# **Automatic Optimization of X-ray Free-Electron Laser at SACLA**

Hirokazu Maesaka<sup>\*,1,2</sup>, Eito Iwai<sup>2,1</sup>, Ichiro Inoue<sup>1</sup>

1: RIKEN SPring-8 Center, 2: Japan Synchrotron Radiation Research Institute (JASRI)

**3rd ICFA Beam Dynamics Mini-Workshop on** Machine Learning Applications for Particle Accelerators osted by Brookhaven National Laboratory ovember 1-4, 2022



### Introduction

- The brightness of an X-ray free-electron laser (XFEL) should be maximized for each experimental user by fine-tuning of accelerator components.
- An optimizer based on Gaussian process regression (GPR) has been successfully applied to XFEL facilities [1].
- We have developed a GPR-based optimizer for the XFEL facility, SACLA.

### SACLA

### Main Linac [2]

#### • Beam energy: 8 GeV maximum

• XFEL photon energy: 4 – 15 keV (BL2, BL3)

### Key Technologies

• Low-emittance CeB6 thermionic electron gun.

#### Second Linac

- Beam energy: 0.8 GeV maximum
- XFEL Photon Energy: 20 150 eV (BL1)

# **New High-Resolution Inline Spectrometer**

- To include spectral information to the objective function, a new high-resolution inline spectrometer (NIspec) was developed and installed [6].
- Halo part of XFEL is scattered by a diamond microcrystals filled in thin capillary and diffracted X-rays are detected by an X-ray image sensor (MPCCD).
- The resolution is a few eV for 10 keV XFEL, which is significantly narrower than the typical XFEL bandwidth of ~40 eV FWHM.





# **Optimizer Design**

- We implemented the GPR based on scikit-learn [3], GPyTorch [4], and BoTorch [5].
- The GP kernel is the convolution of a radial basis function (RBF), constant kernel, and white noise kernel.

 $K(\mathbf{x}_i, \mathbf{x}_j) = k \times \text{RBF}(\mathbf{x}_i, \mathbf{x}_j) + \sigma \delta(\mathbf{x}_i - \mathbf{x}_j)$ 

- The hyperparameters are the factor, k, noise factor,  $\sigma$ , and the length scale of RBF.
- We can set upper and lower limits for each parameter.
- The optimizer can handle up to 32 parameters at a time.
- The optimizer uses the maximum point of the expected improvement,  $\alpha_{\rm EI}$ , to determine the parameters for the next trial.

$$\alpha_{\rm EI}(\mathbf{x}^*) = \int_{-\infty}^{\infty} (y^* - y_{\rm max}) p(y^* | \mathbf{x}^*, \mathbf{y}, X, \boldsymbol{\theta}) dy^*$$

### **Optimization of Spectral Brightness**

The spectrum from the NIspec is fitted by Cauchy distribution. • The spectral brightness,  $B_{avg}$ , is defined by:

$$B_{\rm avg} = \frac{P}{\sigma_{\rm avg}} \qquad \qquad \sigma_{\rm avg} = \sqrt{\sigma_{\rm width}^2 + \sigma_{\rm med}^2}$$

*P*: XFEL pulse energy.

- $\sigma_{\text{width}}$ : FWHM of the Cauchy distribution.
- $\sigma_{\rm med}$ : FWHM of the median/peak photon energy of multiple (~90) shots.
- The spectral brightness was successfully optimized by our GP optimizer.
- The XFEL pulse energy was optimized first, then the spectral brightness was optimized in the case below.





 $Jy_{\rm max}$ 

**x**\*: the parameter vector to make an inference.  $y^*$ : the value of objective function at  $x^*$ . *X*: the parameter vectors ever tried.

**y**: the obtained data of objective function for *X*.  $y_{max}$ : the maximum value of y.  $\boldsymbol{\theta}$ : hyperparameter vector.

## Tuning knobs

- RF phases of 238 MHz, 476 MHz, L-band (main and correction), S-band, and C-band (main and correction) accelerators. (7 parameters)
- Magnetic lens strengths in the low-energy region. (9 parameters)
- Quadrupole magnet strengths. (62 parameters)
- Steering magnet strenghts. (8 parameters)
- Taper of undulator K-values. (7 parameters)

# **Optimization of XFEL Pulse Energy**



- The optimizer has been routinely utilized since 2020.
- However, XFEL spectral width, spatial profile, etc. are sometimes deteriorated after the optimization of XFEL pulse energy.
- Spectrum, spatial profile, etc. should be implemented to the objective function.

#### References

- [1] J. Duris et al., Phys. Rev. Lett. 124, 124801 (2020).
- [2] T. Ishikawa et al., Nature Photonics 6, 540 (2012).
- [3] F. Pedregosa et al., Scikit-learn: Machine learning in Python, Journal of Machine Learning Research 12, 2825 (2011).
- [4] J. R. Gardner et al., Gpytorch: Blackbox Matrix-Matrix Gaussian Process Inference with GPU Acceleration, Advances in Neural Information Processing Systems **31** (2018).
- [5] M. Balandat et al., BoTorch: A Framework for Efficient Monte-Carlo Bayesian Optimization, Advances in Neural Information Processing Systems **33** (2020).
- [6] I. Inoue et al., J. Synchrotron Rad. 29, 862 (2022).

- Both the photon energy jitter,  $\sigma_{\rm med}$  , and the band width,  $\sigma_{
  m width}$  , were improved. • The width of the averaged spectrum was also made narrower and the spectral peak
- was increased by 1.7 times in this case.
- RF phases were more sensitive to the spectral brightness than the other tuning knobs.
- The spectrum was thought to be mainly affected by the longitudinal phase-space profile (bunch length, energy spread, etc.).
- Thus, the choice of the objective function is quite important for XFEL optimization.

# **Summary**

- We developed a GPR-based optimizer for the XFEL facility, SACLA.
- Although we succeeded in the maximization of the XFEL pulse energy, the XFEL bandwidth sometimes became broader.
- Therefore, a new high-resolution inline spectrometer was developed and installed for optimization of the spectral brightness.
- The spectral brighteness was successfully improved by our optimizer.
- We plan to employ some other XFEL performance indices, such as XFEL spatial profile, to deliver better XFEL radiation.