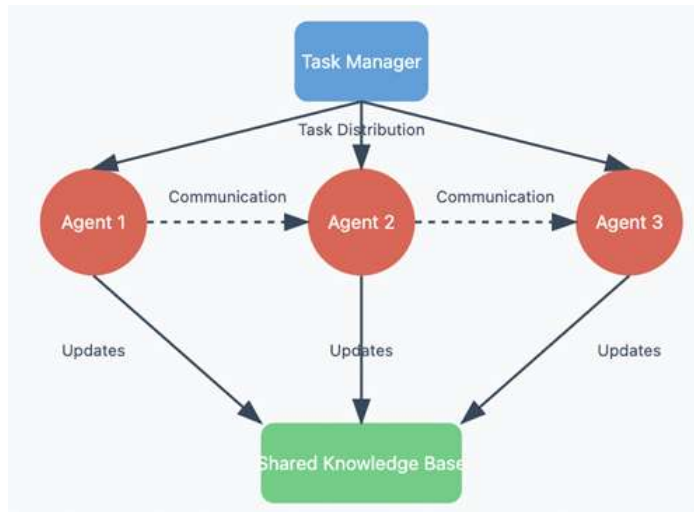
Beam anomaly detection - **Principal Component Analysis** (bad case)

Detuning of RF filed in CAV 1 and CAV 2 (handeld by nutaq)



Machine learning in Synchrotron- plans for the future

- Sector by sector diagnostic
- Fast data aquisition for nutaq (RF tuning and general diagnostic)
- Fast images acquisition for two parasitic optical diagnostic beamlines (PINHOLE, LUMOS)
- Combining all ML output providing a multi agent system



<https://www.digital-alpha.com/ai-trading-financial-markets/>

Thank you