

BPM buttons

non linear beam response meets machine learning

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Introduction

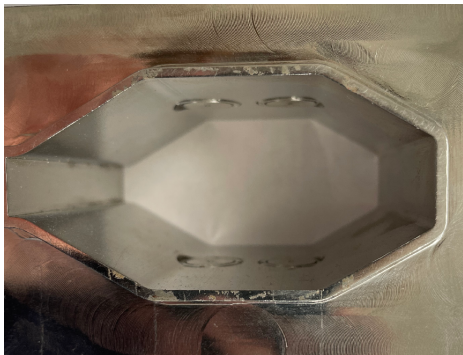
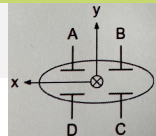


Figure: standard BPM implemented in BESSY II

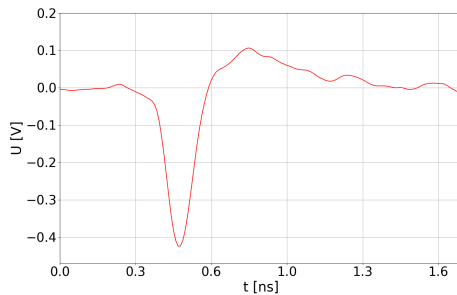
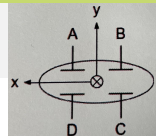


Figure: voltages over time, single bunch signal

Motivation



horizontal scale factor [18.567] mm, vertical scale factor [13.595] mm.

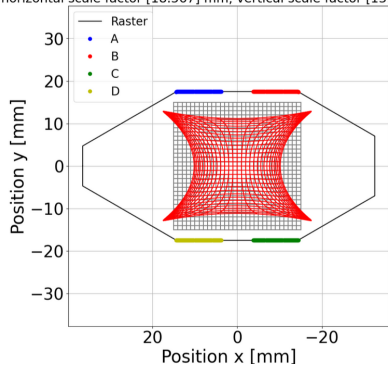


Figure: non-linearity in geometry of BESSY II

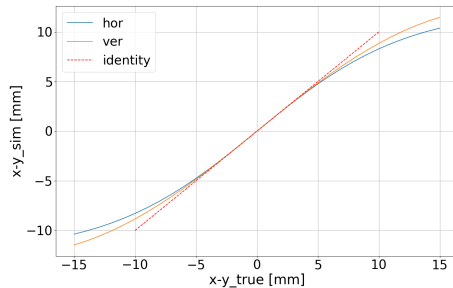


Figure: sensitivity and beam displacement

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Simulation

Principle

- parametrize **geometry** of BPMs & buttons
- simulate **beam positions** in geometry through raster
- calculate **induced charge** for beam positions on chamber wall of geometry
 \Rightarrow boundary element method (**BEM**) [1]
- calculate beam position **backwards** from induced charges on ABCD

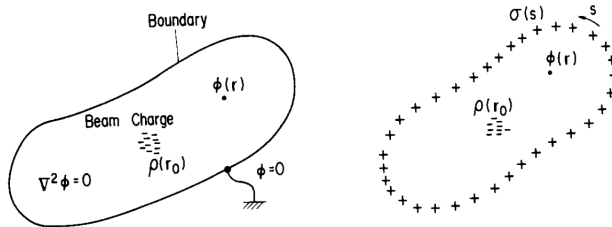


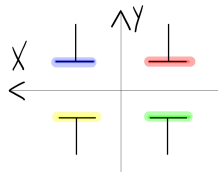
Figure: Reference [1]

Code Conversion

Original Matlab code [2] \Rightarrow Python code

What's included?

- calculation of **induced charge**
 \Rightarrow BEM method
- beam position raster & non linear solution
- Differences over Sum (**DoS**) method
 \Rightarrow applied method to calculate XY-beam-position from ABCD-signals
- calculation of non linear **k-factor**
 \Rightarrow scale factor is change of beam position divided by change of x or y around centres, needs to be applied to DoS solution
- **Newton Raphson Algorithm**
 \Rightarrow iteration method to find XY-position from ABCD signals !time consuming!



The diagram shows a coordinate system with X and Y axes. Four electrodes are positioned around the center: top-left (blue), top-right (red), bottom-left (yellow), and bottom-right (green). Each electrode has a vertical line extending from it, representing the beam's path. To the right of the diagram, the equations for the induced charges are given:

$$X = k_x \frac{(Q_A + Q_D) - (Q_B + Q_C)}{Q_A + Q_B + Q_C + Q_D}$$

$$Y = k_y \frac{(Q_A + Q_B) - (Q_D + Q_C)}{Q_A + Q_B + Q_C + Q_D}$$

The charges are labeled with subscripts A, B, C, and D, corresponding to the four quadrants.

Code Conversion: Matlab to Python



Most important changes

- conversion to **SI** units
⇒ mm → m
- 18 **geometries** included
⇒ e.g. round, rectangular, two buttoned
- **axis** transformation
⇒ according to BESSY II convention

Small Bug

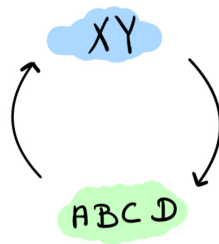
- BEM method → calculation of induced charge
⇒ matrix of Green's functions solving the 2D electrostatic Laplace equation:

$$G_{ij} = \int_{\Gamma_i} \ln\left(\frac{1}{|r_i - r_j|}\right) dr_j = s \cdot \left(1 - \ln\left(\frac{s}{2}\right)\right); s = \text{segments width}$$

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Idea

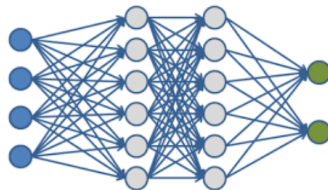
- train **neural network** how to solve
XY-beam-position to ABCD-button-signal
⇒ well known & easy calculation
- generate **training data** with simulation code
- goal: neuronal network learns **inverse** problem
⇒ the inversion we use is a good **approximation** for **small** beam displacements
⇒ we want a **fast** solution for small & large amplitudes



Some Numbers

Example model:

- time of training : $57265.4s \approx 7h$ on CPU
- training data size : $5.2GB$
- model size : $4.6MB$ (not minimized)
- model architecture : 4 layers : $[1000, 700, 500, 300]$
- CPU * times for ML prediction : $\approx 80ms$ for measured dataset **
 \implies CPU * times for Newton Raphson : $\approx 7s$ for measured dataset **
* *13th Gen Intel(R) Core(TM) i7-1365U, 12 threads, 5.2 GHz*
** ≈ 7800 ABCD signals



Outcomes

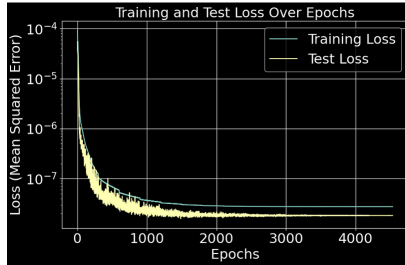


Figure: behaviour of training and testing data during model development

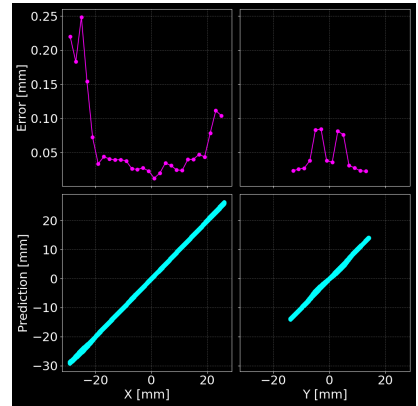
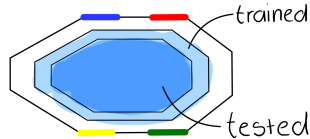


Figure: performance

Outcomes

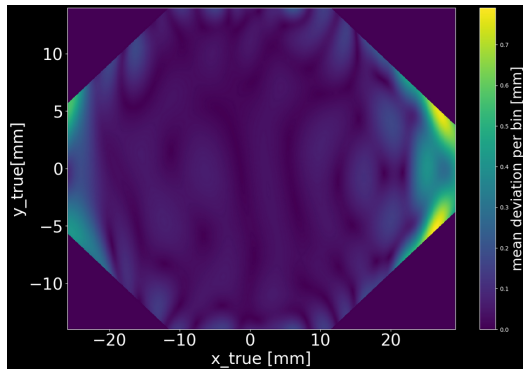


Figure: 2D model performance

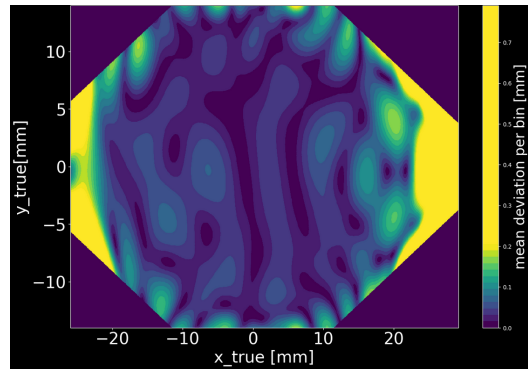
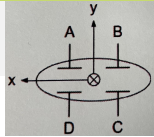


Figure: 2D performance (dicrete area)

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Measurement



measured data of huge beam position shifts by changing masterclock frequency
 $\alpha \approx 7 \cdot 10^{-4}$, dispersion $\approx 0.5\text{m}$

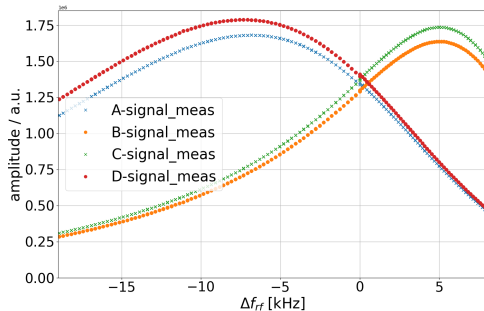


Figure: signals on BPM buttons

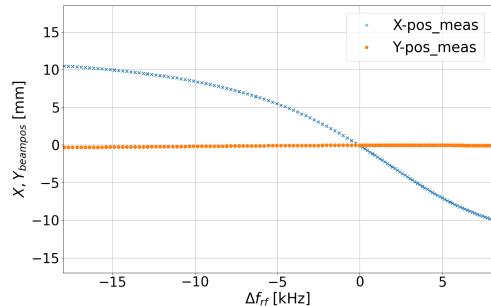


Figure: XY positions

Evaluate model

evaluation in BESSY II model [3] in comparison to measurement results:

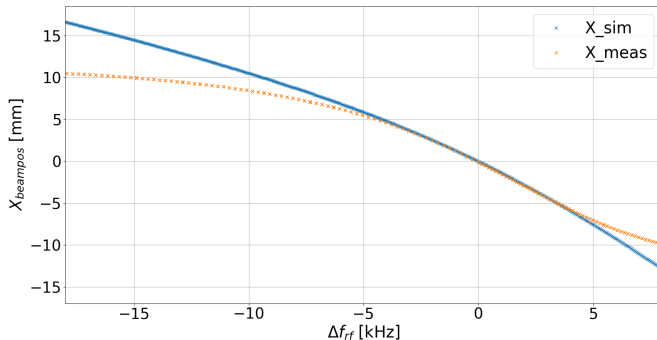
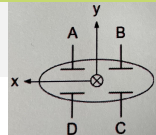


Figure: X positions comparison

Comparison to Newton Raphson

application of measurement data on Newton Raphson algorithm:

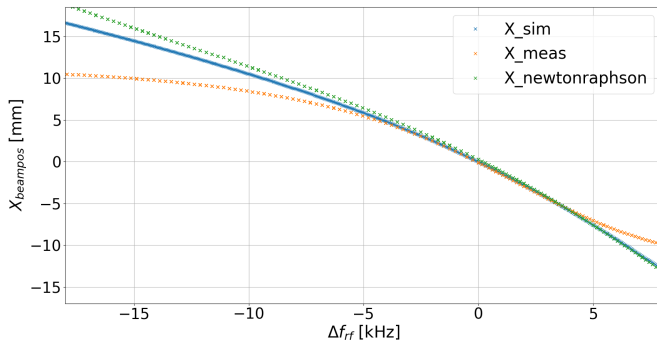
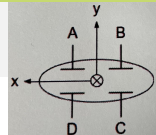


Figure: X positions comparison

Comparison to ML model

application of measurement data on ML model:

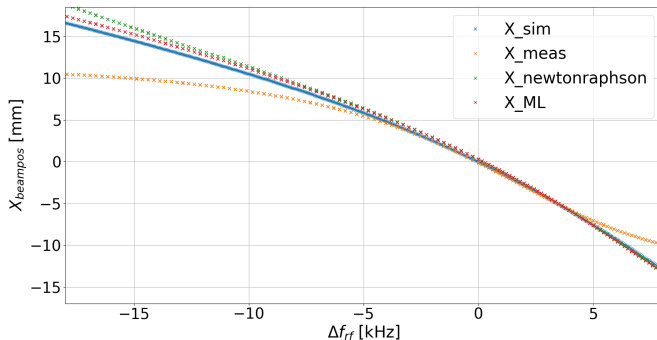
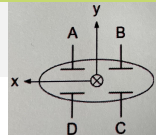


Figure: X positions comparison

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- convert python code into **python package**
⇒ loadable & easy to apply
- implementation of different **BPM button** geometries
- apply **beam current** to ML model
⇒ atm needs to be fixed by **norm factor** on ABCD
- apply ML model to **FPGA** system

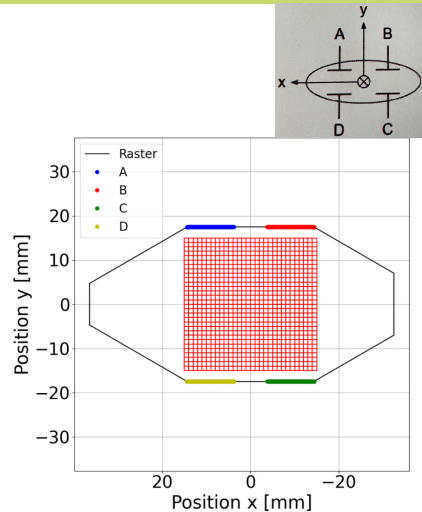




Figure: HZB, BESSY II [4]

References

- [1] Shintake et al., "Sensitivity calculation of beam position monitor using boundary element method", 1987, journal NIM-A, volume 254, pp. 146-150
- [2] A. Olmos, F. Perez, and G. Rehm, "Matlab Code for BPM Button Geometry Computation", in Proc. DIPAC'07, Venice, Italy, May 2007, paper TUPC19, pp. 186-188
- [3] Pyat documentation: <https://atcollab.github.io/at/p/index.html>
- [4] Dirk Laubner https://www.helmholtz-berlin.de/forschung/quellen/bessy/index_de.html

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