# SuperKEKB positron beam tuning using ML

Takuya Natsui KEK

## Back ground

KEK positron/electron injector

## Accelerator at KEK Tsukuba





## Beam timing and beam mode switching



The interval between beams is 20 msec (50 Hz). The time between beams is used for PFN charging, and the other pulsed devices are at idle. A different beam mode can be launched every 20 msec. (HER e-. LER e+, PF e-, PF-AR e-) Pulse magnet, RF phase, etc. are switched every 20 msec. There appear to be multiple injectors from each ring. It is called Virtual Accelerator







## So, beam tuning has become increasingly complex in recent years

In the fall of 2023, large pulse magnets have been installed not only in the 3-5 sector but also in the Jarc area, allowing independent adjustment of each beam mode. Independent tuning of each mode is important for SuperKEKB, which requires beams with large charge and low emittance.

Thus, more and more tuning parameters are being increased. However, since time and manpower remain the same, some new mechanism was needed to extract machine performance.



## Automatic tuning using machine learning



Large pulse magnets were also introduced upstream and downstream of J-arc. Although each beam can now be matched independently, the degree of freedom has increased, and the degree of difficulty of adjustment has also risen.

## Principle of Automatic tuning

Bayesian Optimization and Downhill simplex method. Automatic Tuning of Accelerator Operation as a Minimization Problem.

What is machine tuning? It is "to change the parameters of the equipment to bring it closer to a better state". It is important to consider the parameters and the resulting values as a multivariable function for automatic tuning. For example, if the current value I[A] of the magnet is "x" and f(x) is the inverse of the beam charge Q[nC], the problem is to find x where f(x) is the minimum.



parameters that require an optimization algorithm for the multivariable function f(x1, x2, x3, ... xn).

> For this type of minimization problem, Bayesian optimization or the Downhill Simplex method (Nelder-Mead method) can be used. (f(x) is unknown)

Defines a numerical value to be minimized by computing the measured value.

### Gaussian Processes and Bayesian Optimization

The Gaussian process is a method for predicting the entire function from a small number of observation points with a distribution of errors. Bayesian optimization is a method that determines the next point to be observed based on the prediction and the error range, and searches for the point of minimum (maximum) value of the function.



Gaussian Process Regressor

X

### Bayesian optimization and downhill simplex method

Bayesian optimization is an algorithm that finds the optimal solution by predicting the entire function considering uncertainty. In contrast, the Downhill simplex method is an algorithm that converges to the optimal solution with a simple move and flip operation that considers the number of search dimensions plus one point.





### Downhill simplex method

Move to anticipate the function of the entire search area without falling into local solutions.

Seek optimum point if smooth function, may fall into local solutions. 11

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#### **Bayesian optimization**

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## Example of Automatic Tuning

#### Operation Panel

I use GPyOpt. It is python library for Bayesian optimization.

Select tuning knob. We can use any variable EPICS records and select minimum and maximum value for tuning range.

Basic setting. Number of iteration and so on.

X

Select optimization method. Bayesian or Downhill.

Operator will select setting file and just push this Start button.



## Example 1



[mm]

Y [mm]







It took about 10 minutes to double the amount of positron charge.

### Example 1



SH\_A1\_S1 (KBP) 146.7°  $\rightarrow$  146.1° (-0.6°) SH\_A1\_S8 (KBP) 48.0°  $\rightarrow$  48.6° (+0.6°) PX\_AT\_22-KBP 0.364A  $\rightarrow$  0.170A (-0.194A) PY\_AT\_22-KBP 1.435A  $\rightarrow$  1.451A (+0.016A) PX\_A1\_M-KBP 0.009A  $\rightarrow$  -0.018A (-0.027A) PY\_A1\_M-KBP -1.137A  $\rightarrow$  -1.176A (-0.039A)

## Example 2

The problem was the large beam loss of positron in the capture section. The transmission was lower than the simulation results, but the cause had not been identified.

It was thought that the Q-magnets and steering magnets would need to be adjusted to improve this situation. However, about 200 parameters had to be adjusted, making it impossible to do so manually.

The parameters were adjusted with automatic tuning from the upstream with divided regions. Parameter tuning was fully automated, resulting in a significant improvement in the transmission of the positron beam.







### Before tuning

File Data Mag BPM Update 2023/10/20 14:34:57 v8.2 Linac KEKB e+ Orbit 2023/10/20 14:34:57 DX 1st RMS: 0.978 DX [mm] 2 Max: 11.887@S8\_DN\_01 n Min: -1.901@SP\_15\_T -2 -4 SX [A] XS 5 -DY 1st 4 RMS: 0.554 DY [mm] 2 Max: 3.662@SP\_R0\_02 Min: -1.033@SP\_26\_1 -2 -4 Positron beam loss, after target S [A] VS Purthind 5 S8\_DN\_01 (QLF1N) -6 . 4 [nC] Ū L DX(1st): 11.887 mm 9 O DX(2nd): 0.000 mm 5 DY(1st): -0.074 mm 0 DY(2nd): 0.000 mm SP\_C7\_4 SP\_C6\_4 SP\_C6\_4 SP\_C4\_4 SP\_C3\_4 SP\_C2\_4 SP\_C2\_4 88\_15\_5 SP\_12\_4 SP\_11\_4 SP\_08\_4 Q(1st): 3.144 nC Q(2nd): -100.000 nC Beam Gate --- FC 15 LTR BS RTL BS Bucket Sel-Bunch e+/e-DR pulse Beam Rep ON 18.498 kV ACC 2nd n01: Close n02: Open 55.33 [%] 5.000 5.000 [Hz] 100.000 [%] Open Open ON s01: Open 1st Show Range DX 5 - DY 5 - Qe- 13 - Qe+ 7 -Replot Cur-Ref Gold Ave10 2023/06/06 03:28:45 Set Ref 🔳 Cur 🔳 Sector Bunch -Sigma-KBE PFE QFE ARE JBE JBP RFE SFE ZRE 👅 1st 🔄 2nd **A** BT visible В 6

🗆 chg th 🛛 A 🖂 SP AT 0 🖂 1st 🔤 0.1 [nC] 🗆 P.H 🔲 conti 🛛 300 💴 SP\_DN\_21 (QLF5N) : DX=[ 0.00, 0.00] DY=[ 0.00, 0.00] Qe+=[ -100.00, -100.00] resize 📷

### After tuning



## Design and tuned Q magnet field in positron beam line



- The error due to energy is estimated to be about 10%. Magnetic field errors are usually smaller than that.
- Similar to the design in some areas, but values differ by more than 20%, especially near Target.
- The large energy error near the target is thought to be a contributing factor.
- Due to misalignment effects, the Q magnet acts as a steering magnet?

Although the adjustment itself was made with the improvement of charge quantity as an index, a trend of discrepancy with the design became visible as a result.

清宮氏提供データ

Automatic tuning significantly improves positron charge

Achieved a positron charge (before damping BT) almost equal to the calculated value. (2023/10/25)



## Summary

- In the KEK injector linac, there are many tuning parameters for four-ring injection, and the number of tuning knobs has been increasing in recent years due to the increase in the number of pulsed magnets.
- Automatic tuning was introduced to ensure sufficient tuning in a small amount of manpower and time.
- We use Bayesian optimization and Downhill simplex method as optimization algorithms.
- In accelerator facilities where many adjustment knobs are controlled by a control system such as EPICS, such automatic adjustment is very useful.
- In fact, the automatic tuning contributed significantly to the increase in positron production.
- In the future, we would like to make the automatic tuning easier to use and apply it to routine beam stabilization.