

Development of a framework based on PyTorch Lightning for training Graph Neural Networks (GNNs) for tracking

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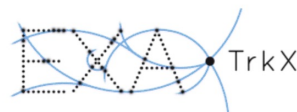
Sylvain Caillou

(On behalf of L2IT « Computing, Algorithms and Data » Team)

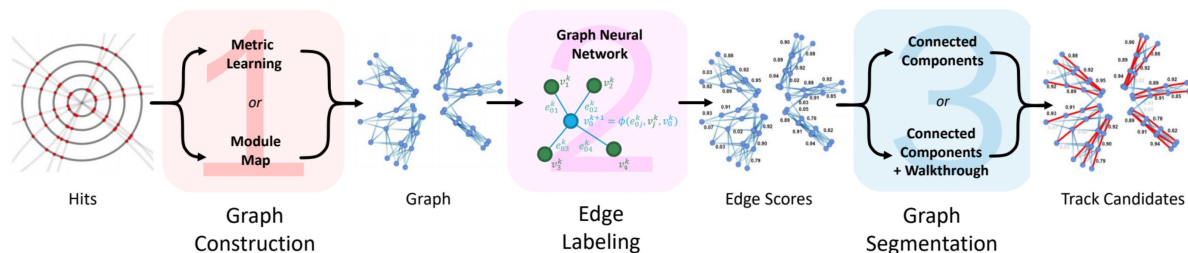
Context

- ⇒ Geometric ML have become one of the hottest fields of AI research
- ⇒ Increasingly popular for a large number of LHC physics tasks
- ⇒ GNNs perform pretty well to learn geometric pattern of particles tracks

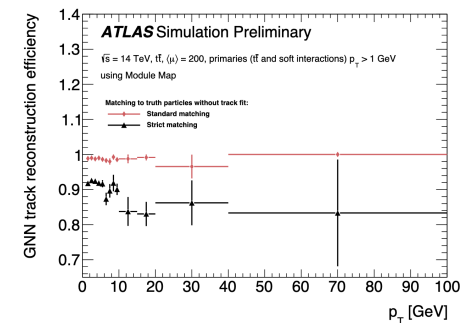
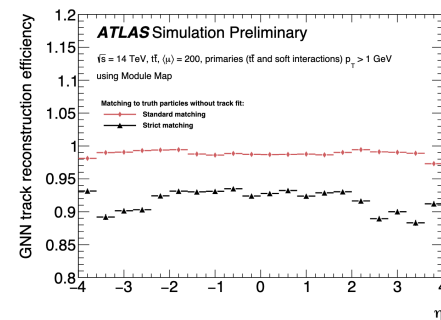
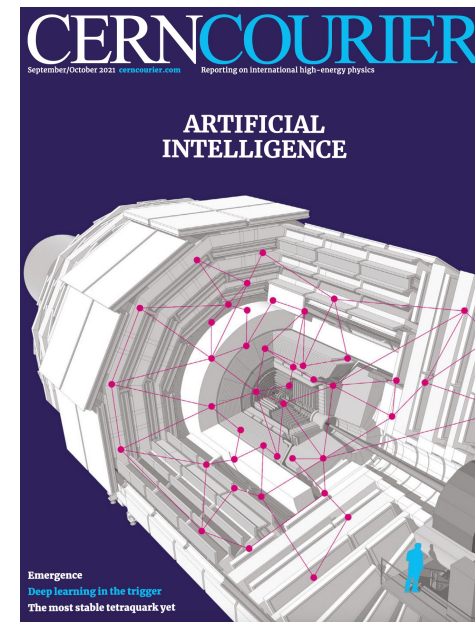
Collaboration L2IT & ExatTrkX



⇒ Construct a GNN-based track reconstruction algorithm for ATLAS Itk



- ⇒ First results on ITk published in 2022 more than encouraging
- ⇒ GNN-based algorithms now appear as competitive solutions
- ⇒ Put into production (ACTS) for the HL-LHC



[ATLAS ITk Track Reconstruction with a GNN-based pipeline, C.Rougier et al., CTD 2022](#)
[Graph Neural Network track reconstruction for ATLAS ITk, D. Murnane et al., IML 2022](#)

Motivation

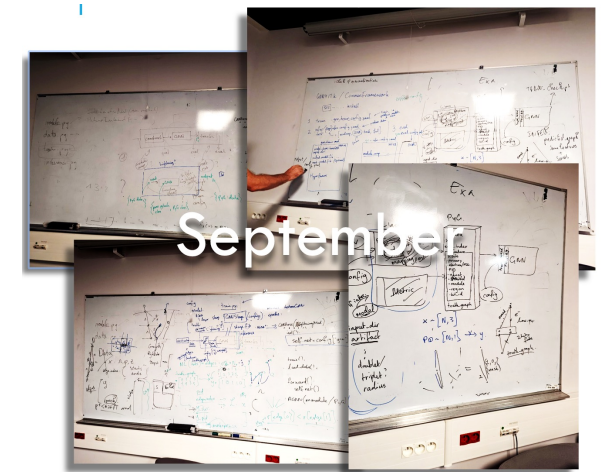
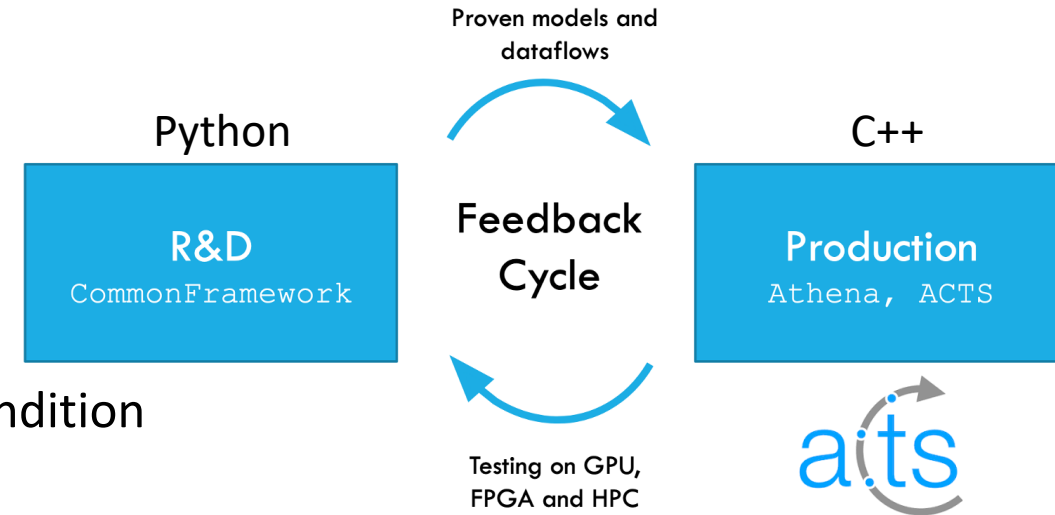
ML R&D in HEP take a LOT of time and (GPU) resources

- ML R&D: Models and hyperparameters exploration
- Data: Complexity of the simulated detector data in HL-LHC condition (Pythia + GEANT4 + ITk geometry)
- (International) Collaboration:
 - Do we speak the same language (share the same semantic) ?
 - Data format
 - Experimental conditions
 - Performance metrics definition & implementation
- Reproducibility: Lack of traceability of experimental conditions
- Software quality: Messy idea implemented, no documentation, no test, etc...

Improve the process => save some R&D time

- ⇒ Get faster for research
- ⇒ Use less (GPU) resources and do more sober R&D

⇒ Need to develop of a common R&D framework

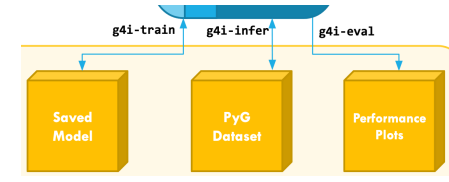


gnn4ITk common framework (developped with D. Murnane, LBNL)

PyG Data

<p>Spacepoint features</p> <ul style="list-style-type: none"> • r, ϕ, z, \dots (required) • Hit ID (required) • Cell shape • Module • Region • Shared • Ghost 	<p>Track features</p> <ul style="list-style-type: none"> • Truth graph (required) • Particle ID (required) • Primary • p_T • Num hits • PDG ID <p>Graph Features</p> <ul style="list-style-type: none"> • Edge index (required) • Y (required)
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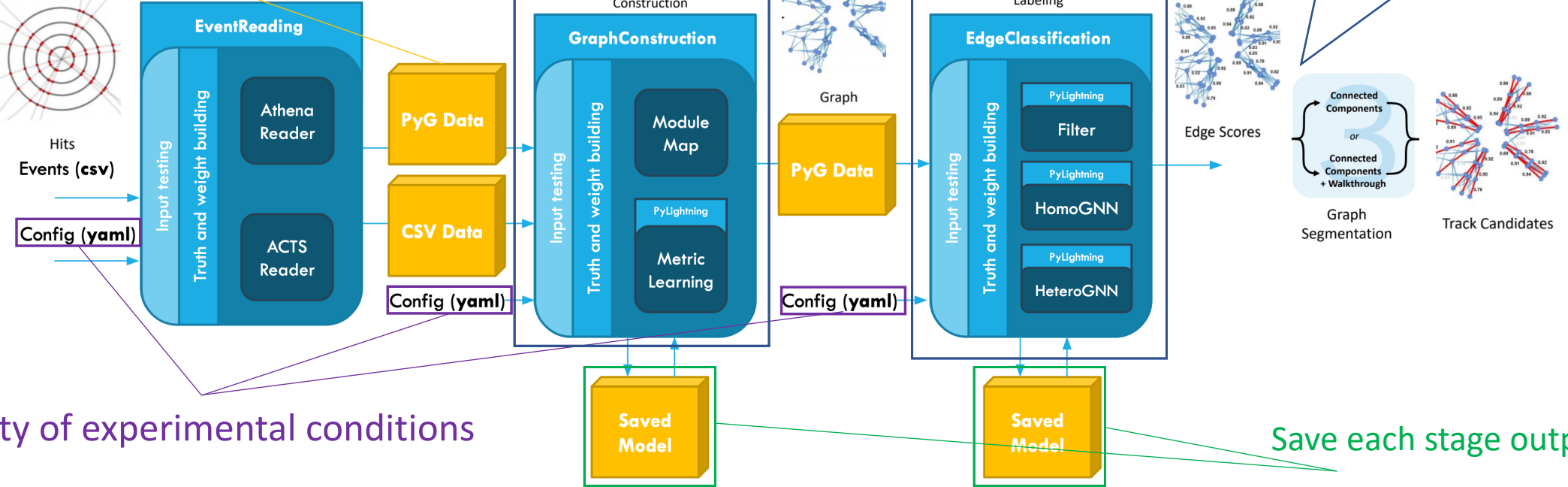
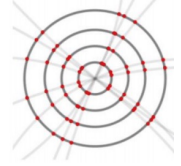
Modularity: Pipeline divided in stage
 Each stage can include several models
 It can be *train* or *infer* or *evaluate*
 Share the same evaluation method



Also try to:

- Be Easy to use
- Be clean (keep messy dev & random ideas out of the repository)
- Be well documented
- To Include Test

Well defined common data format



Traceability of experimental conditions

Save each stage output

Next steps...

- Reproduce performance on approval dataset
- Use the framework (and no more home code) for the next studies
- Publish it with open data demo
- Towards a common platform / tools for Geometric ML R&D in HEP ?