

# Machine Learning for Longitudinal Phase Space Reconstruction

Preliminary Results

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# Structure

## This Presentation

- Scalar & Spectral Virtual Diagnostics
- Intro to KNN & ANN
- Adaptive Resolution Reconstruction
- A Comparison of Methods
- Deploying to the Control Room
- Where to go from here?

# Longitudinal Phase Space Inference

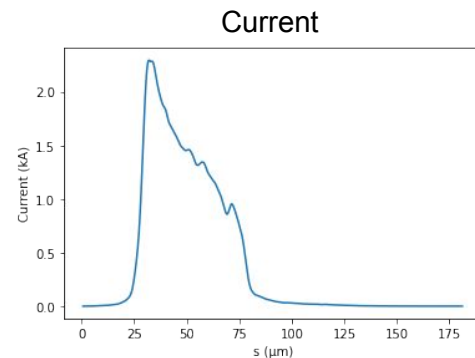
## Initial Approach (Fixed Resolution)

### Scalar Virtual Diagnostic

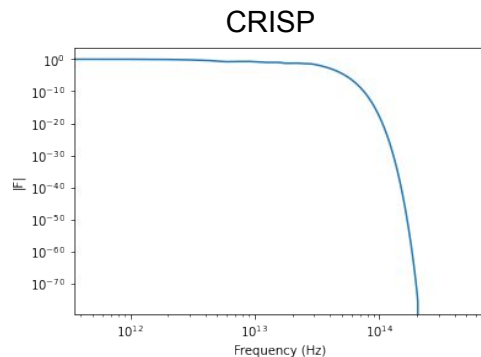
RF Parameters



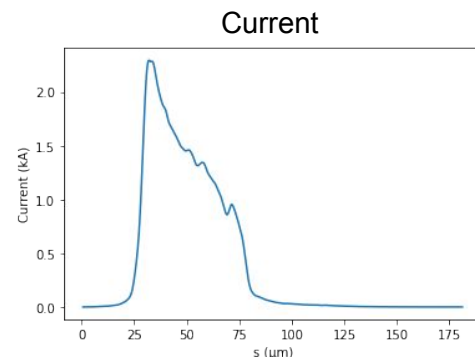
Machine Learning Model



### Spectral Virtual Diagnostic



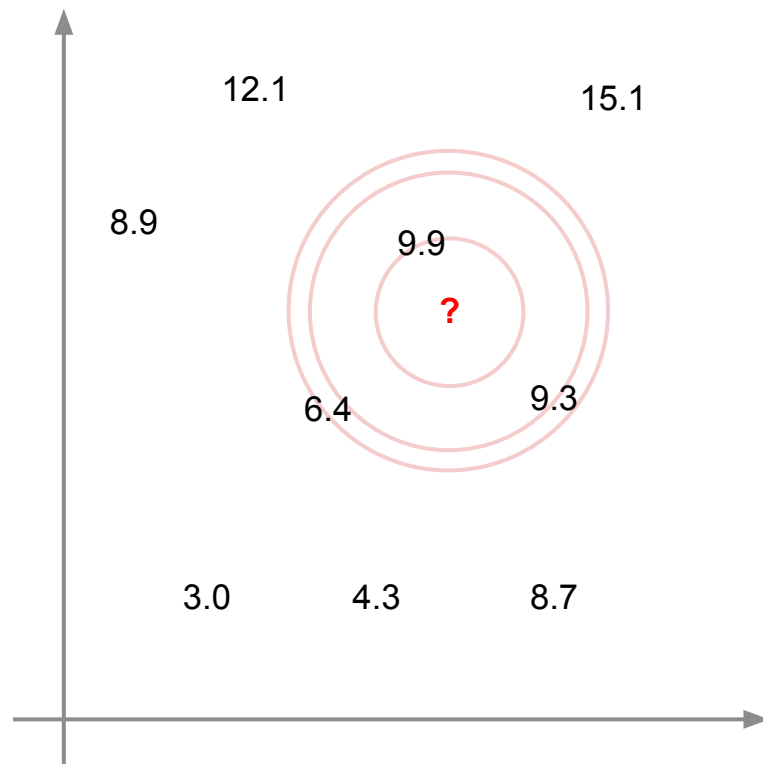
Machine Learning Model



# K-Nearest Neighbour (Regression)

## Brief Introduction

- (Weighted) mean of closest neighbours
- If  $k=1$ , basically a lookup in database
- How to compute distance?
- Interpolation?
- Potentially large database and expensive lookup

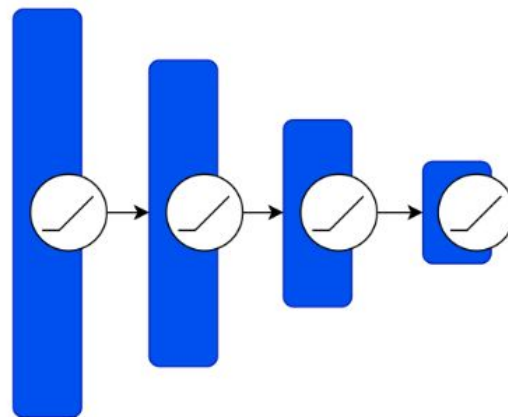
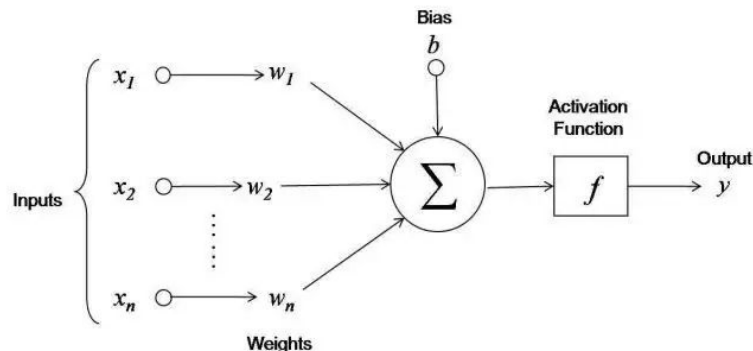


# Artificial Neural Networks

## Brief Introduction

- General purpose function approximator
- Trained via some kind of stochastic gradient descent on gradient of *loss function* over weights
- Basically compress the learned dataset by learning function
- Approximate “correct” interpolation
- Potentially very fast, especially on easily available dedicated hardware.

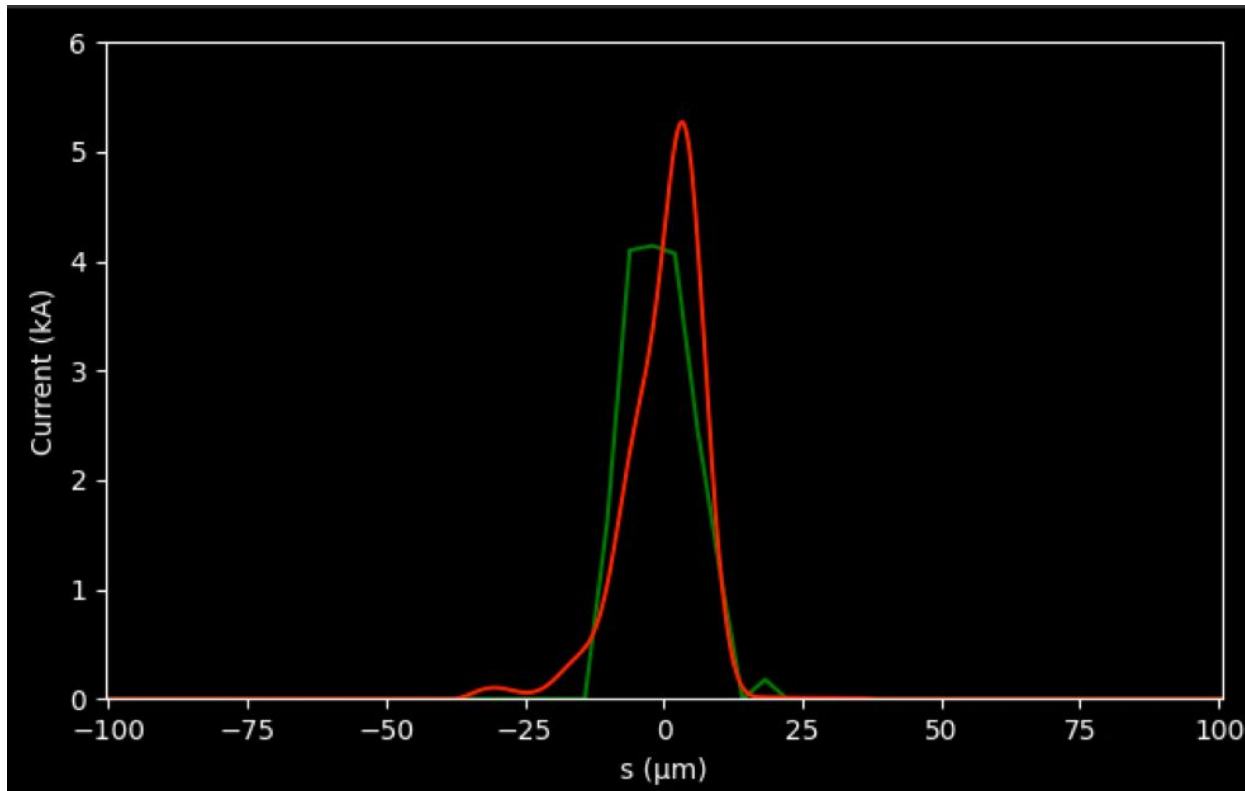
$$\mathbf{y} = \sigma (W \mathbf{x} + \mathbf{b})$$



# Fixed Sampling

## The Problem

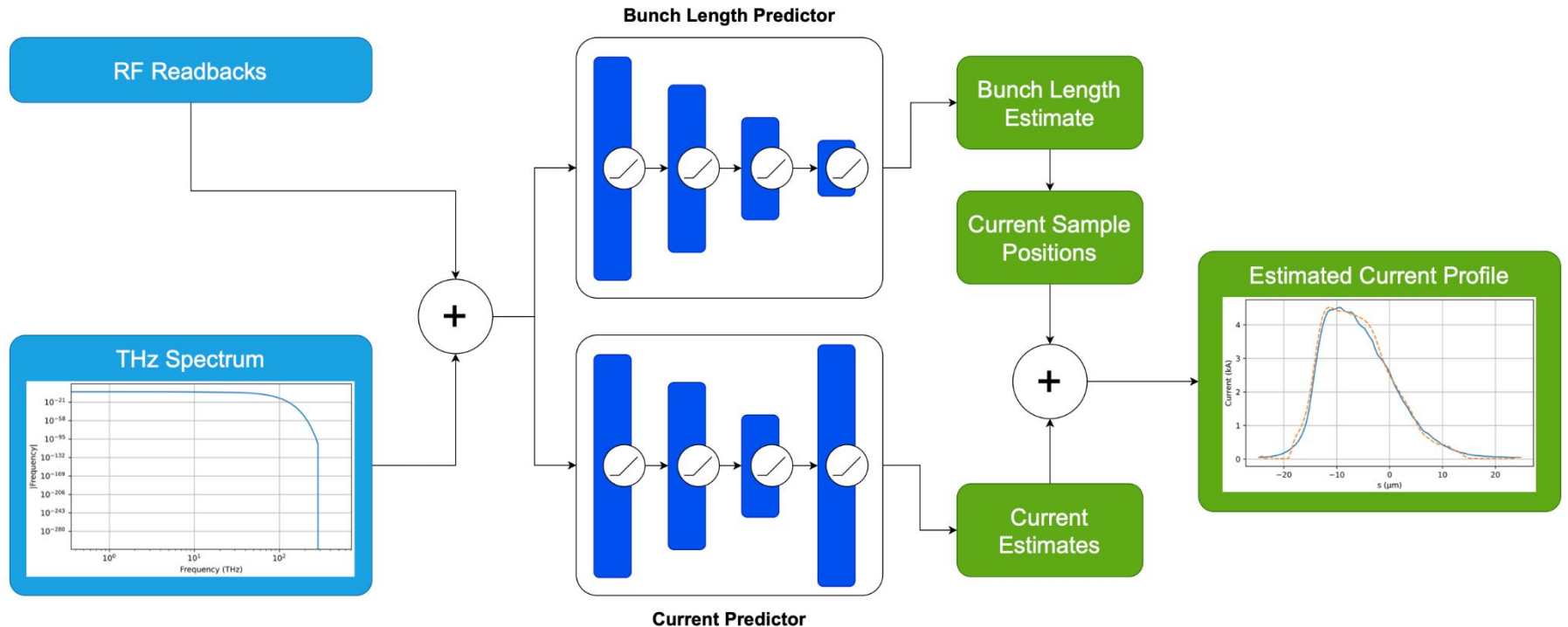
- Range covered needs to be very long for long bunches
- Number of samples needs to be infeasibly high to resolve short bunches



# Improved Approach

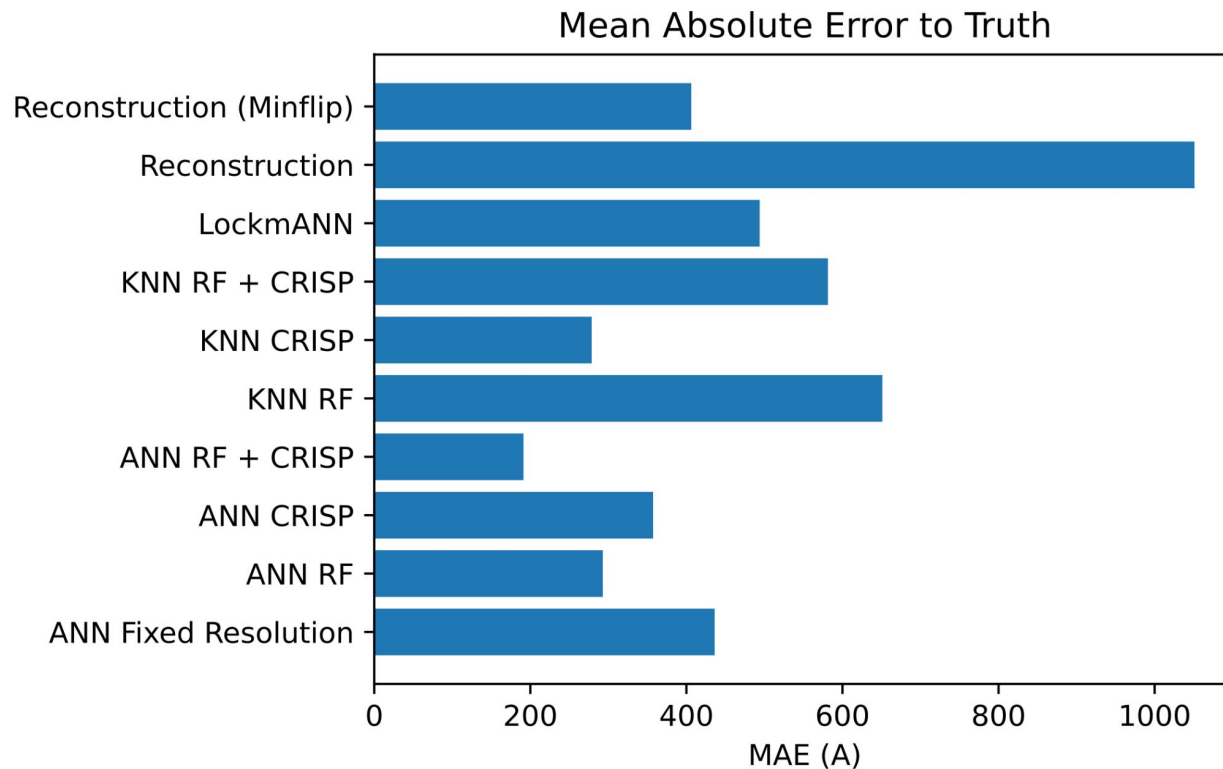
## Adaptive Output Resolution

- Second ML model to predict width of bunch and determine positions of  $n$  longitudinal samples.



# Comparison of Different Methods

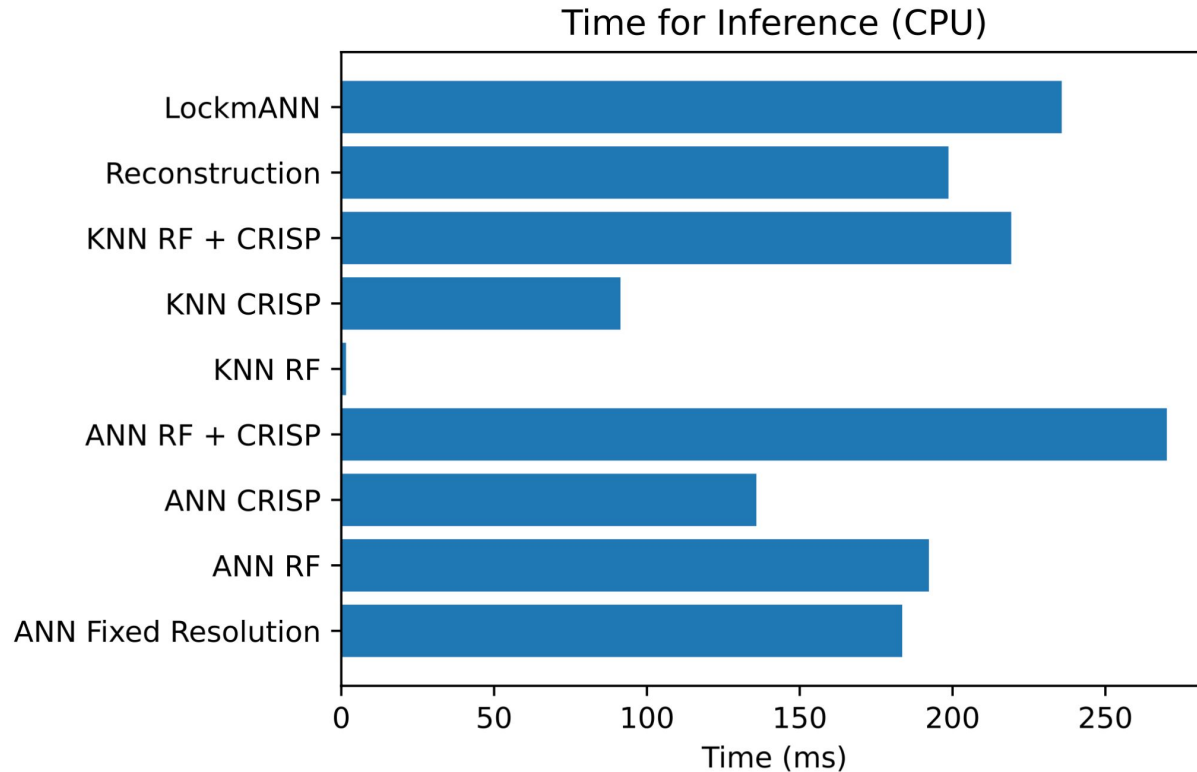
## ANN vs. KNN vs. Reconstruction





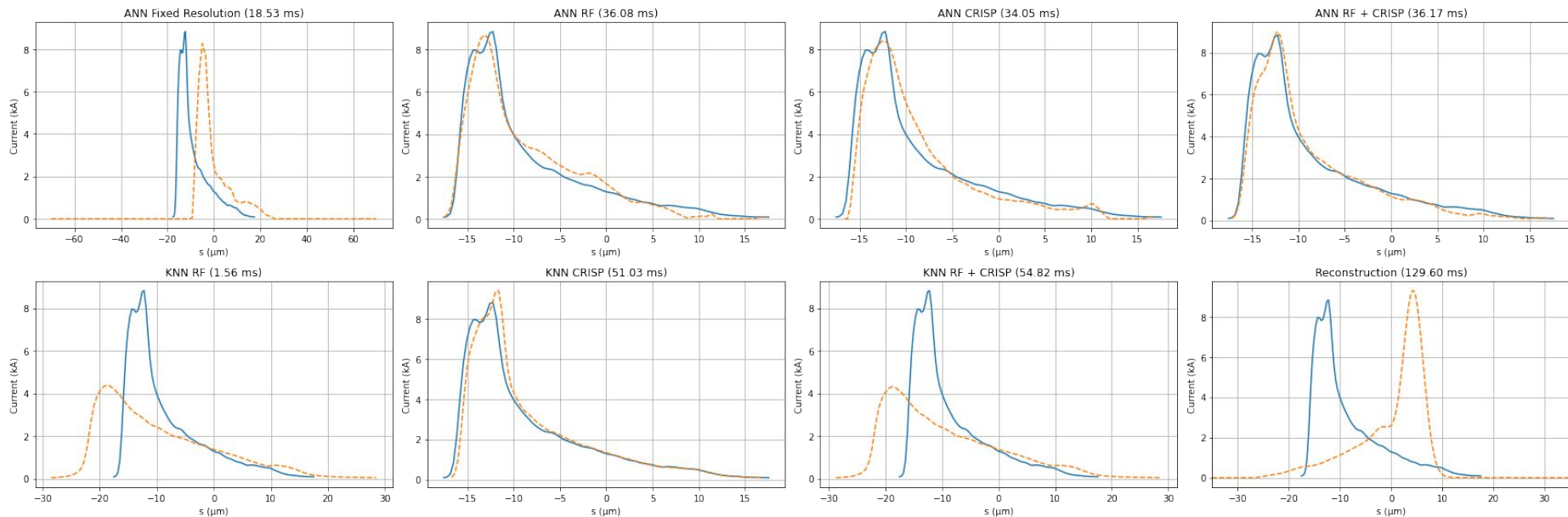
# Comparison of Different Methods

## ANN vs. KNN vs. Reconstruction

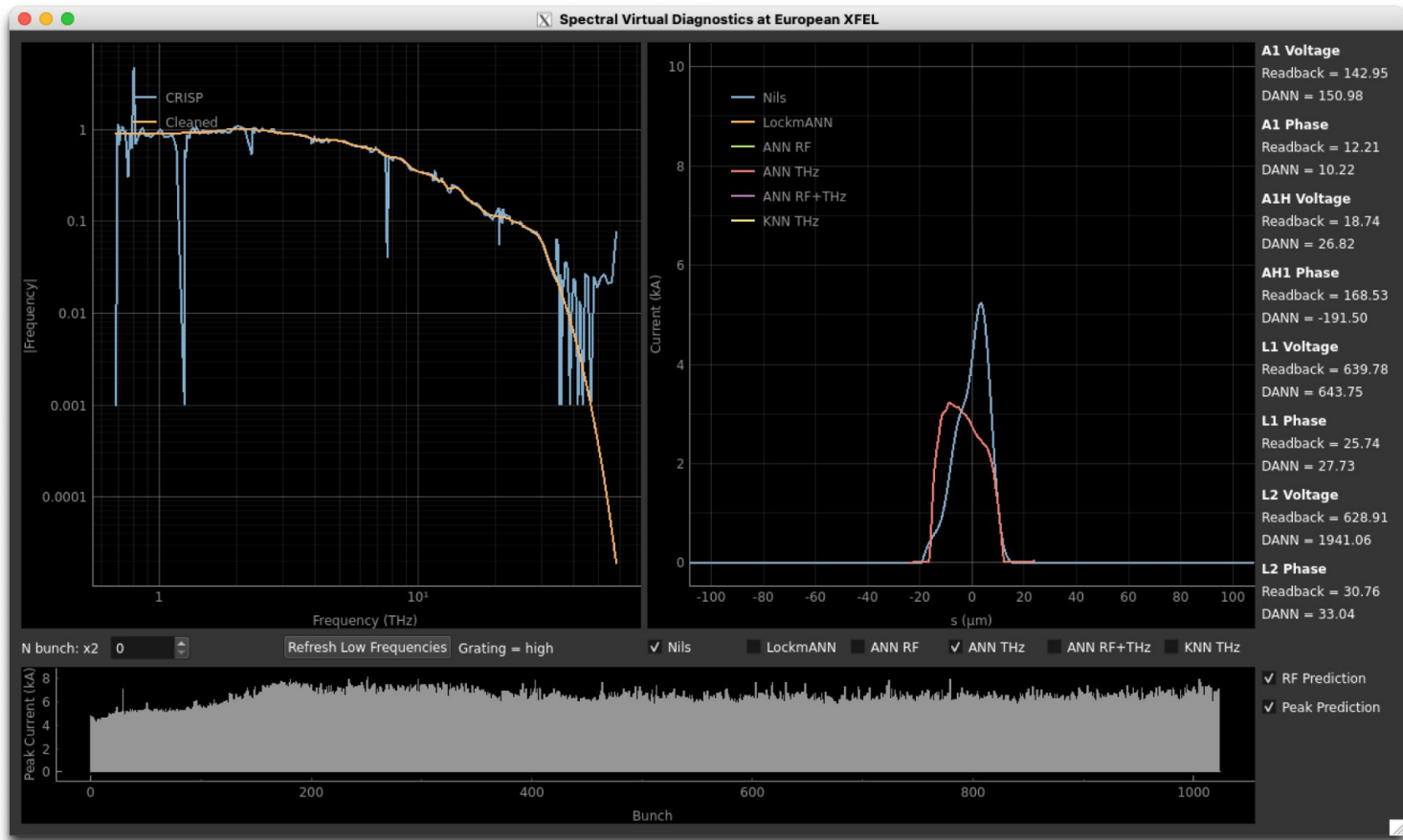


# Comparison of Different Methods

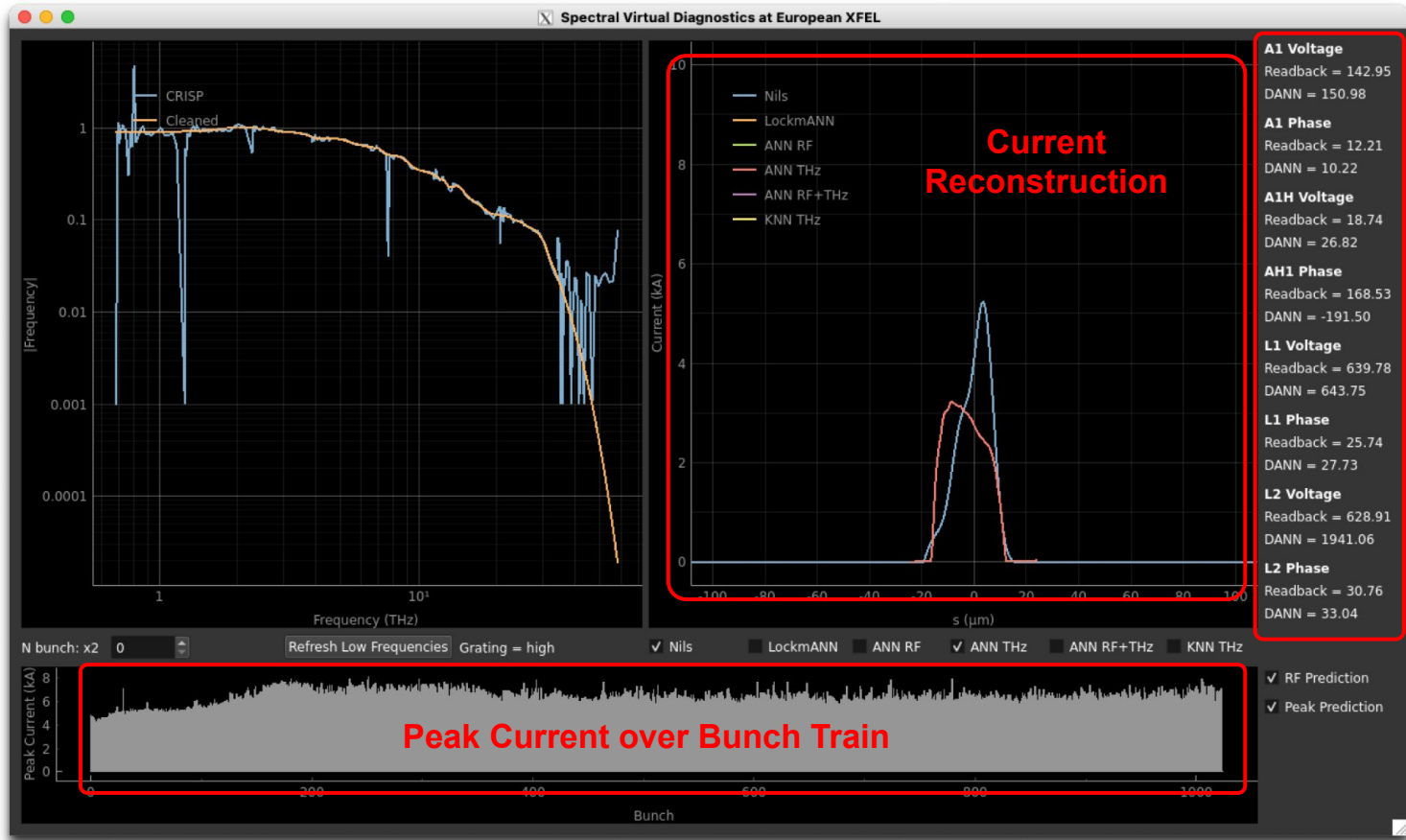
## ANN vs. KNN vs. Reconstruction



# Python Application



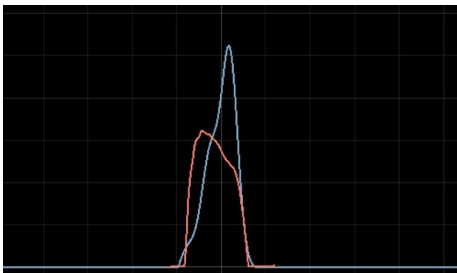
# Python Application



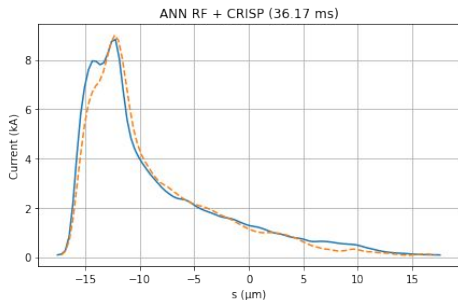
# Where to go from here?

## Future Work

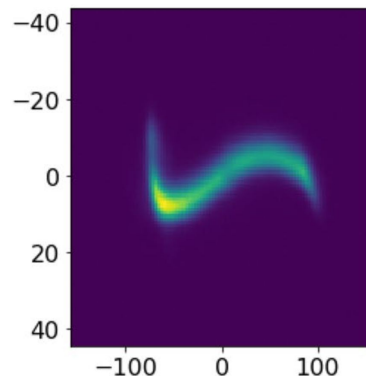
- New dataset to train for better real-world agreement



- More suitable loss function
  - Replace MSE with loss functions from image generation (e.g. perceptual loss, GAN)



- 2D LPS



- Improve model architecture?
  - Share length and current weights
  - CNN? LSTM?
- Deploy to server with better reliability (dependencies)

**Thank you!**

## Contact

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