



Enhancing Detector Alignment with ML

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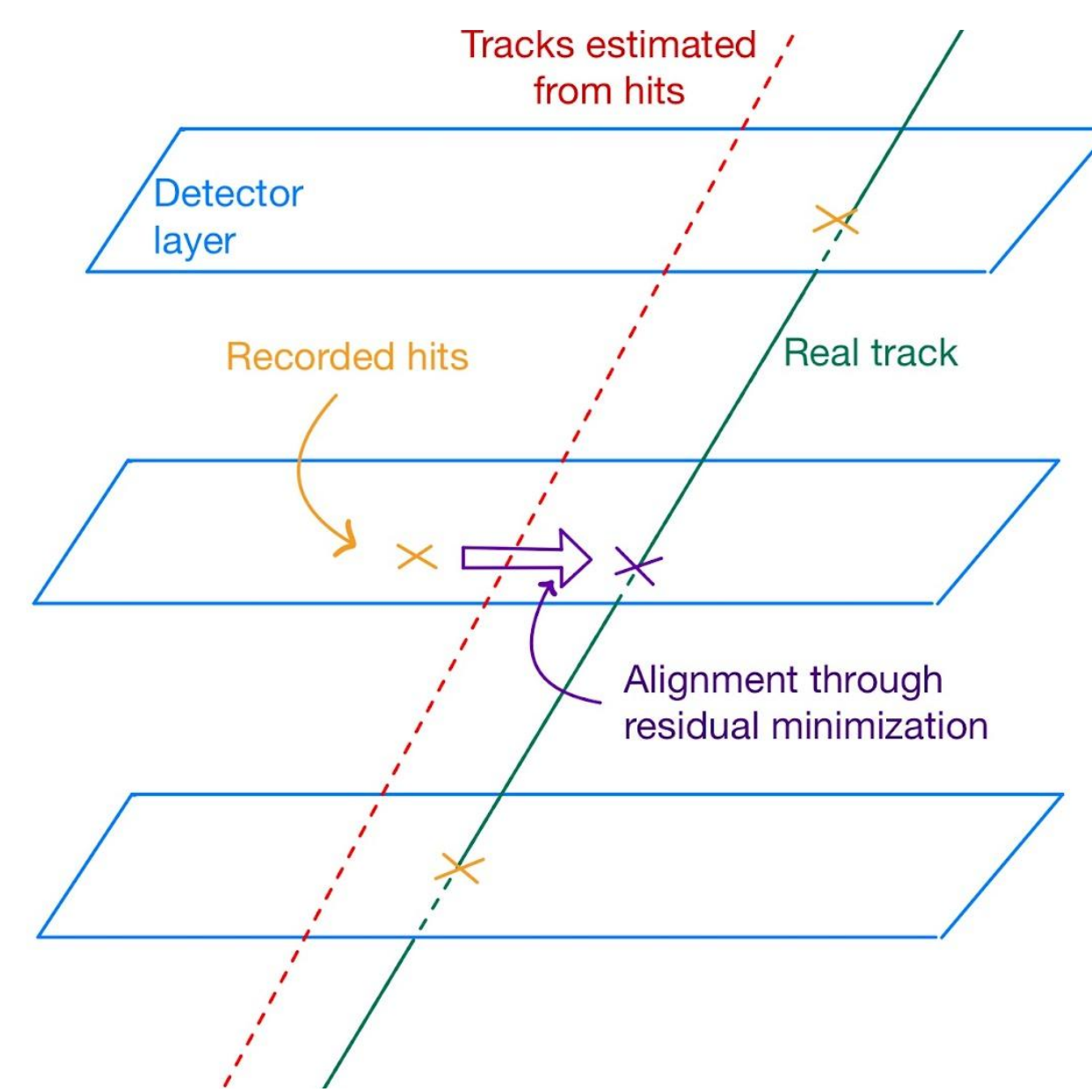
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Motivation

Current alignment techniques:

- Fit real tracks to reconstructed tracks by reducing residuals
- Classical fitting algorithms can be computationally heavy (100s of thousands of parameters)

Our goal: use GNNs to reduce the areas for which classical alignment algorithms are used.

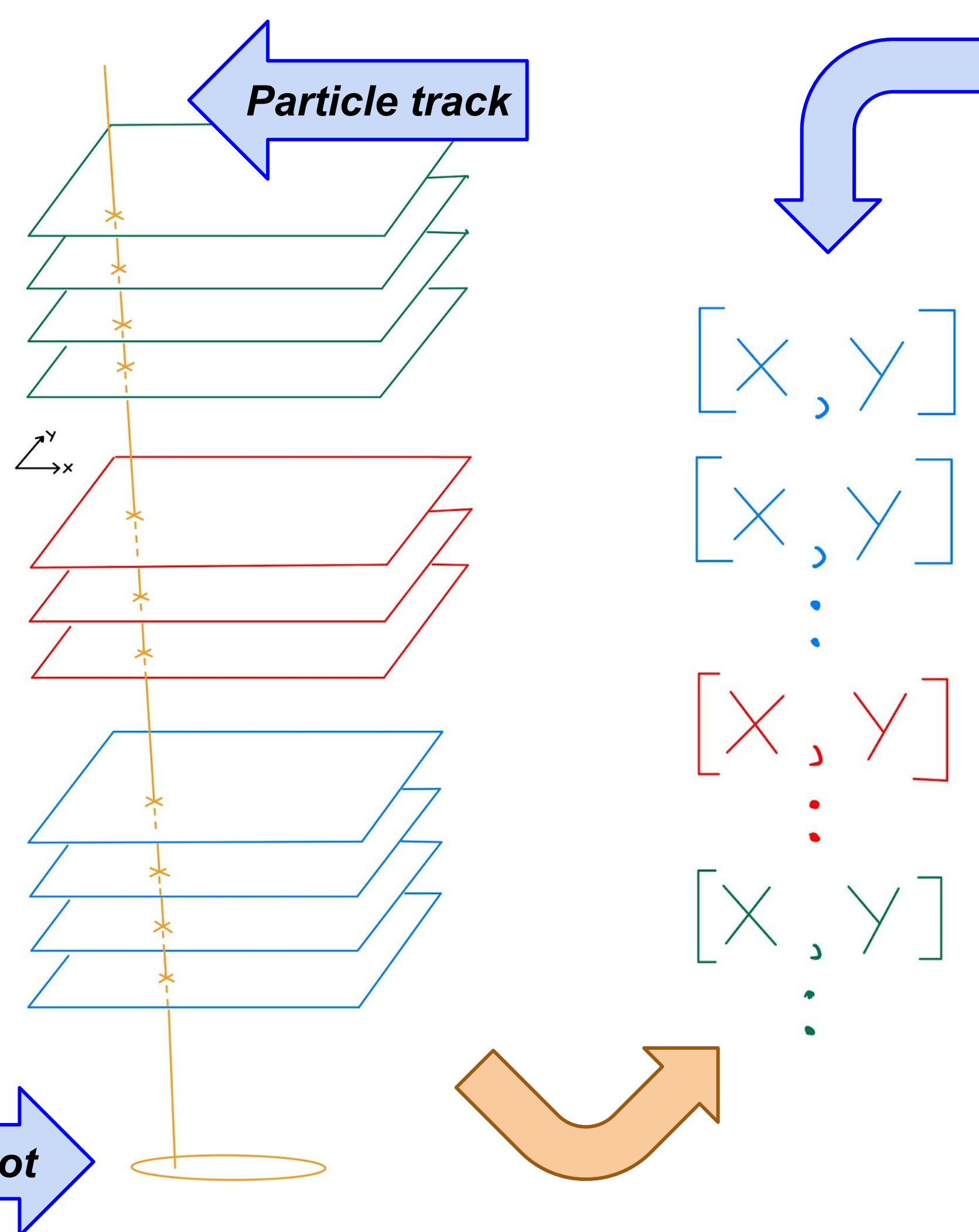


Proof-of-concept test:

- Telescope detector with 10 parallel layers
- Trained on one or two neighbouring misaligned layers
- Tested on two random misaligned layers
- Promising results

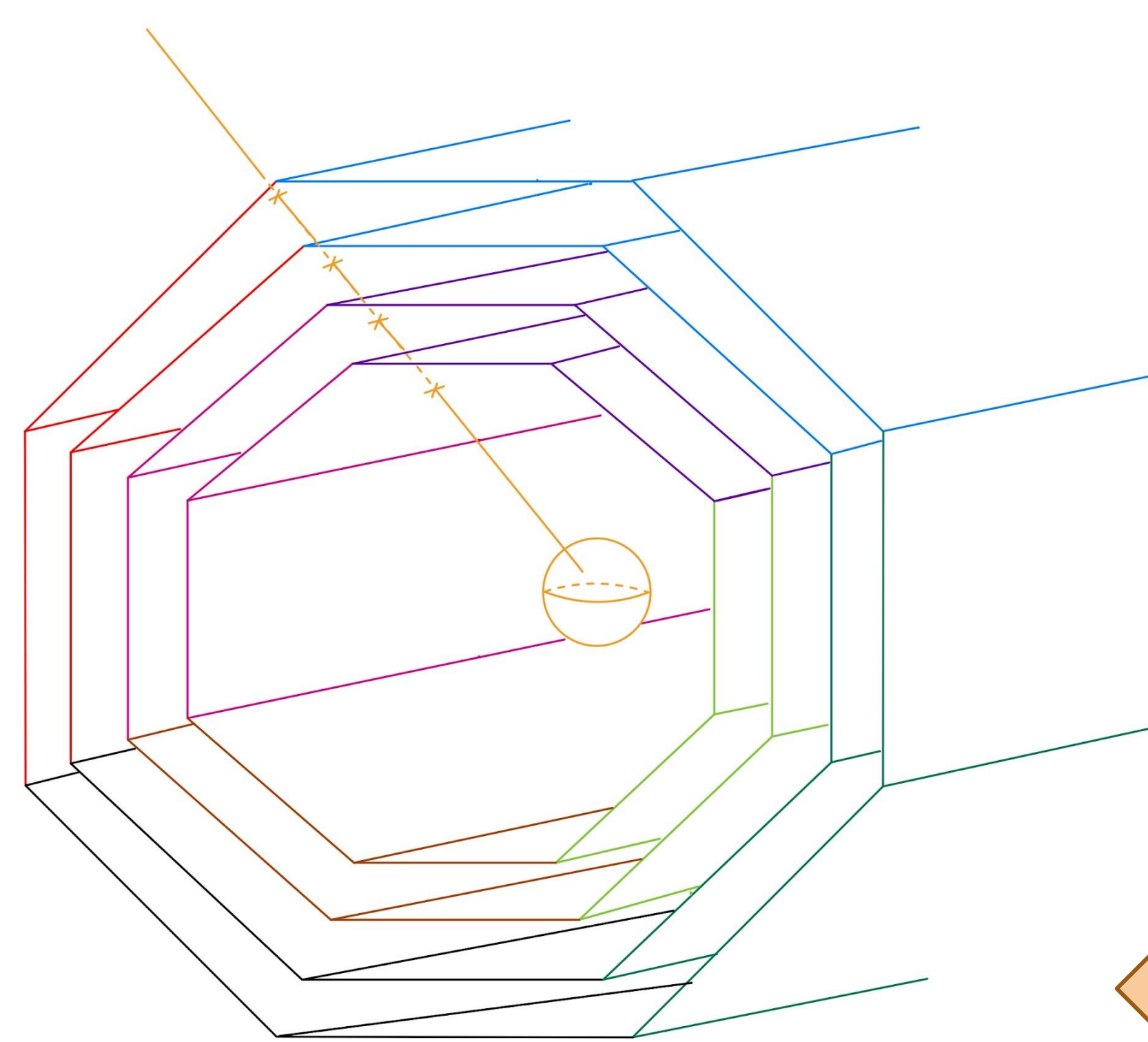


Detectors



Note!

To create training datasets with misalignments, some readings are artificially shifted after calculating the correct intercept values.



$[id_Z, id_P, x, y]$
...
 $[id_Z, id_P, x, y]$
...

Networks

[graphs]

GCN

- A. Two SAGEConv hidden layers
- a. 32 hidden nodes
 - b. ReLu activation
- B. Final Linear layer
- a. 4 out features
 - b. No activation function

0.97
4.32
-1.15
0.01

Loss

CrossEntropyLoss for multiclass classification.

Weights based on the occurrence rate of each class.

Note!

Undergoing testing of hypernodes/hyperlinks in the architecture.

There were observed problems during the training that make the current implementation unpredictable.

Note!

Matching the 4 possible node types:
(0) no shift, (1) shifted with the block, (2) shifted alone, (3) shifted alone and with the block.

Results

True class	0	1	2	3	
	5836940 (88.4%)	607618 (9.2%)	116167 (1.8%)	39275 (0.6%)	
	352672 (25.2%)	882094 (63.0%)	131708 (9.4%)	33526 (2.4%)	
	362850 (20.2%)	730706 (40.6%)	628561 (34.9%)	77883 (4.3%)	
	0 (0.0%)	0 (0.0%)	100000 (50.0%)	100000 (50.0%)	
Predicted class		0	1	2	3

Telescope detector:

- 30 runs training, 100k particles/run
- Adam optimizer, 1e-3 learning rate;
- Convergence in around 10 epochs
- 99% accuracy on validation set
- Second evaluation on 10 runs with different shift types
- Inaccuracy in very defective setups (more than 50% of layers shifted)

Cylindrical detector:

- 10 runs training, 100k particles per run
- Adam optimizer, 1e-3 learning rate
- Convergence not achieved in 50 epochs
- 80% accuracy for training validation
- Second evaluation on 10 runs with similar errors
- Decent at finding block errors

True class	0	1	2	3	
	6661107 (86.3%)	747737 (9.7%)	282200 (3.7%)	25223 (0.3%)	
	391605 (26.8%)	1014561 (69.4%)	30507 (2.1%)	24830 (1.7%)	
	452607 (69.1%)	88199 (13.5%)	105232 (16.1%)	9188 (1.4%)	
	35225 (26.9%)	76537 (58.4%)	9471 (7.2%)	9728 (7.4%)	
Predicted class		0	1	2	3

Prospects

Hyperstructures:

- More complex architectures require better defined hyperstructures for understanding the geometry
- Working idea: hyperlinks between nodes in the same block
- Future idea: hypernodes for block evaluation

Geometry:

- Current geometries still similar to “proofs-of-concept”
- Working idea: train using data from standard tracking software e.g. ACTS
- Future idea: creating a “reasonable enough” model that is easily adaptable for a large range of geometries

Pipeline integration:

- This work should not fully replace global fittings, just help narrow down areas simulated
- Working idea: making the frameworks compatible with popular data types and formats
- Future idea: potentially use it as a “real-time” misalignment monitoring tool