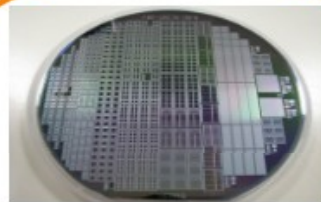
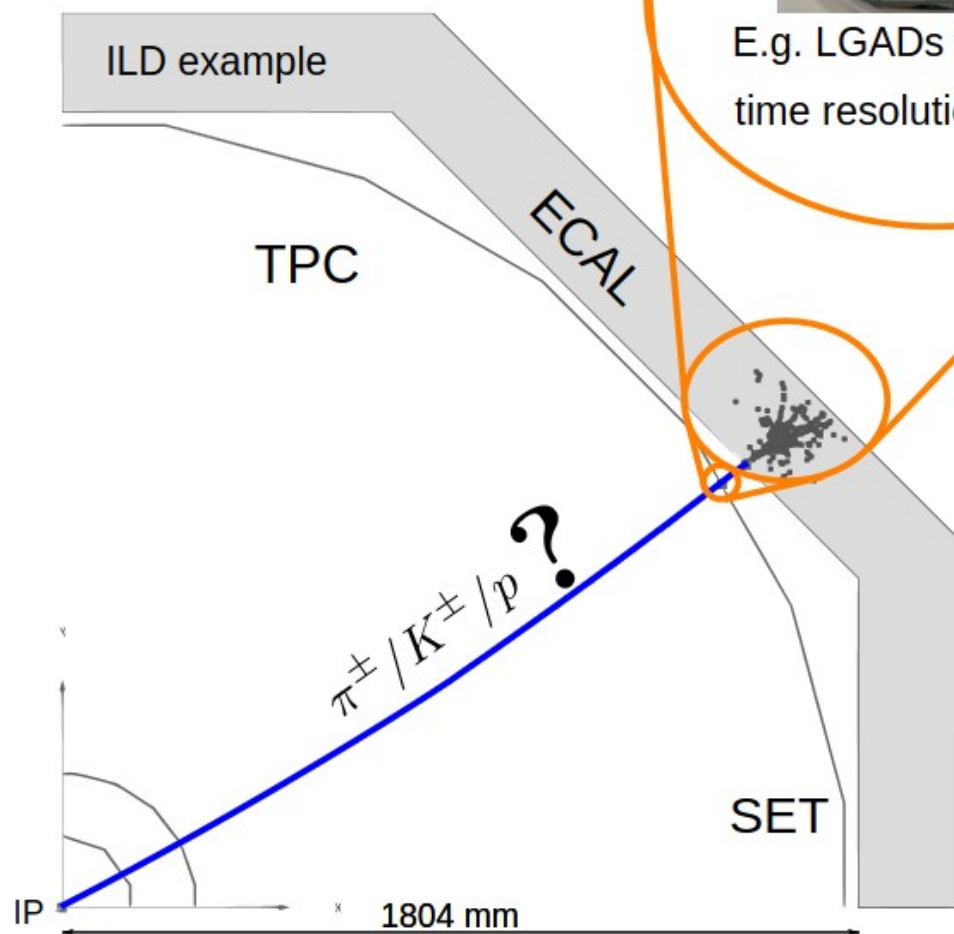


Updates from machine learning of TOF

Khrystyna Trofimiuk
02.03.2023

Basic principle of TOF

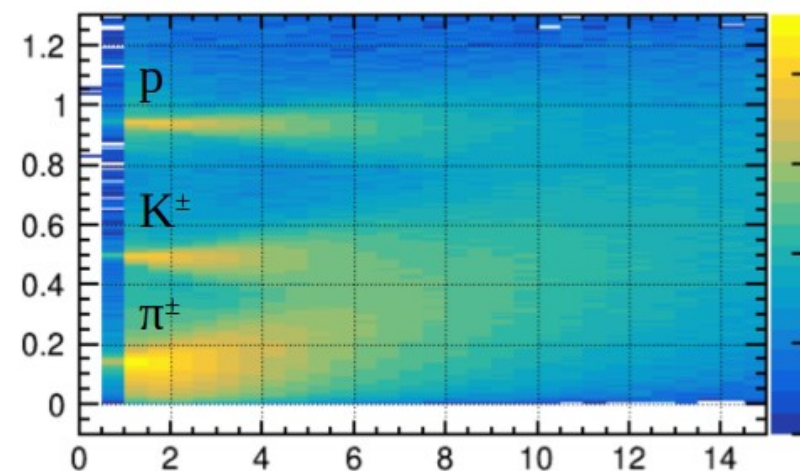
Study by Bohdan Dudar



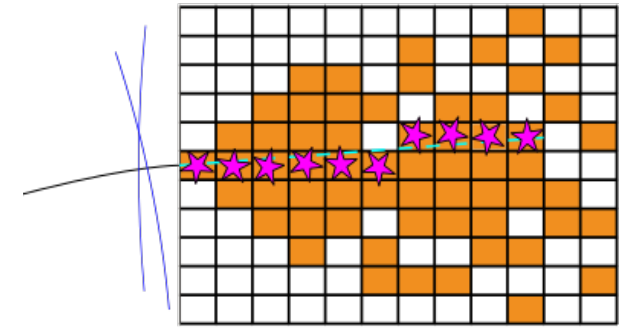
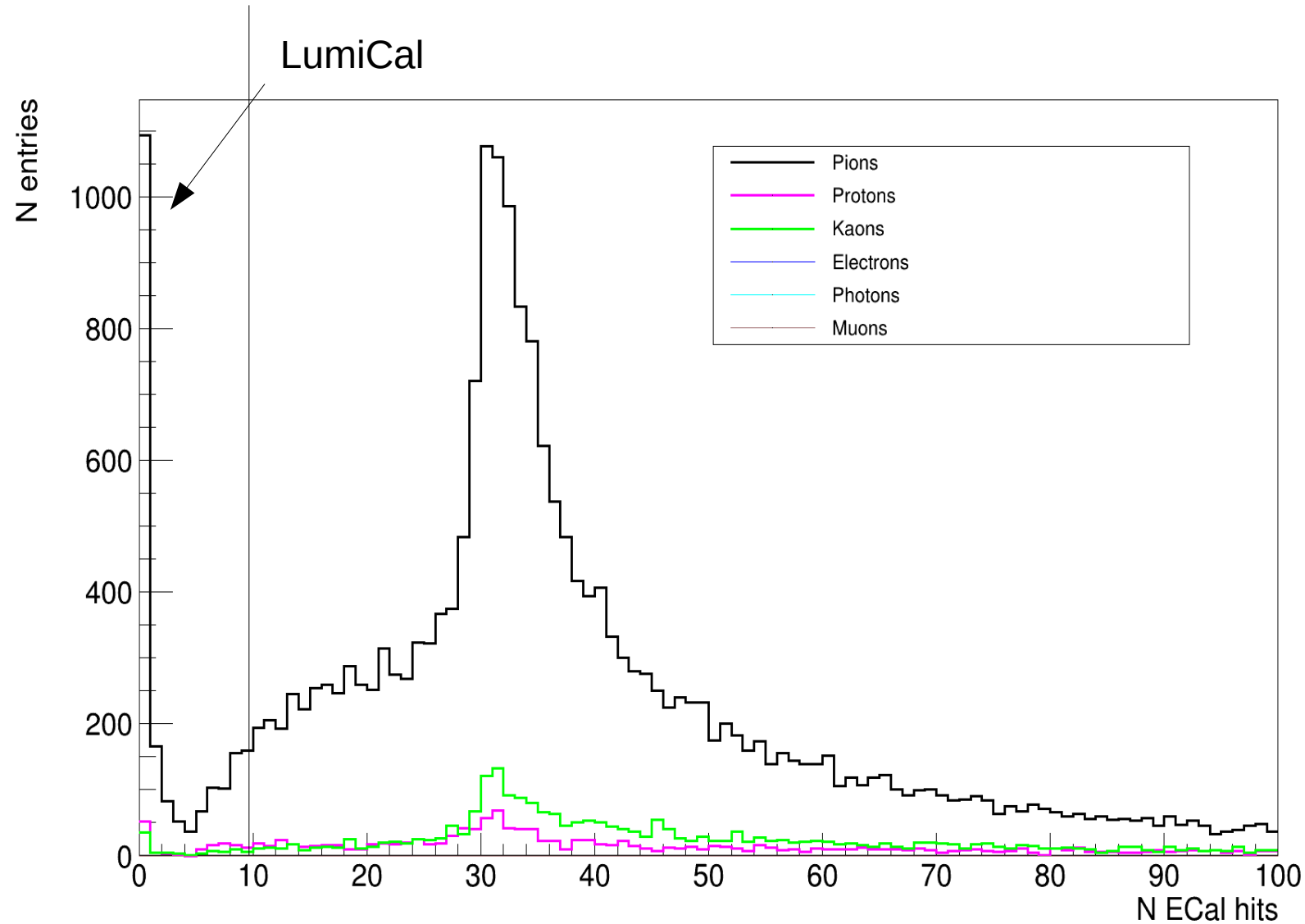
E.g. LGADs Si sensors
time resolution < 50 ps

$$\beta = \frac{\ell_{\text{track}}}{c \cdot \text{TOF}}$$

$$m = \frac{p}{\beta} \sqrt{1 - \beta^2}$$



N hits in the ECal for different particles



Average or fit of 10 hits
closest to the extrapolated
line

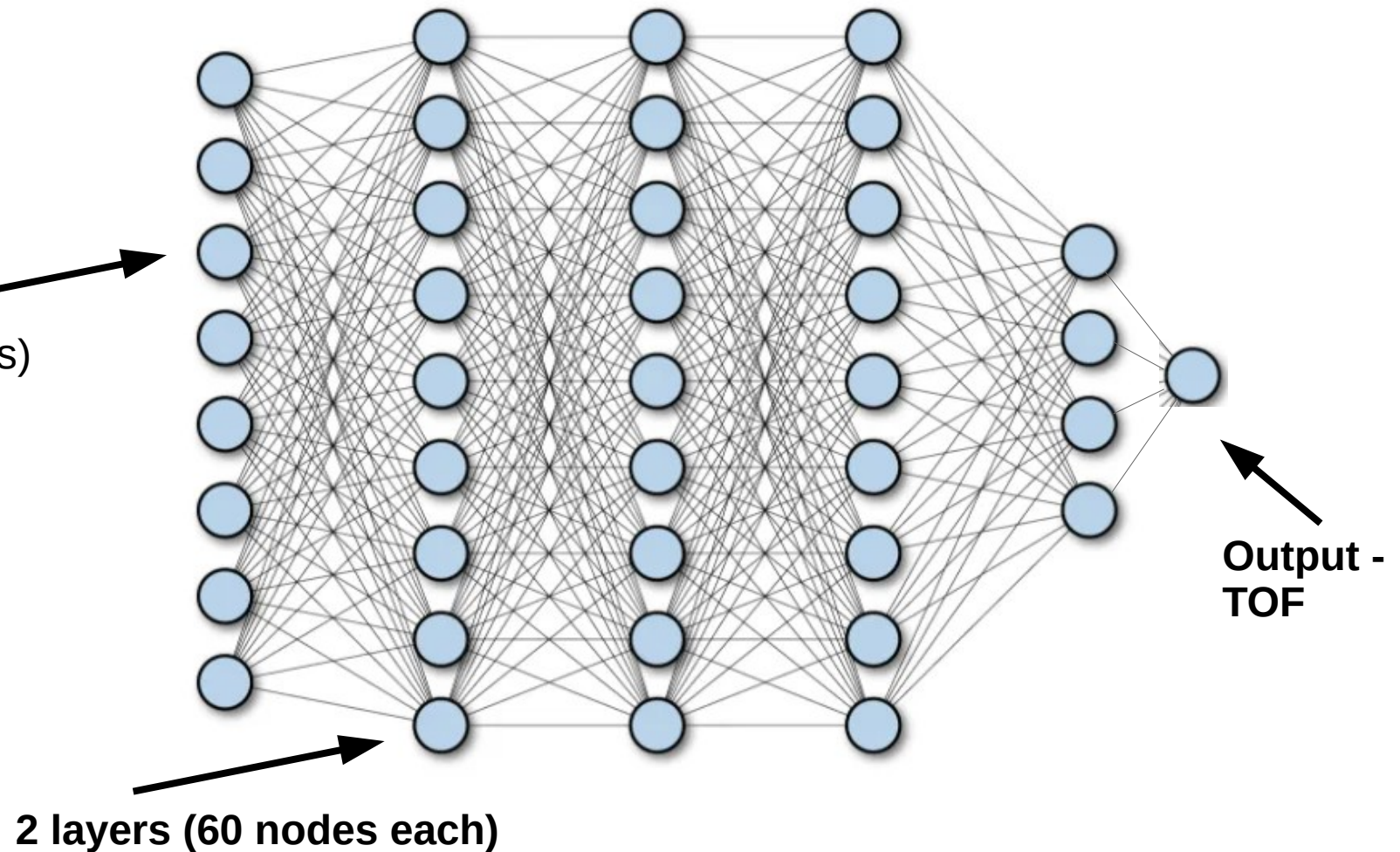
**Improving understanding
of number of hits in the
shower (**Ecal**)**

Fully connected neural network

First choice, main reason: the easiest one to start off

Input layer: 53 nodes:

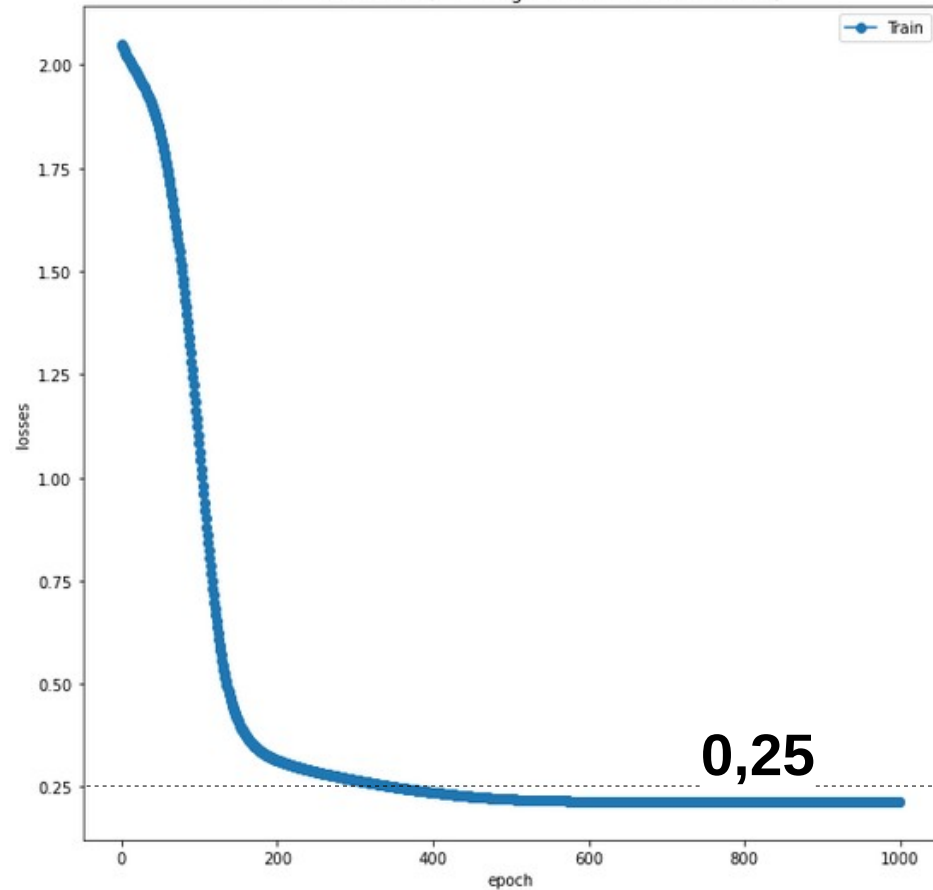
- Track p, pT, theta
- For each of 10 Ecal hits:
- time (smeared 50 ps)
 - energy
 - distance to track (coordinates)



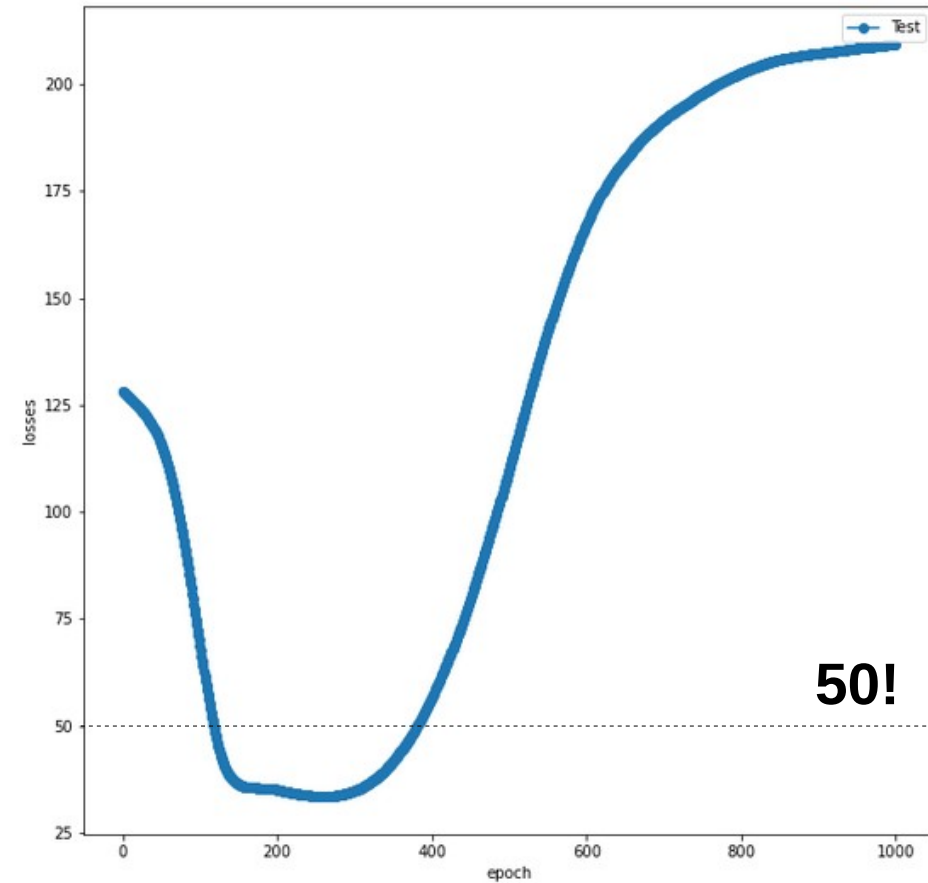
Fully connected neural network: trying to overfit 1 batch

Training loss

Train vs Test Losses (overfitting with one batch normalised)



Test loss

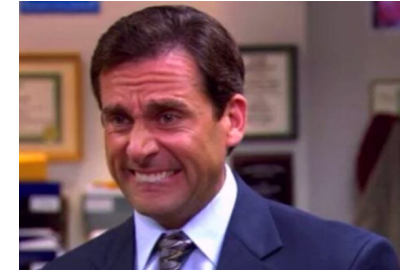


**“Successful
overfitting”**

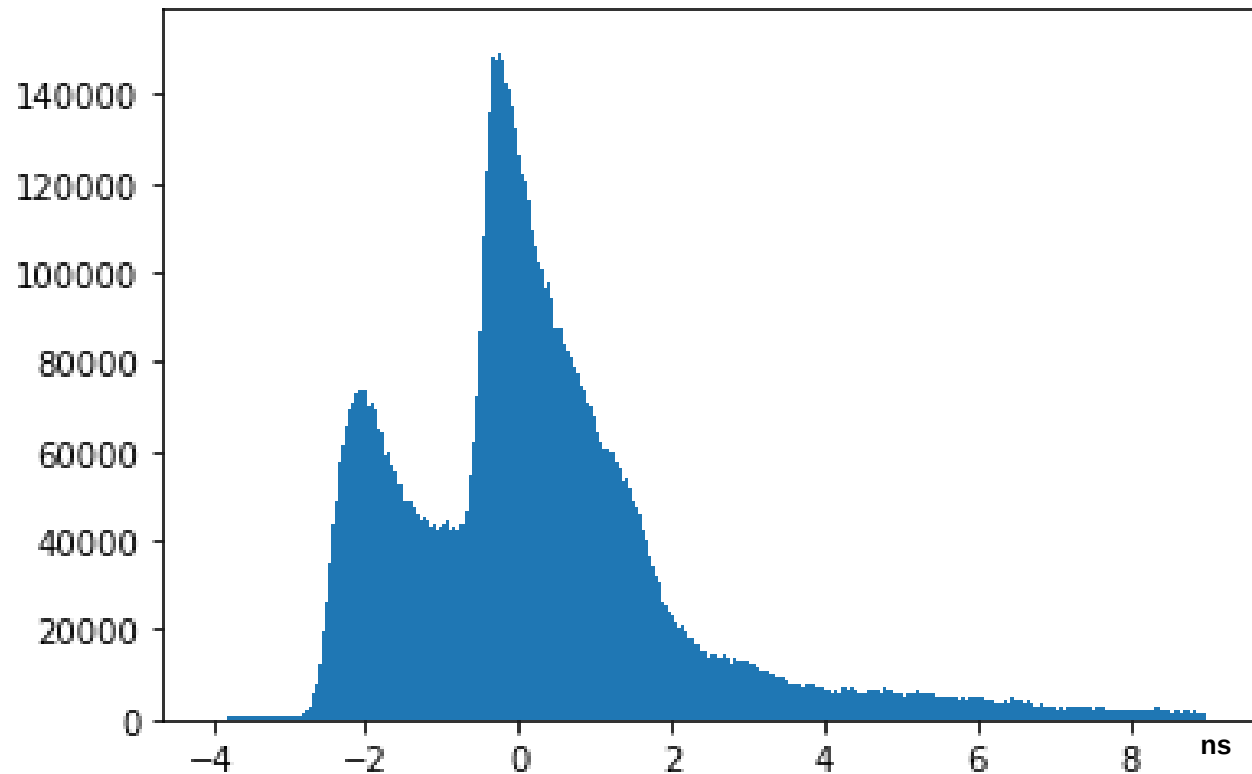
Fully connected neural network

How we count accuracy:

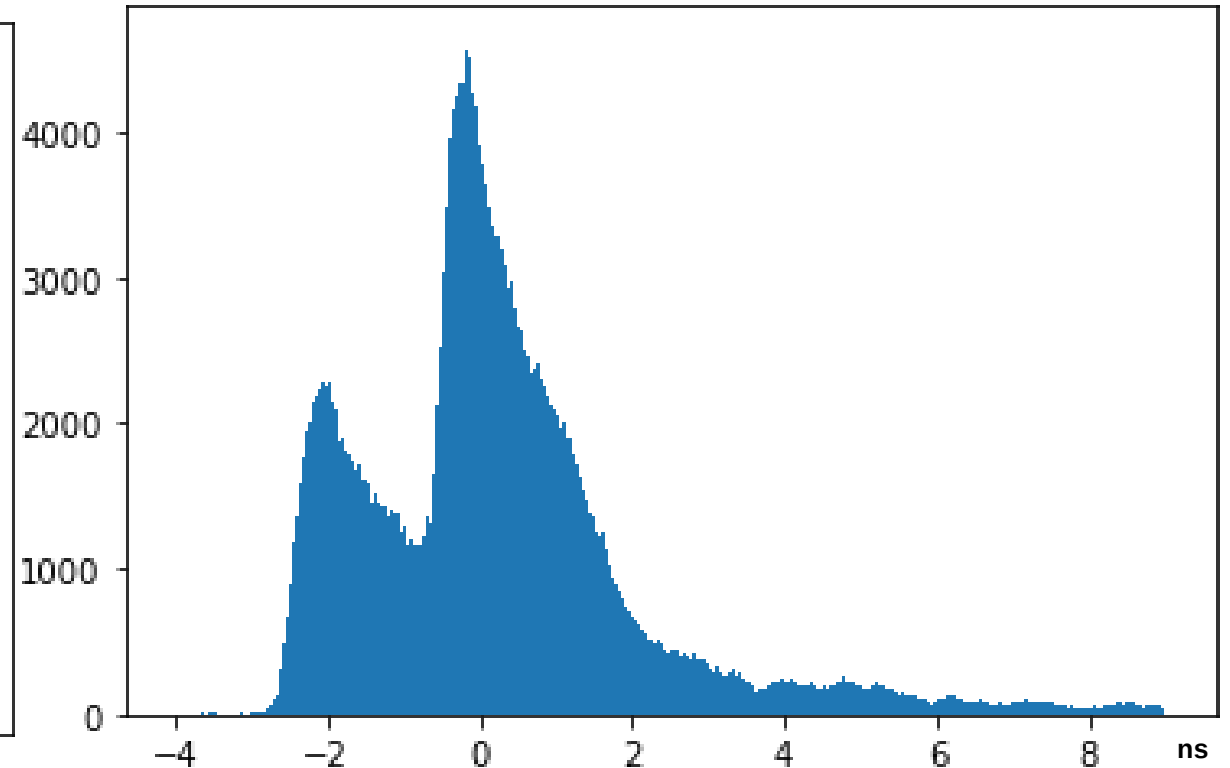
histogram with difference between
correct values and predicted ones



Differences for training (100 epochs)



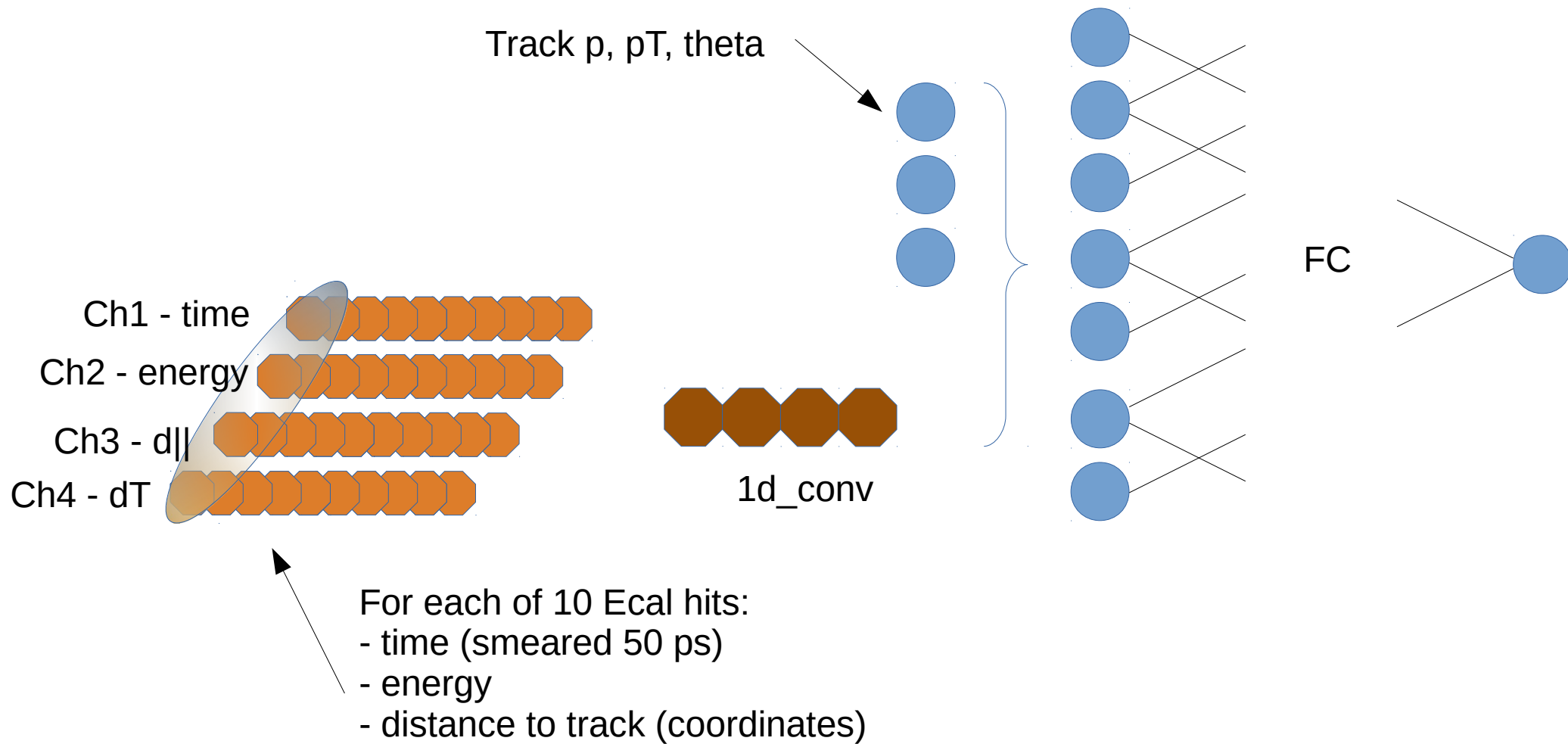
Differences for testing (100 epochs)



**Anatolii came up with a better
NN architecture idea!**

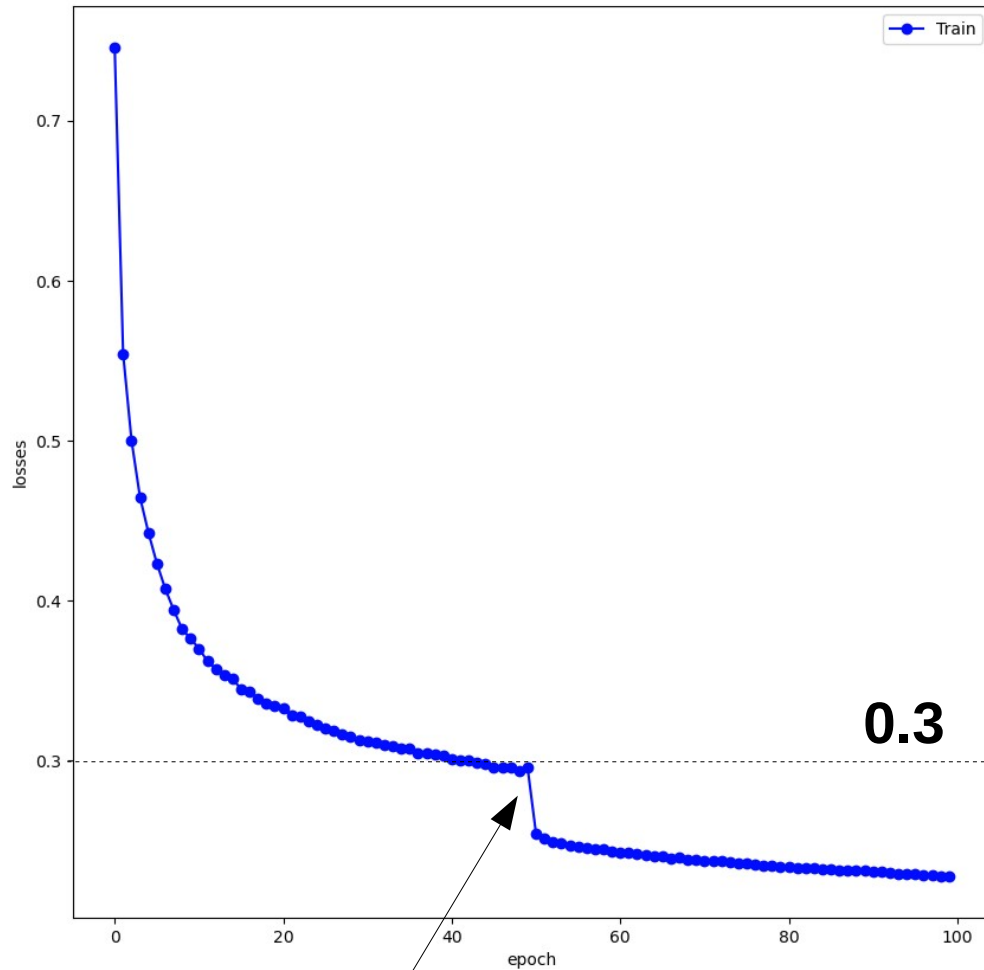
Convolutional Neural Network

Second choice

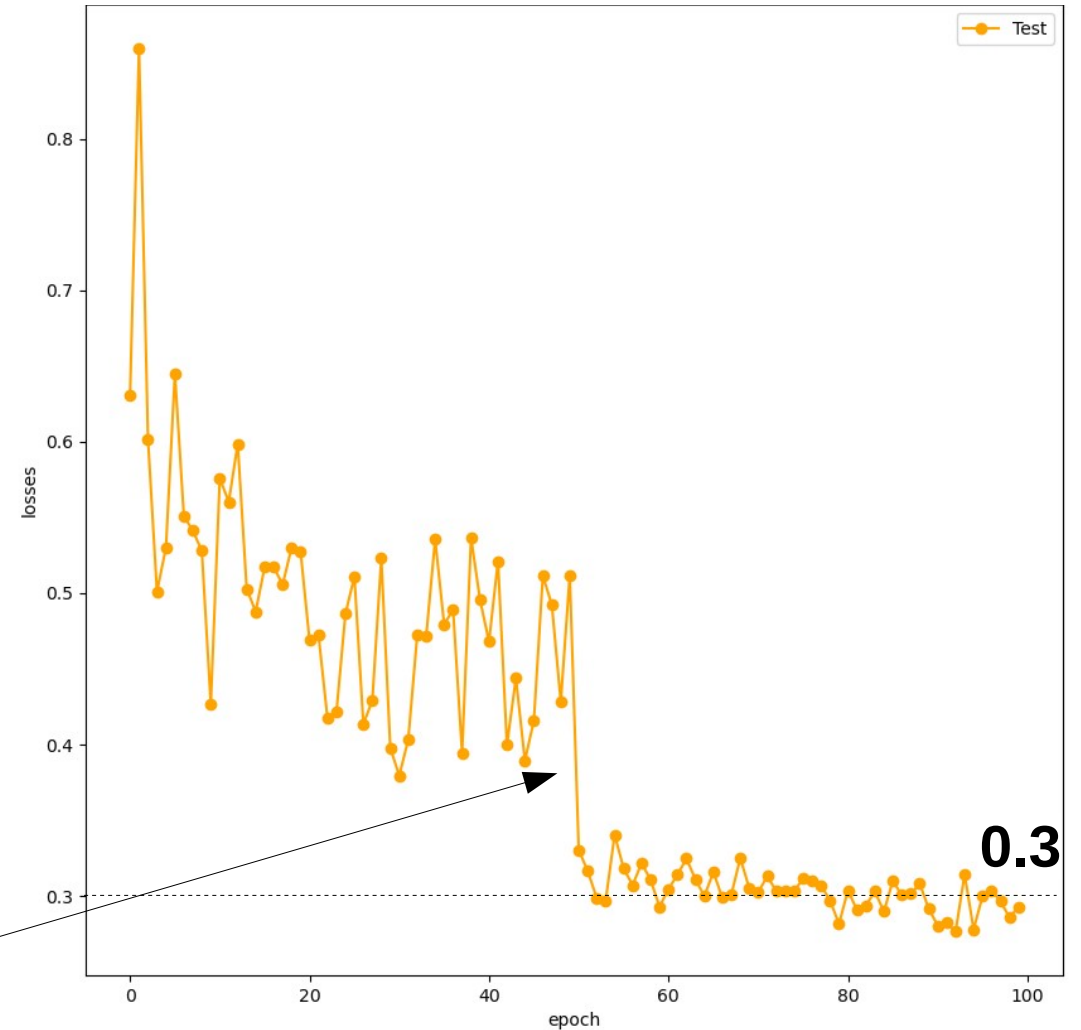


Convolutional Neural Network

Train losses for 100 epochs



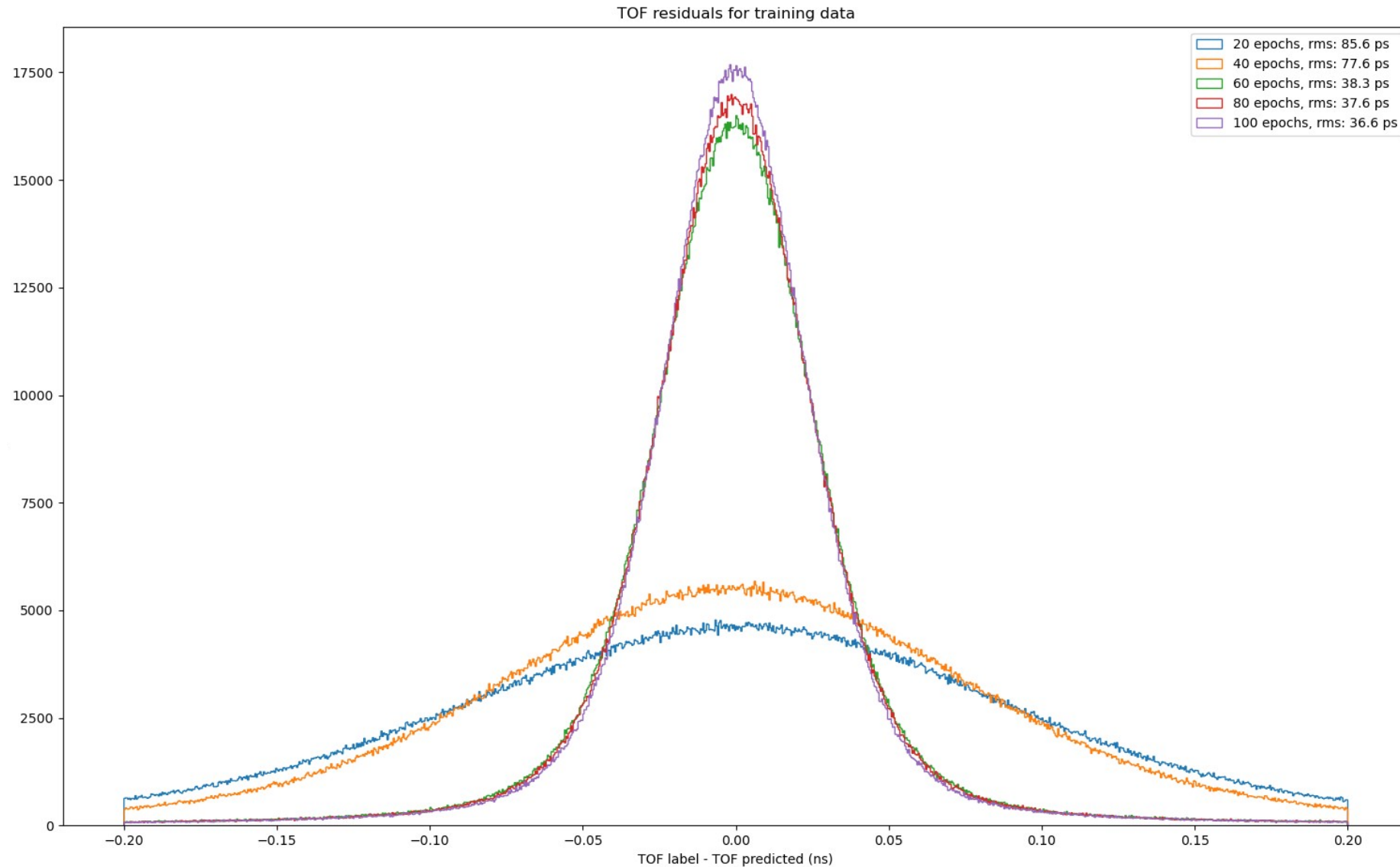
Test losses for 100 epochs



After 49 epochs training mode was changed on validation

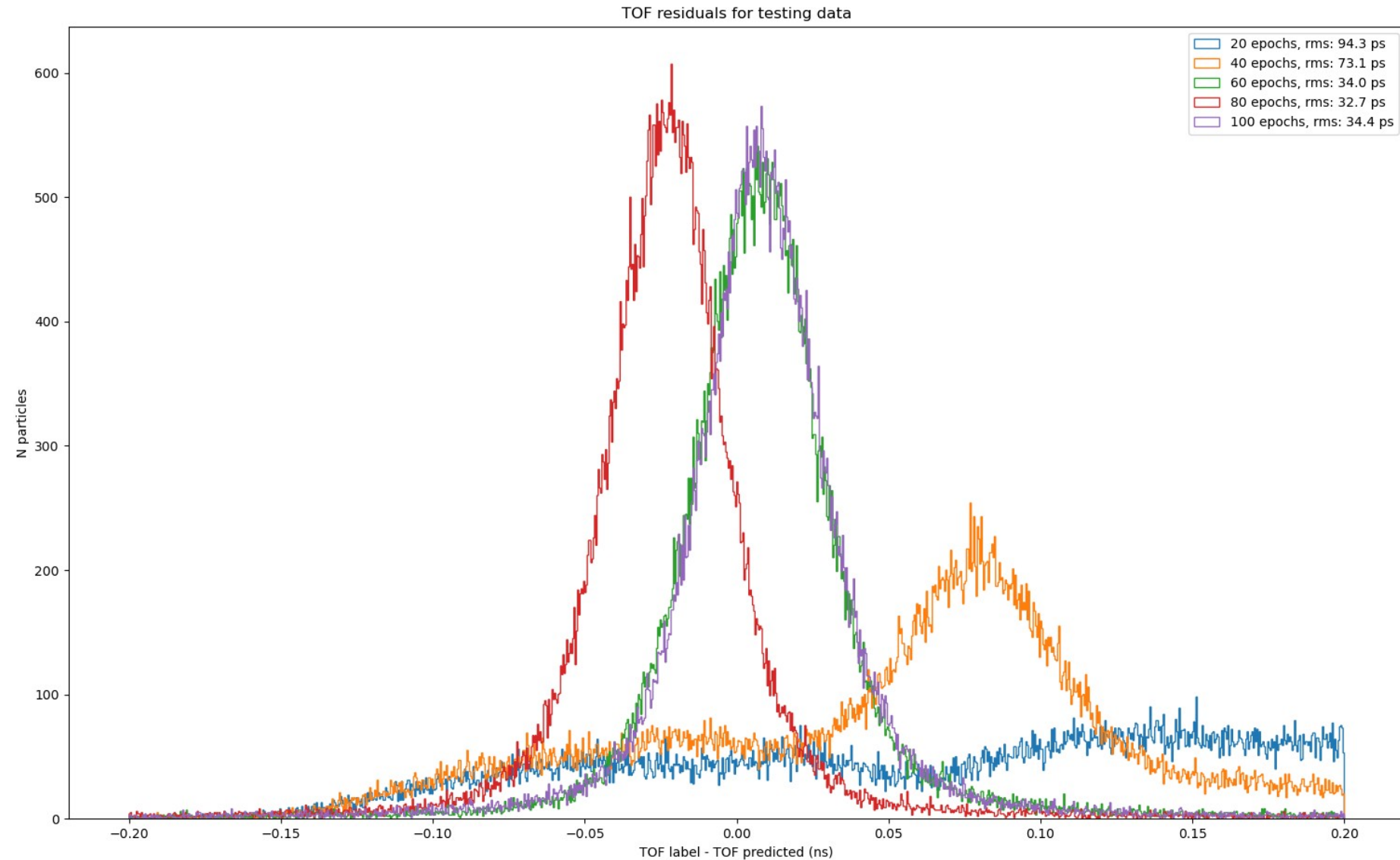
CNN: residual

histogram with difference between correct values and predicted ones (training)



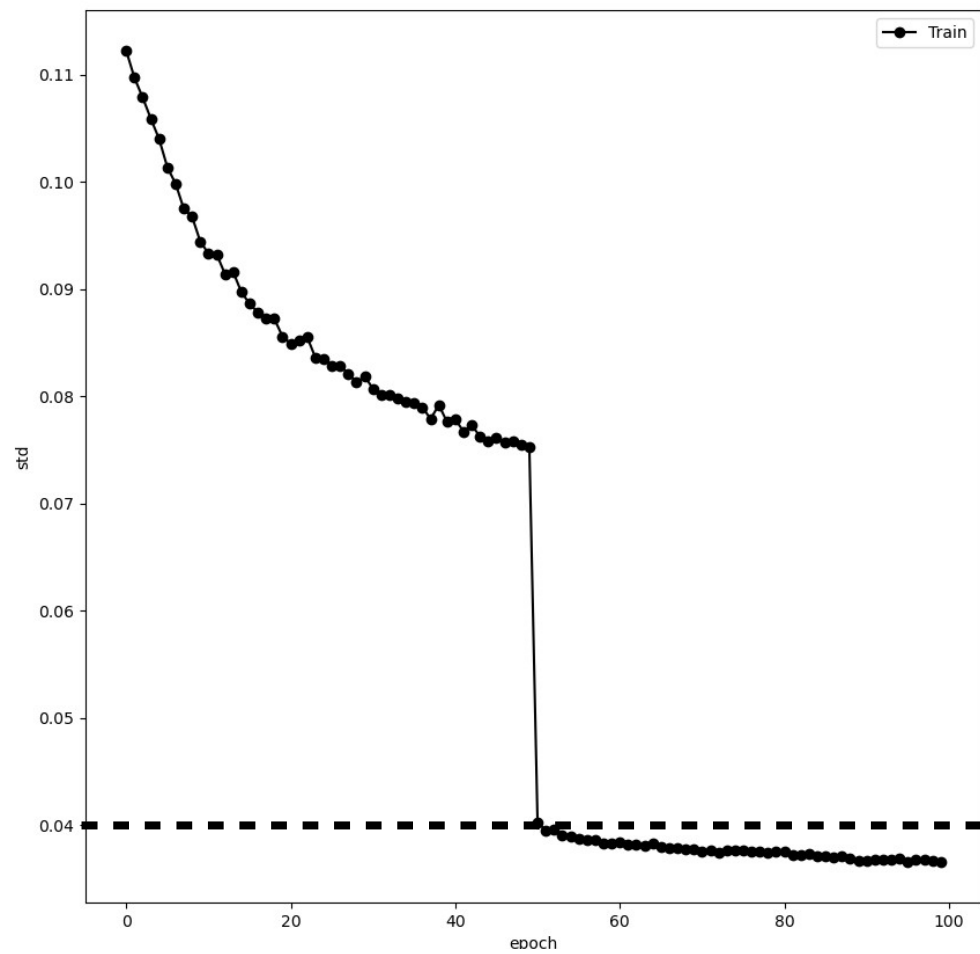
CNN: residual

histogram with difference between correct values and predicted ones (testing)



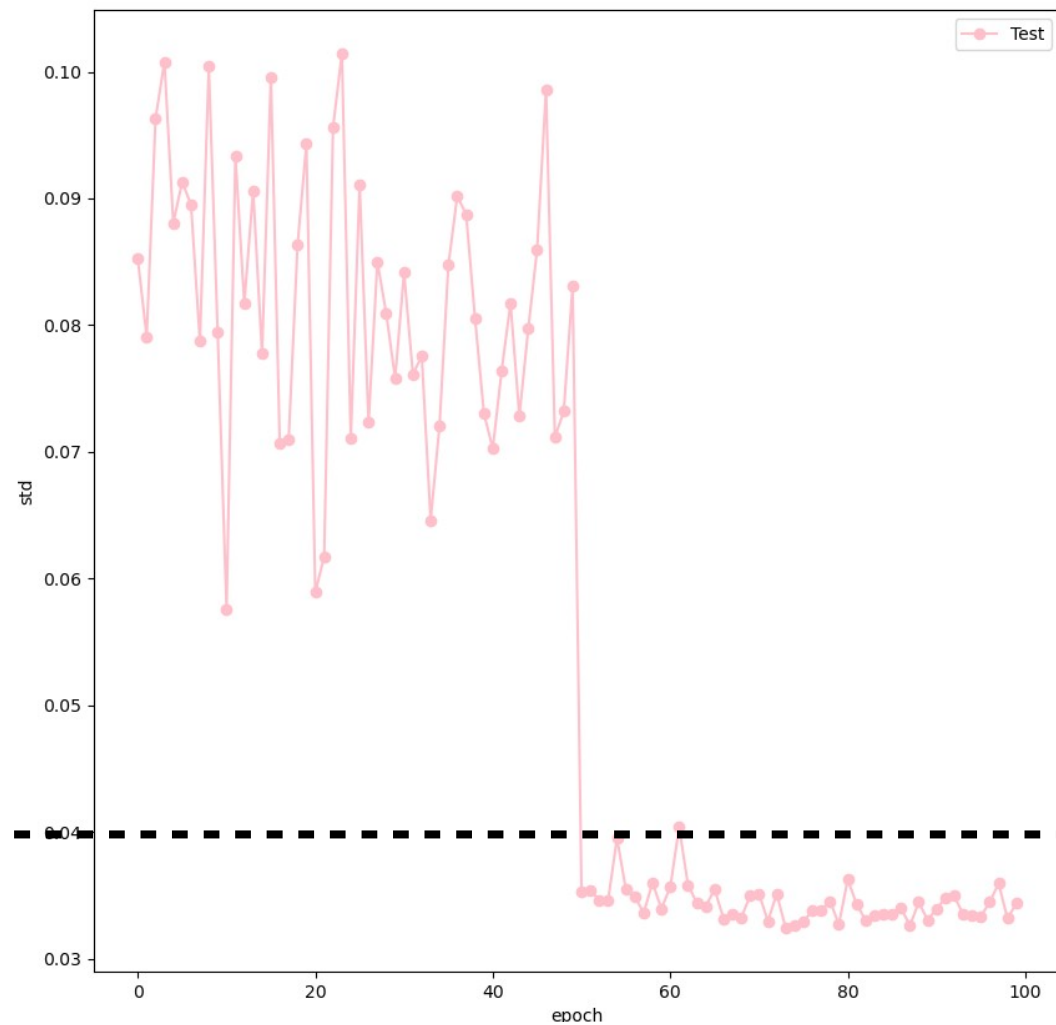
Convolutional Neural Network

Train standard deviation for 100 epochs



After 49 epochs training mode was changed on validation

Test standard deviation for 100 epochs

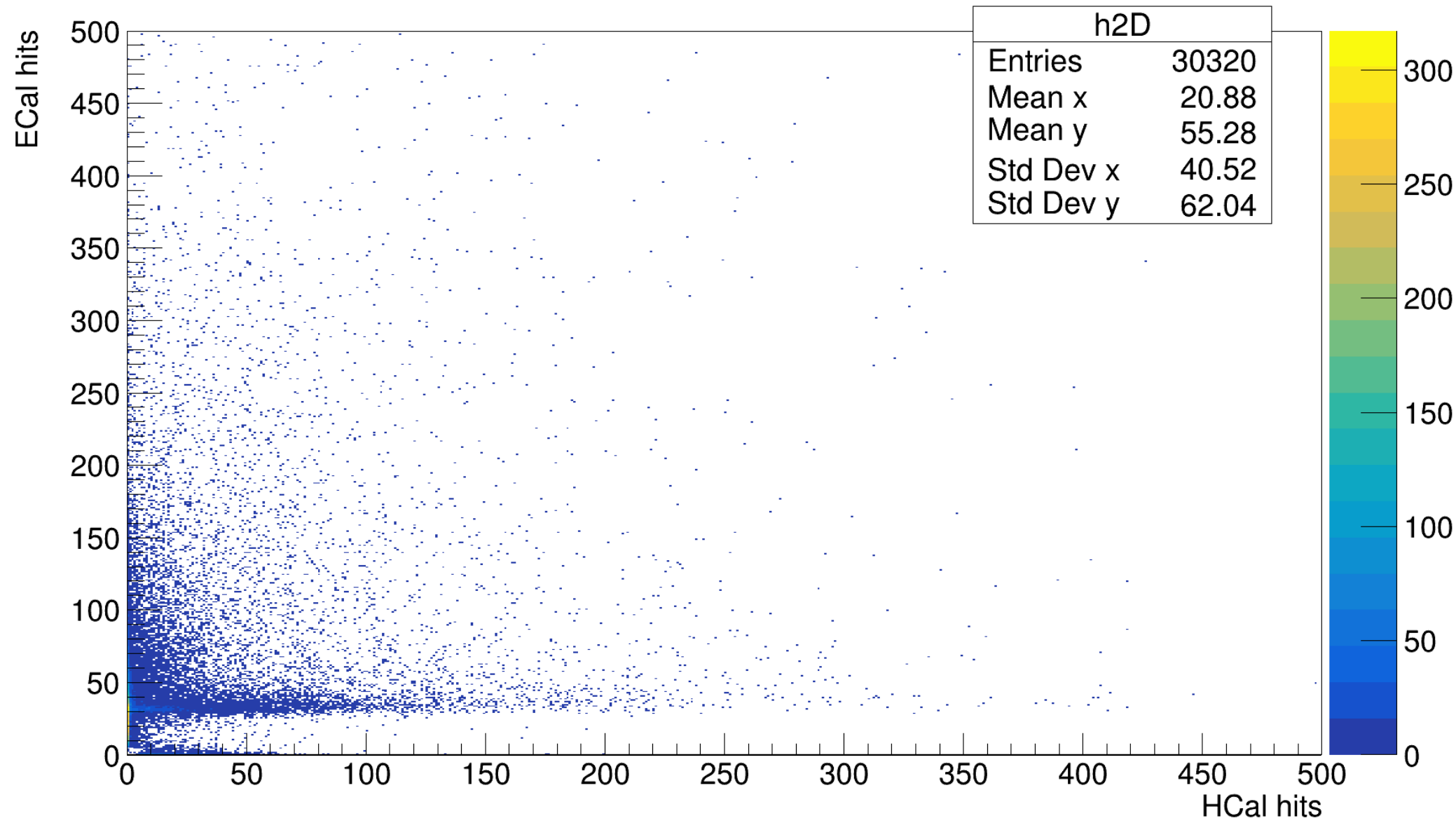


Conclusions

- It was my first time working on python and machine learning, so this experience is priceless
- It works!
- We have achieved **34.4 ps** TOF resolution using CNN after 100 epochs compared to **15.8 ps** of Baseline method “Frank’s average” ($50/\sqrt{10}$)
- Not the final result, it is still possible to improve
 - Train more epochs
 - Use more data (for now – we use 160 000 events with total ~2 800 000 particles)
 - Play with learning rate and NN architecture
- No systematic hyperparameter scans done yet

Thank you!

Correlation between HCal and ECal hits for all particles



Backup: math for TOF

$$\ell_{\text{track}} = \sum_{i=0}^n \ell_i = \sum_{i=0}^n \sqrt{\left(\frac{\varphi_{i+1} - \varphi_i}{\Omega_i}\right)^2 + (z_{i+1} - z_i)^2}$$

$$\beta = \frac{\ell_{\text{track}}}{c \cdot \text{TOF}}$$

$$p_i = e \frac{|B_z|}{|\Omega_i|} \sqrt{1 + \tan^2 \lambda_i}$$

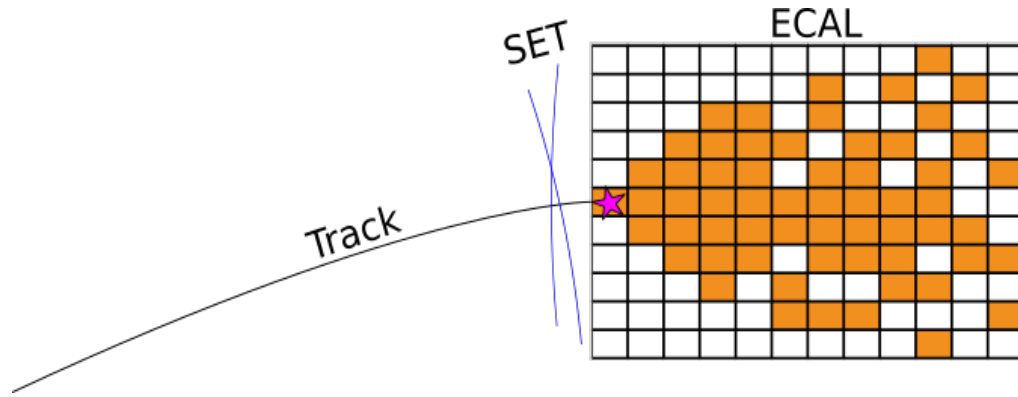
$$p = \sqrt{\langle p^2 \rangle_{HM}} = \sqrt{\sum_{i=0}^n \ell_i / \sum_{i=0}^n \frac{\ell_i}{p_i^2}}$$

$$m = \frac{p}{\beta} \sqrt{1 - \beta^2}$$

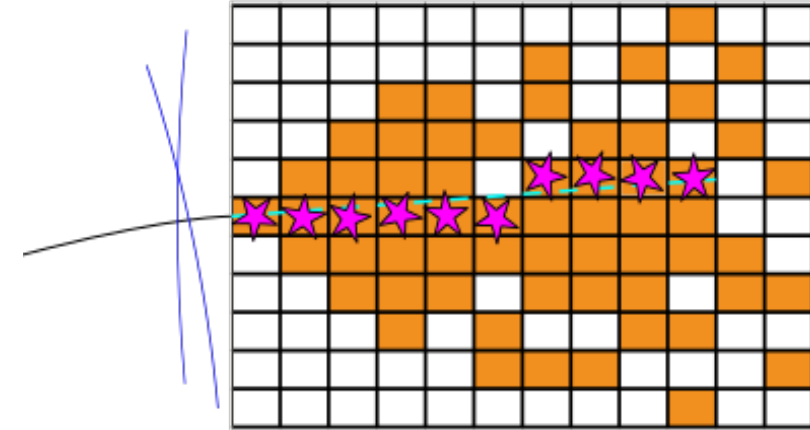
Winfried A. Mitaroff

arXiv:2107.02031

Backup: How do we implement TOF in the detector?



Single closest/fastest hit



Average or fit of 10 hits
closest to the extrapolated
line

Contact

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Department

E-mail

Phone