



Predicting the LINAC3 Beam Current

Summary of work done by BE-ICS

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22.07.2020

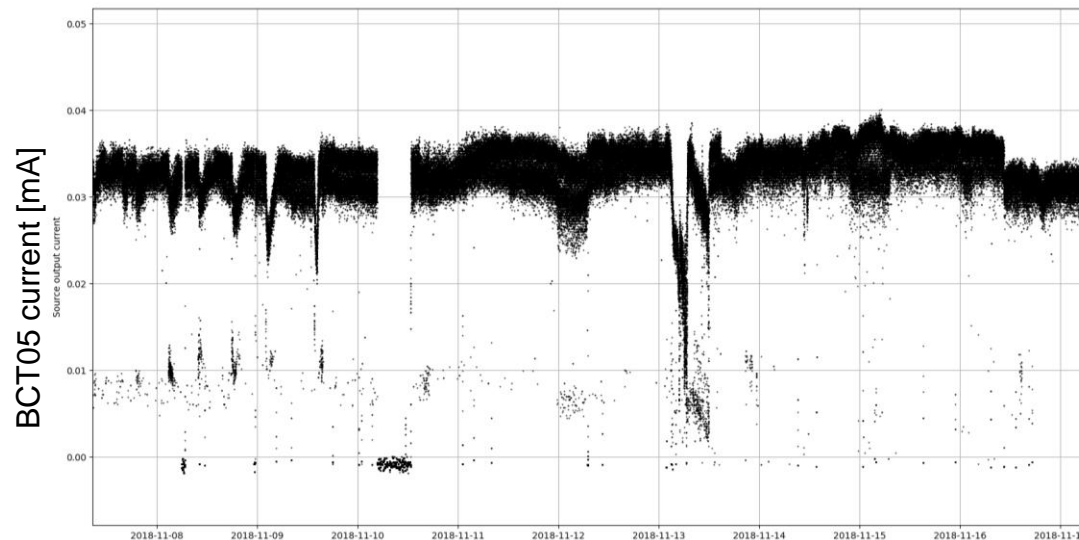
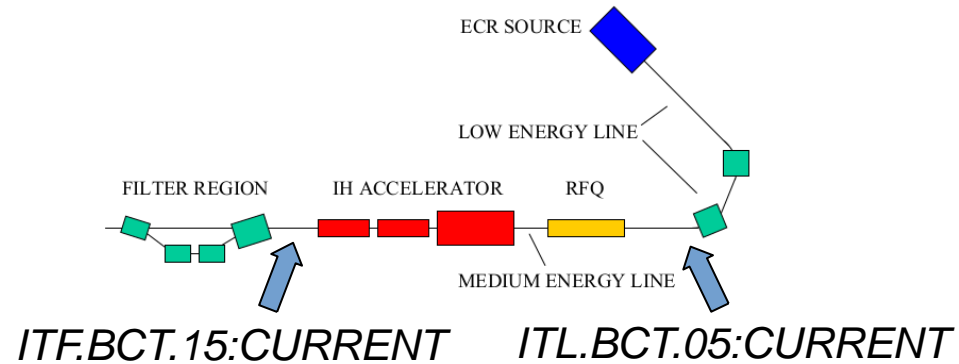


Overview of the Cooperation

- Three year old cooperation with BE-ICS/Siemens
- Develop a Machine Learning Model supporting LINAC3 operation
- Since October: Filip Široký and Marc Bengulescu
- Presented multiple times at ML-coffee (informal CO platform on machine learning)
- **Goal:** Predicting the Beam current of the LINAC3 using Neural Networks

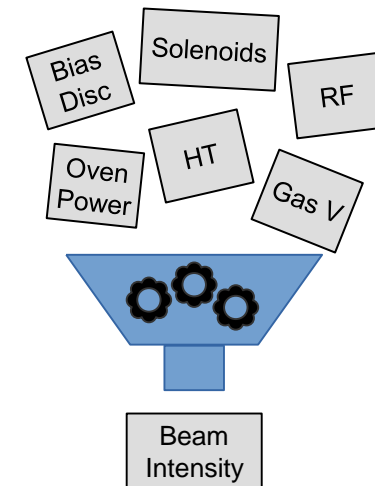
Thank you to Filip and Marc as everything presented here is their work.

Layout of LINAC3 and the objective



Objective:

Using all source Acquisitions, can we predict what the intensity of the beam will be in the near future?

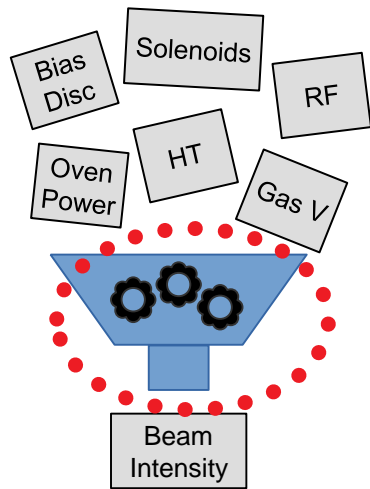


Numerous possible applications

- **During operation:**
 - Automatically correct (small) instabilities
 - Have a pre-warn system that alerts operators if the current is about to degrade
 - Will the beam intensity drop over the next 30min?
 - Are there going to be many bad shots in the next hour?
- **Offline analysis:**
 - Study how certain source settings might influence the current (simulation)
 - Understand what variables cause the beam intensity to change

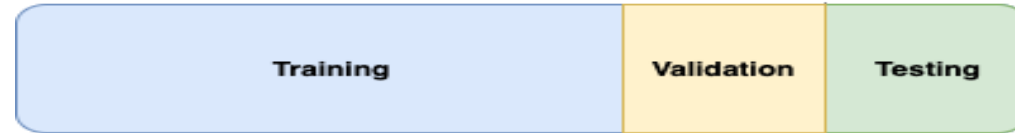
Machine learning crash course

How do you train a (supervised) model?



This is what we call a model

At the beginning, the model knows nothing.

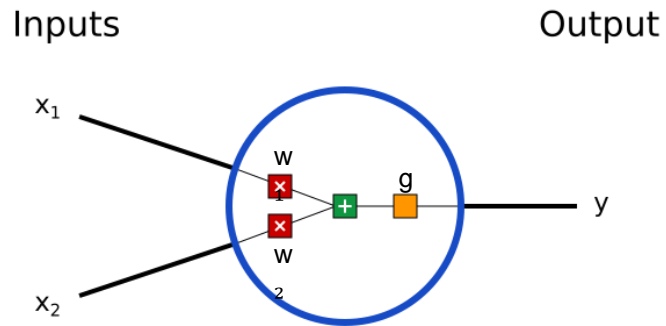


Take all your data and split it up

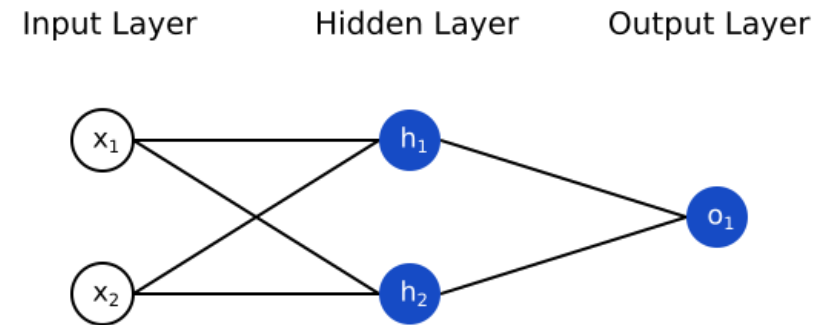
1. Present it a lot of (input, expected output) pairs and adjust model to minimize an error function
1. Measure performance on **unseen** data
1. Test behaviour in real application

Machine learning crash course

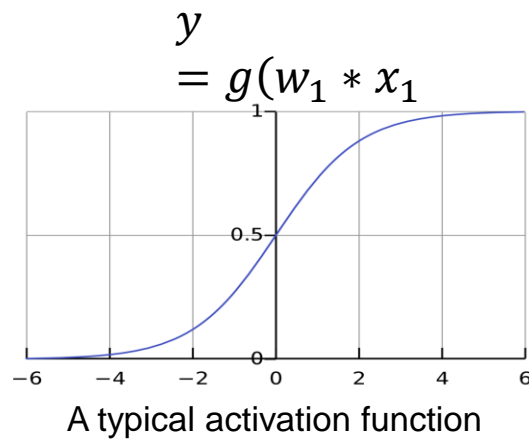
Neural network basics



Building Blocks: Neurons



Combine units to get a neural net

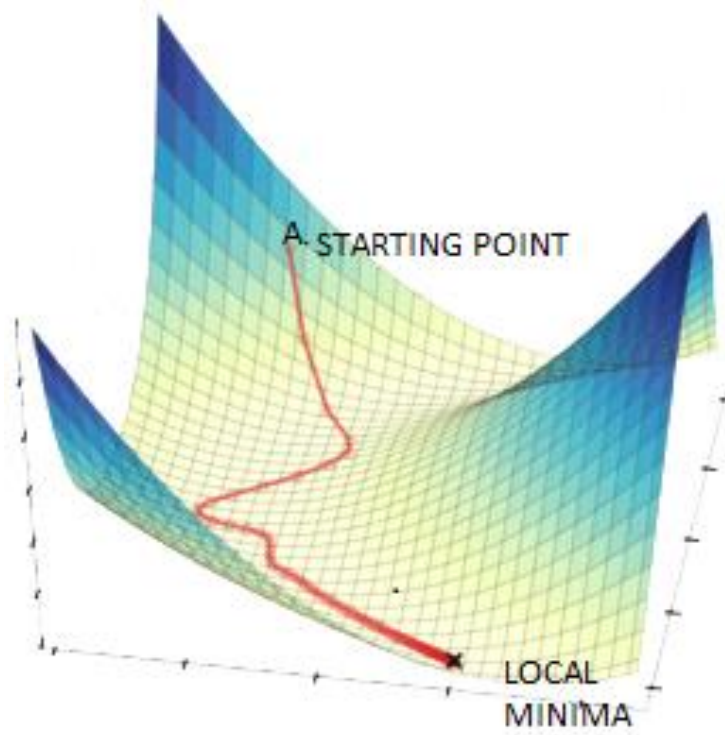


← You need this, to model non-linear functions!

Images from <https://towardsdatascience.com/machine-learning-for-beginners-an-introduction-to-neural-networks-d49f22d238f9> (03.07.2020)

Machine learning crash course

Neural network basics



The most common error function is

$$\sum_{(x,y) \in \text{Training}} (f(x))$$

In our case we have

$$f(x) \equiv f(x, w_1, \dots, w_n)$$

Partial derivatives of error function with respect to w_i give the direction you have to move w_i to decrease the error.

➡ Can in theory approximate any function

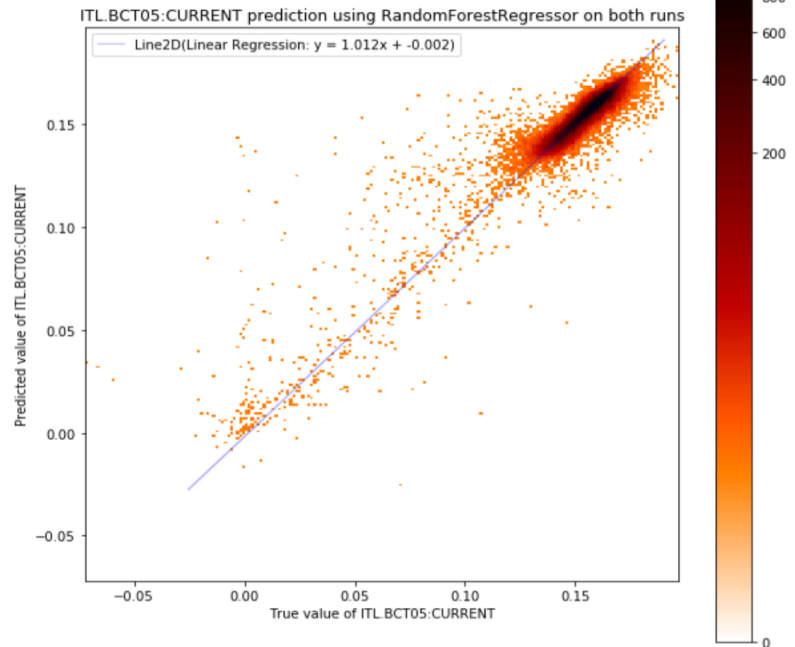
Images from <https://towardsdatascience.com/machine-learning-for-beginners-an-introduction-to-neural-networks-d49f22d238f9> (03.07.2020)

A first, promising result

Testing on two weeks of training data

Pearson correlation squared: 0.88
MSE of NN: 1.7e-05

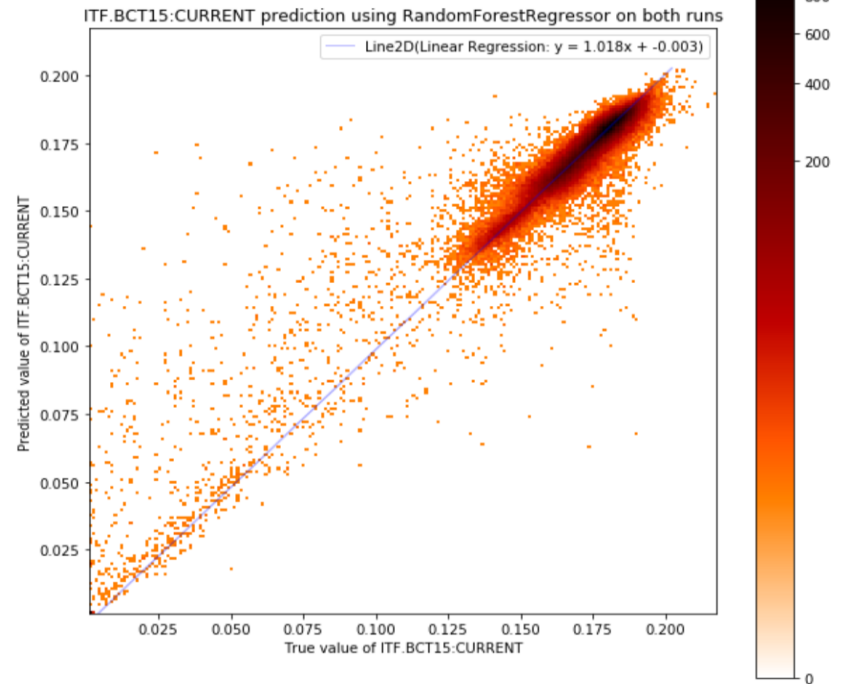
ITL.BCT05



- True ion current on x-axis
- Predicted current on y-axis

Pearson correlation squared: 0.92
MSE of NN: 3.2e-05

ITF.BCT15

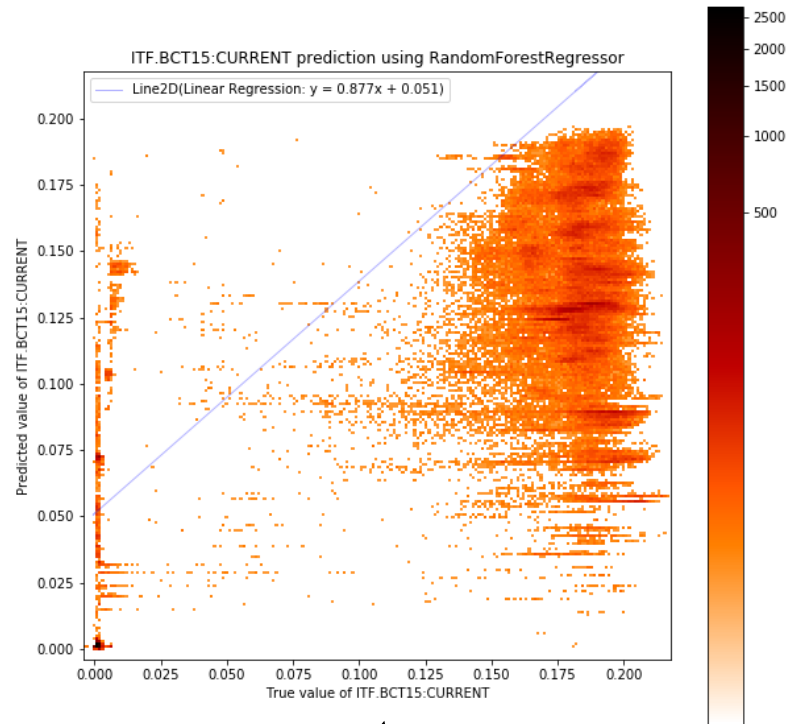


➡ Almost perfect prediction

Plots by Filip Široký

However, it generalizes poorly

Applying the previous model to new data



True BCT15 current (x-axis): constant ~180 uA

Predicted BCT15 current (yaxis): full range 25 uA - 200uA

Plots by Filip Šíroky

Why?

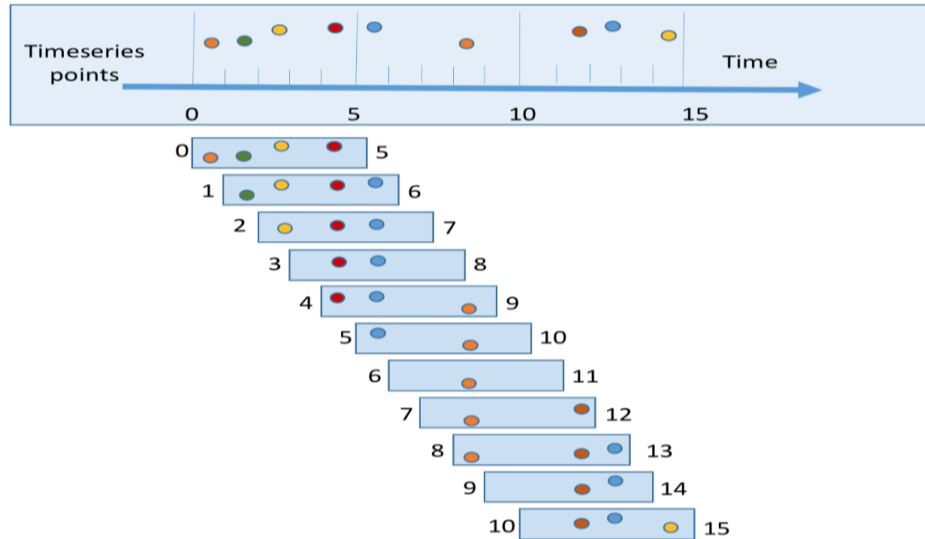
- Training data of only two weeks
- No time component included

A surprise?

- Supports Cluster Analysis that showed no direct link between used setting and source performance
- Many settings used throughout a year,
- test data might be from unknown regime

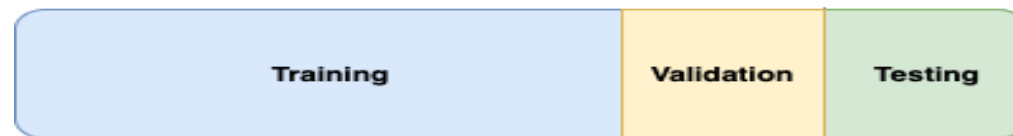
➡ Need to model **time series**

Modeling the next instant



Given the data from n points, what will the $n+1$ point look like?

- Model should learn the near future
- Can be viewed as a real-time simulation
- Input data can be multidimensional (HT current, Bias Disc Voltage, Oven Power, ...)



Classical training approach

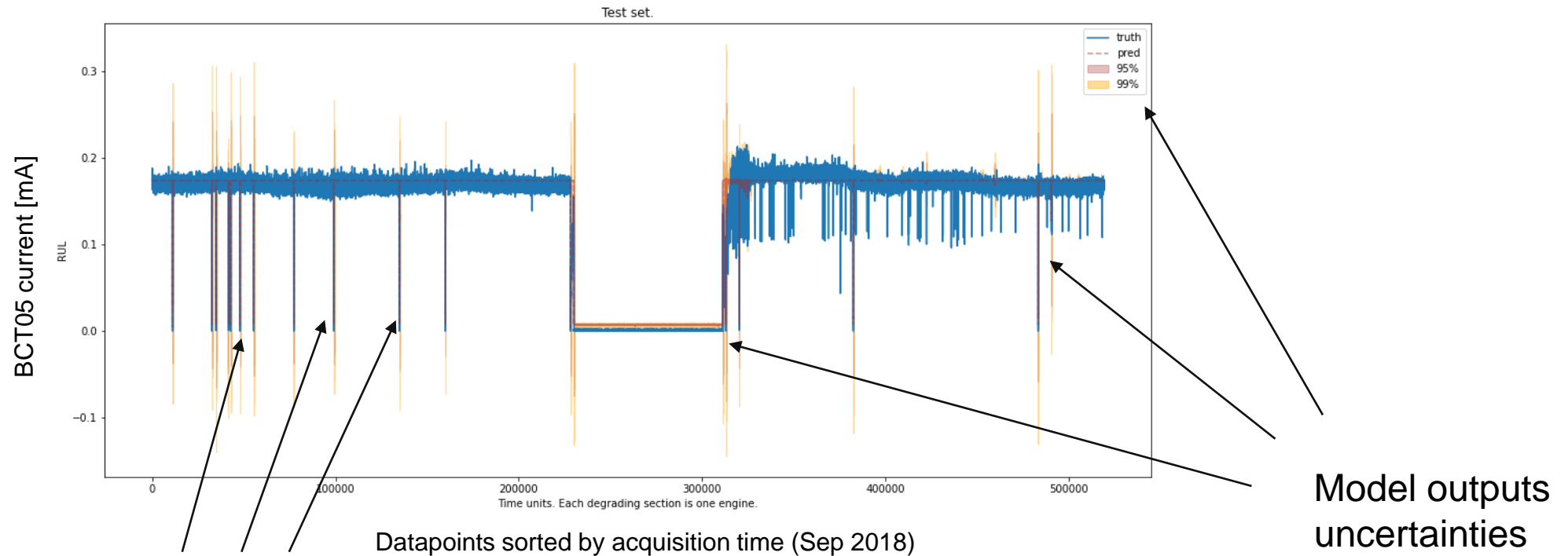


Time series training

➔ **In our case:** Given all information from the last 60s, what will the beam intensity be in the next few shots (1s).

Predictions of a temporal model

Results of a model trained on periods of September 2018



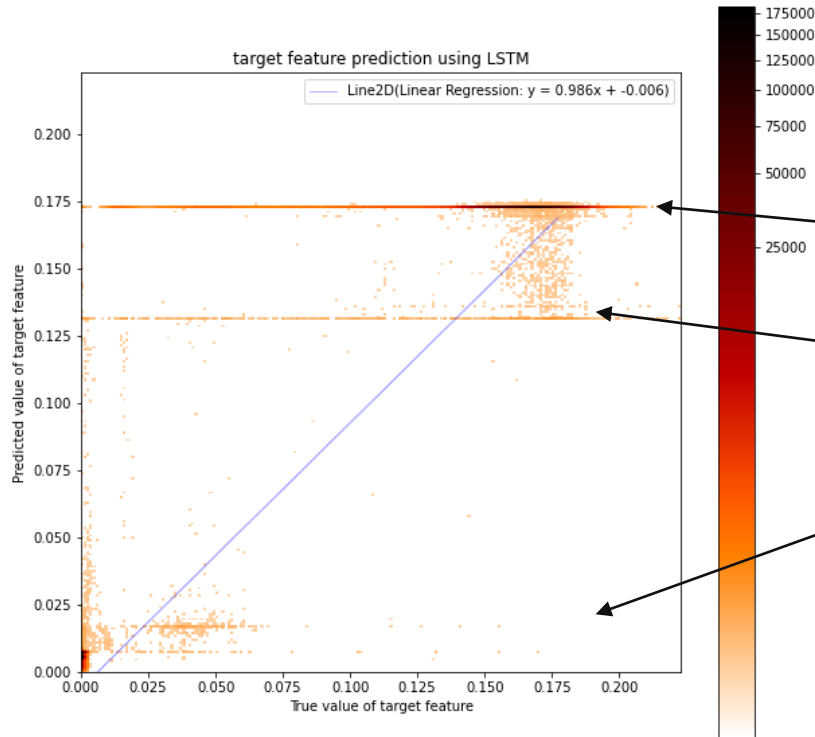
Model learns HT voltage
breakdown artifacts

➡ There is potential to learn
(stable) source operation

Plots by Filip Šíroký

Predictions of a temporal model

Results of a model trained on periods of September 2018



Model learns to predict three distinct levels

Average Beam current

Average value of bad shots

Source was off

Possible Interpretation?



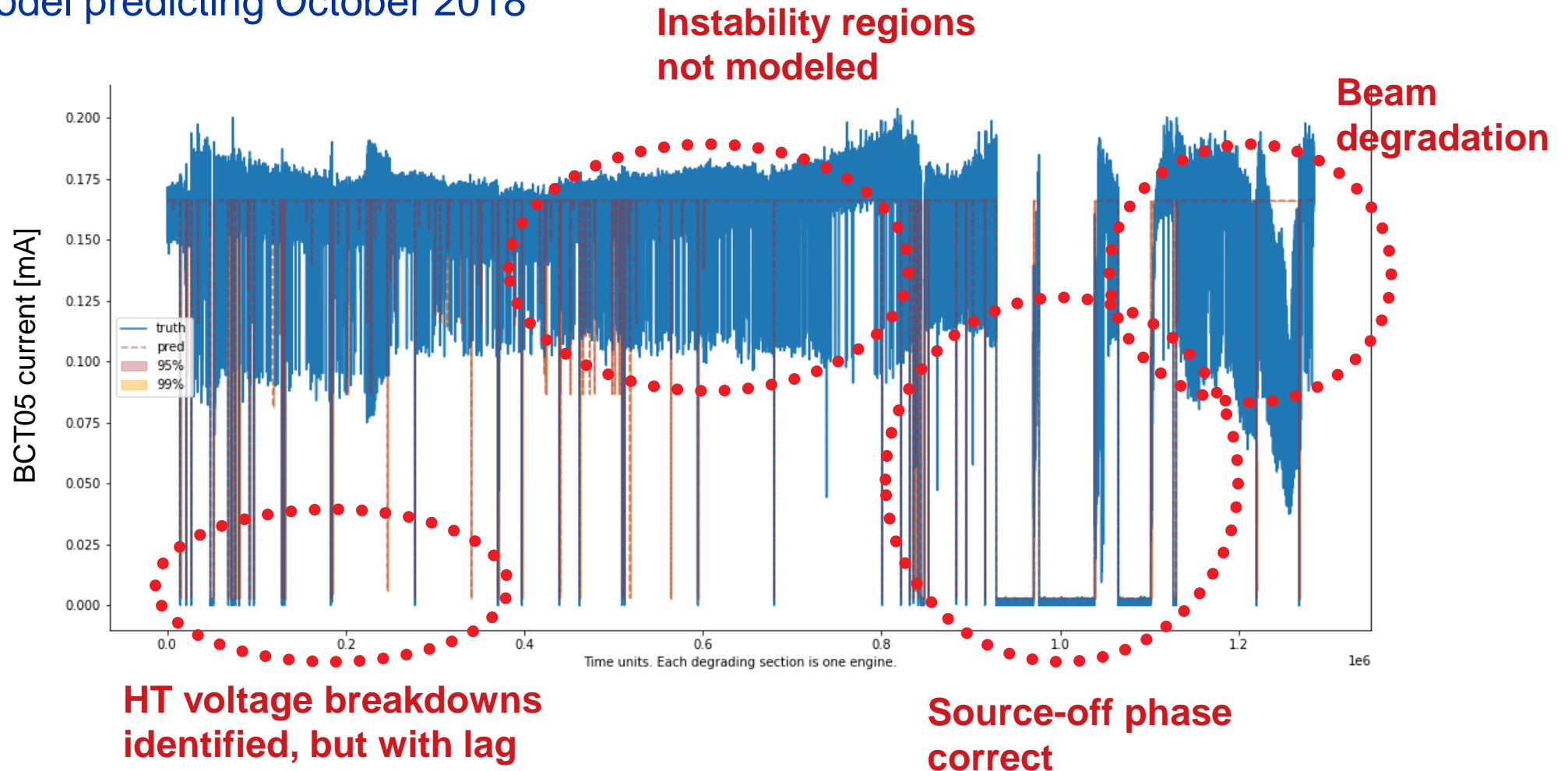
Some Key features of interest for operation could not yet be represented

Training and predicting Sep 2018

Plots by Filip Široký

Forecasting new horizons

Previous model predicting October 2018



Plots by Filip Široký

Conclusion and next steps

- Promising time series models that already learned very basic current features
- Improvement over randomly shuffled data (without time component)
- Output of uncertainty can be of very great interest to indicate degradation of the beam
- Predict Longer future (e.g. next 30min instead of next second)
- Understand lag and prediction of discrete states
- Try out different ML approaches
- Have a different representation of the target (Wavelet/FFT Transform of current)

