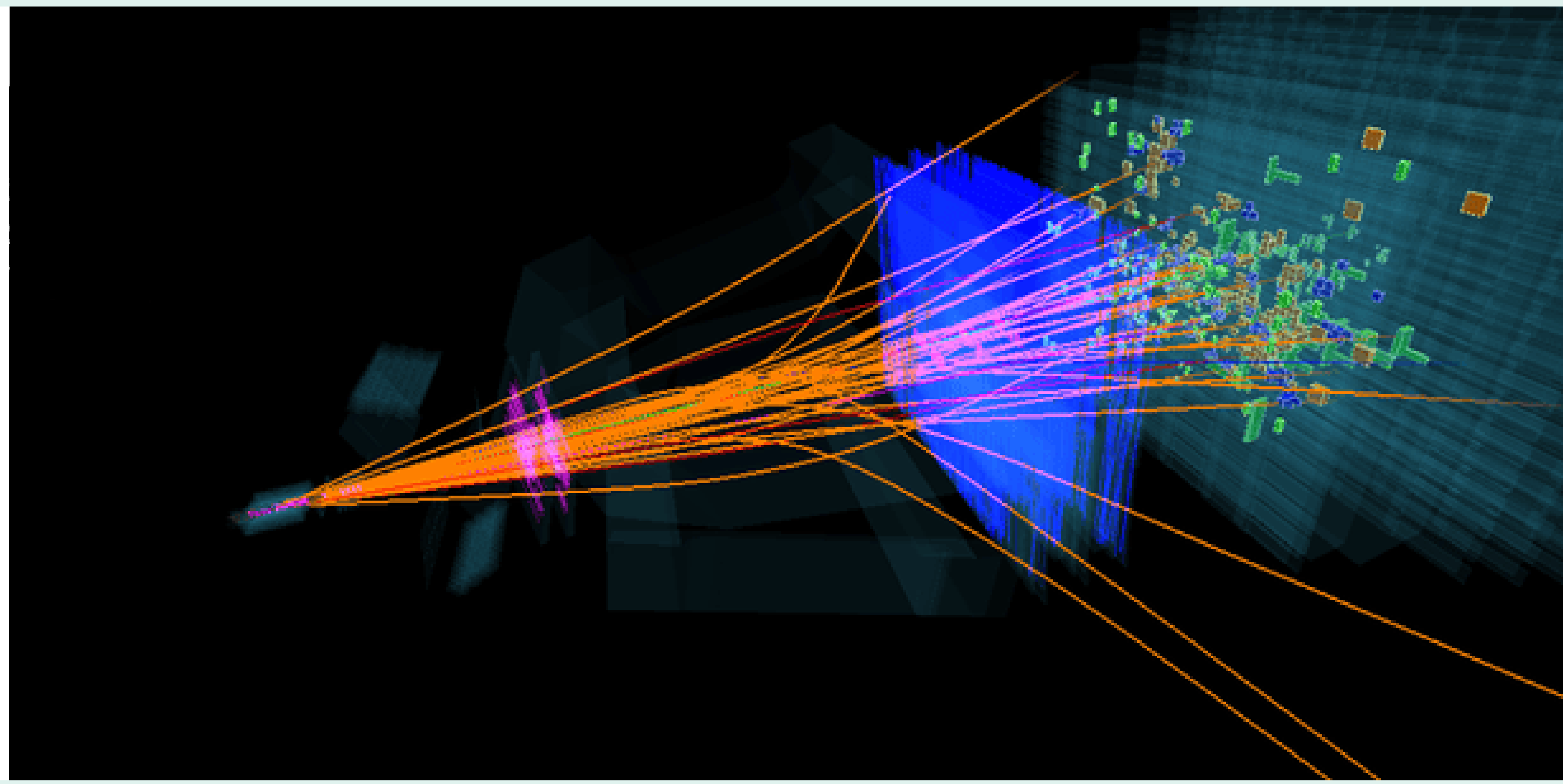
GNN for Deep Full Event Interpretation and hierarchical reconstruction of heavy-hadron decays in proton-proton collisions



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CHEP 2023

Norfolk
11 May 2023

Outlook

Performance

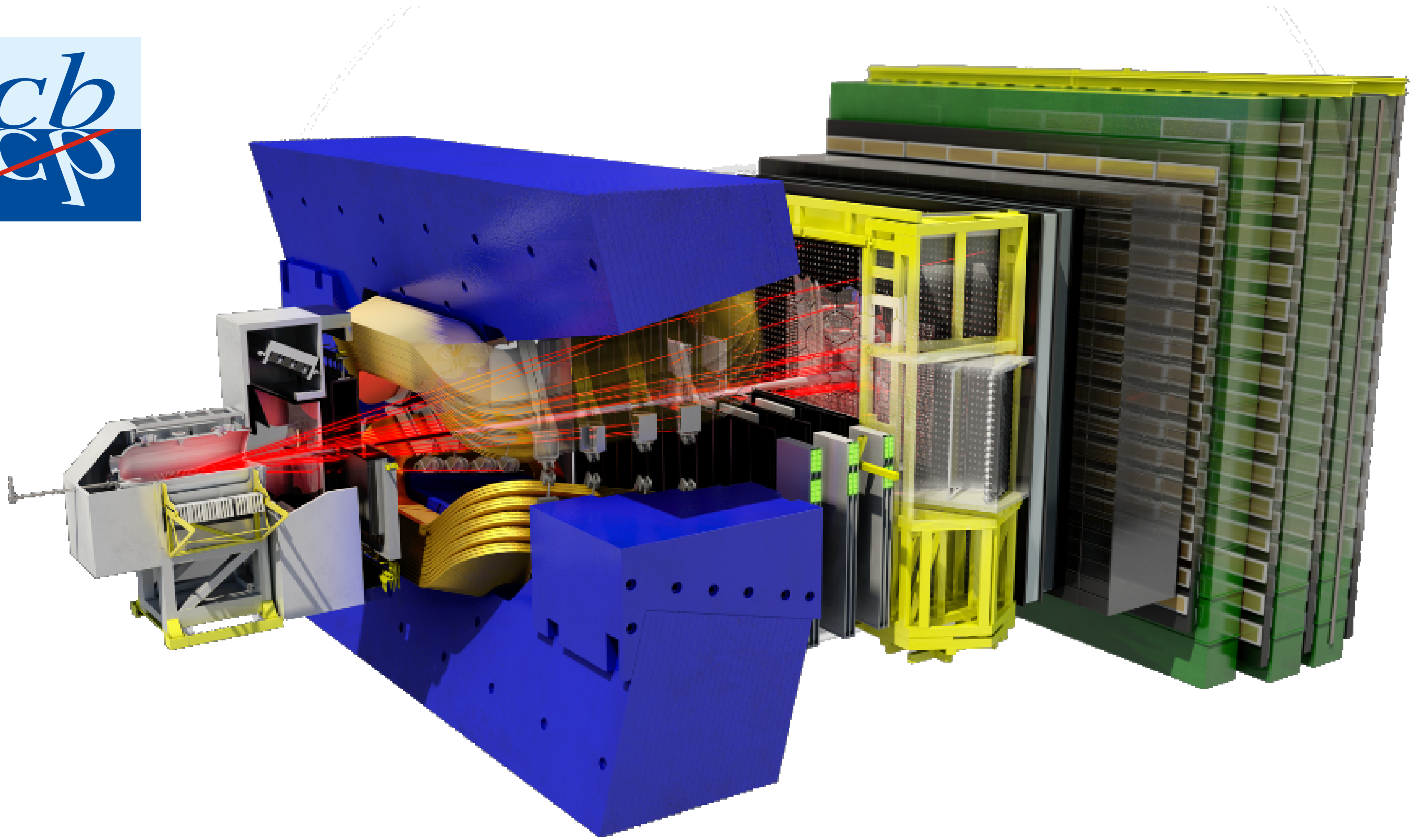
The algorithm

Motivation

[DFEI-arXiv:2304.08610]

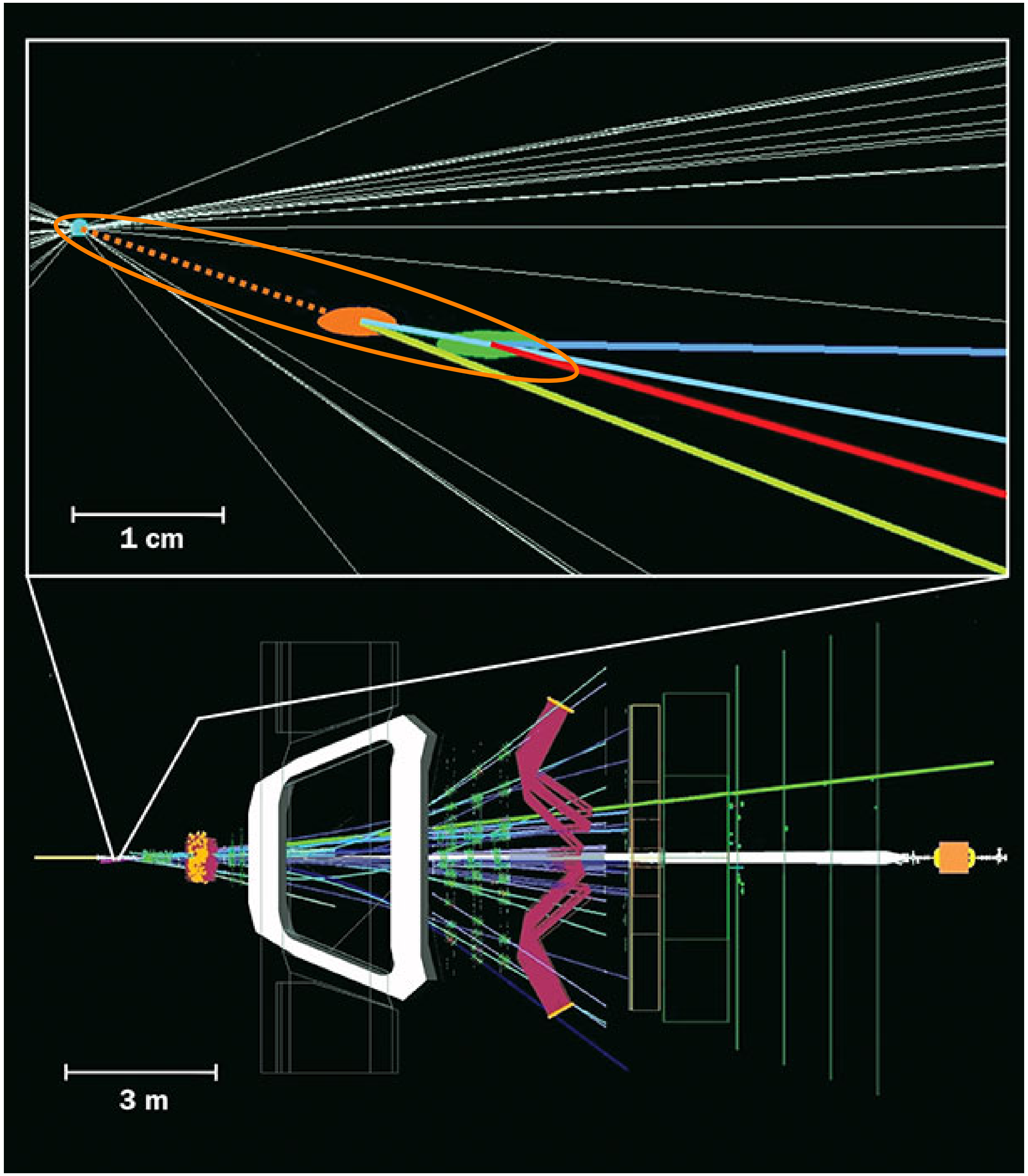
The LHCb detector

Single-arm forward spectrometer, studying the **decays of beauty and charm hadrons**.



Excellent vertexing capabilities, momentum resolution and PID performance.

LHCb trigger



Current trigger

OR between decay modes

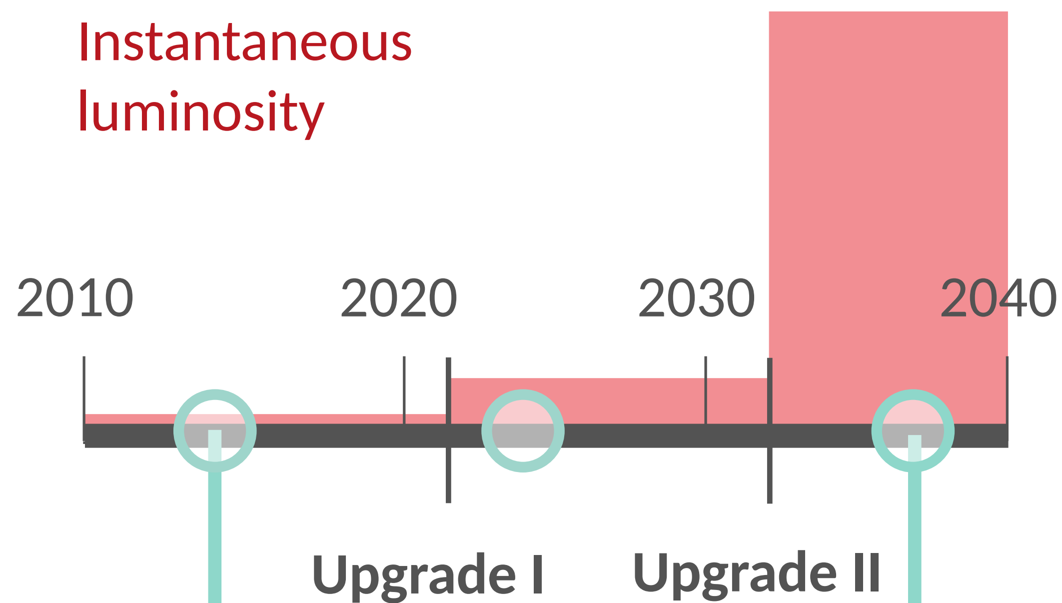
Store whole event*

*some efforts at reducing size exist:[JINST 14 (2019) 04, P04006]

Evolution in the LHCb trigger

≪ 1 signal
~ 50 tracks

Which events are interesting?
⇓
Trigger strategy:
signal based.



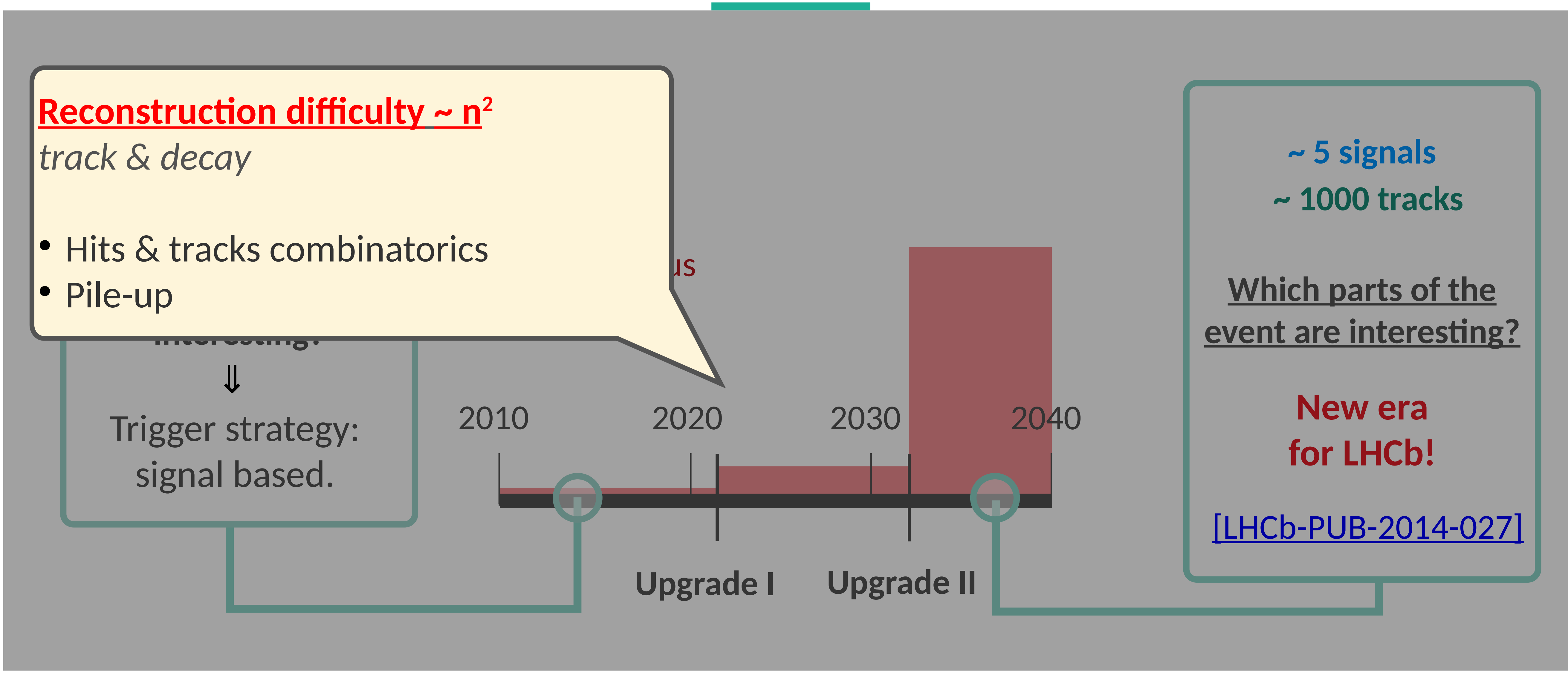
~ 5 signals
~ 1000 tracks

Which parts of the event are interesting?

New era for LHCb!

[LHCb-PUB-2014-027]

Evolution in the LHCb trigger



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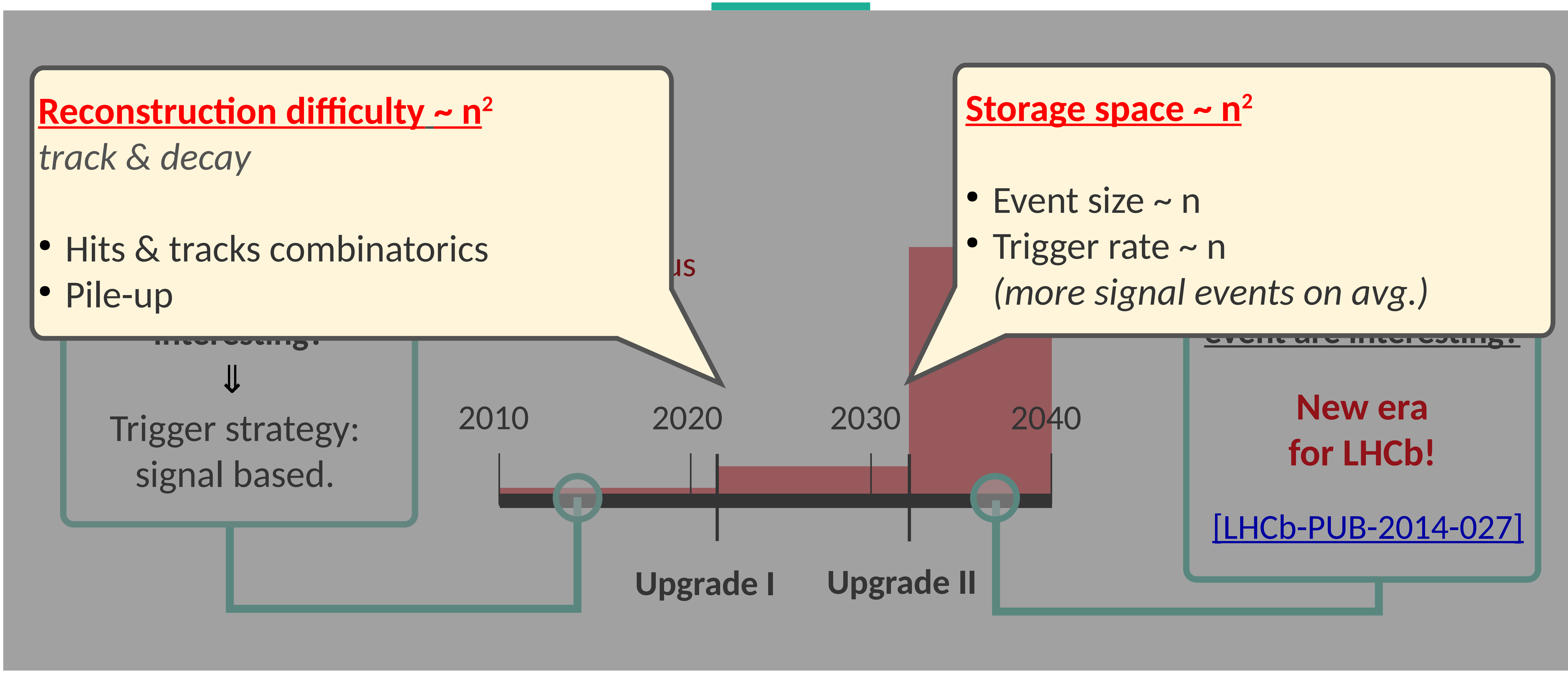

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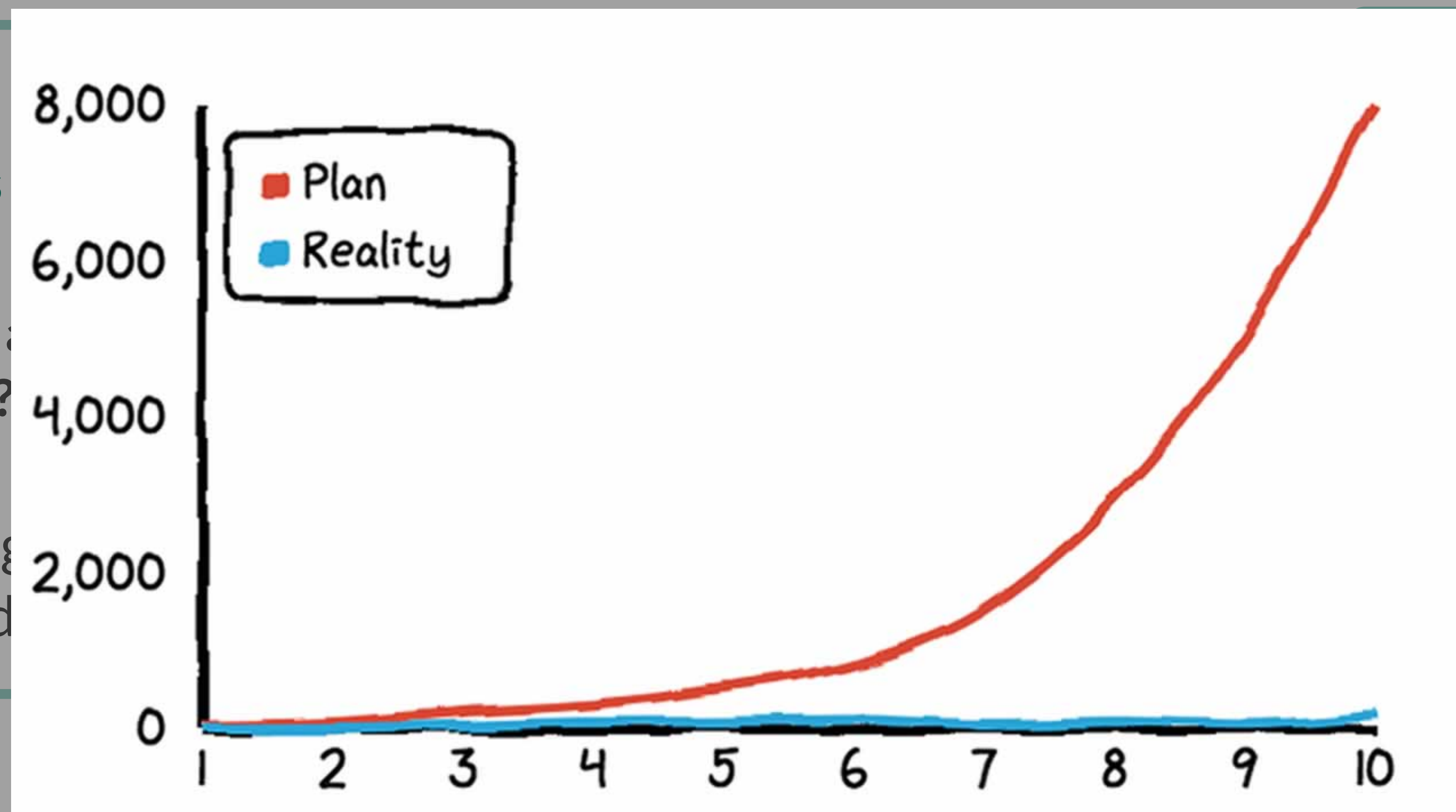

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Cb-PUB-2014-027]

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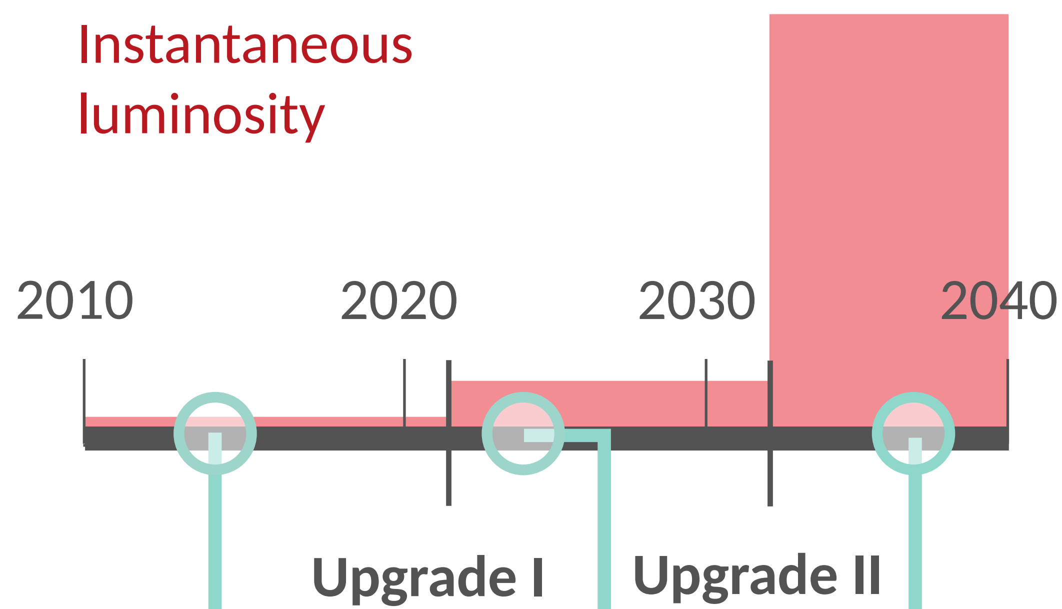


~ 5 signals
~ 1000 tracks

Which parts of the event are interesting?

New era for LHCb!

[LHCb-PUB-2014-027]

~ 0.5 signals
~ 140 tracks

Fully software trigger, CPU + GPU [JINST 14 (2019) 04, P04006].
➡ Online alignment and calibration,
offline-quality online reconstruction.

Key developments that enable more ambitious trigger strategies.

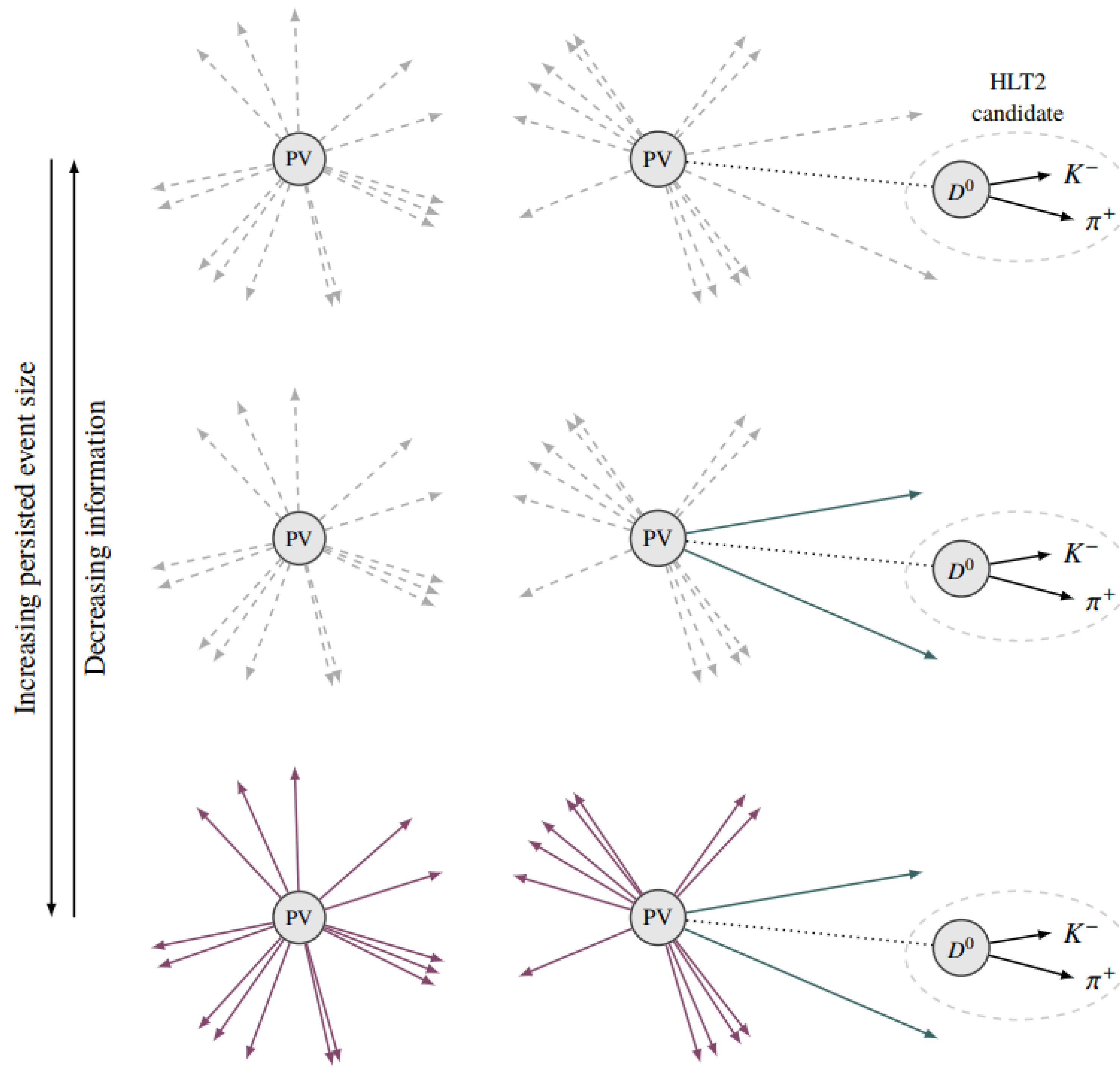
Event information and size

Full event

vs

disk space

Crucial to have <u>all</u> tracks of decay (finding resonances…)

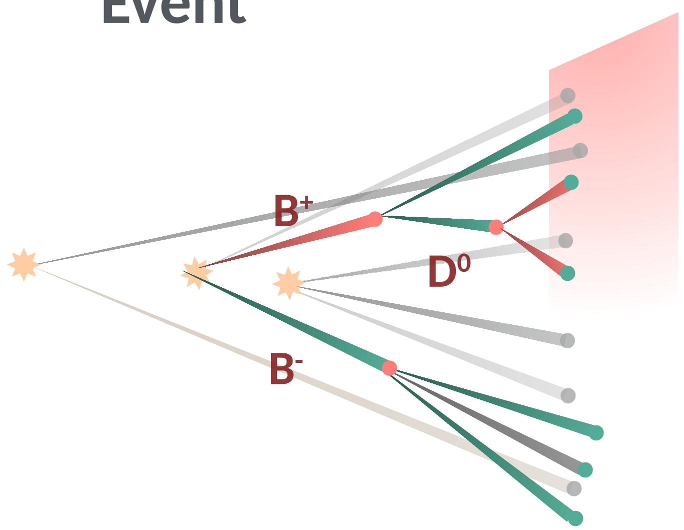


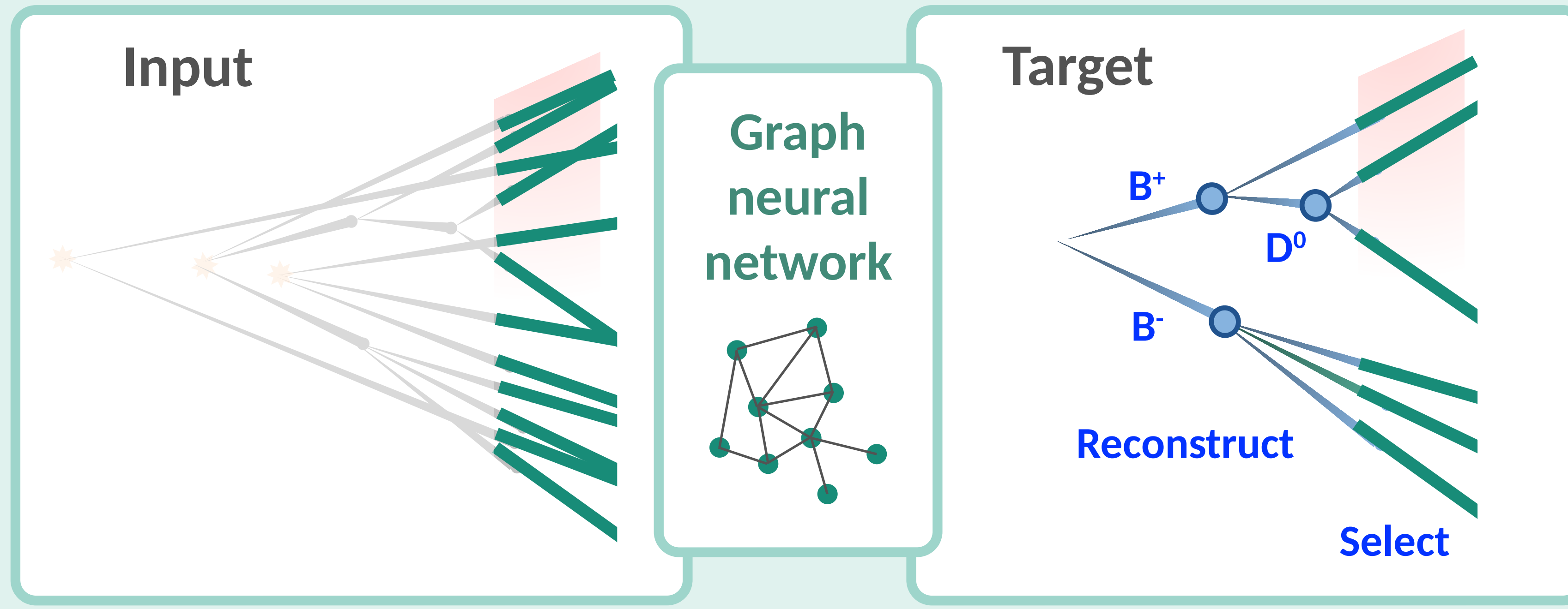
Facing the new era with machine learning

Novel approach proposed

DFEI:
Deep-learning based Full Event Interpretation

Event





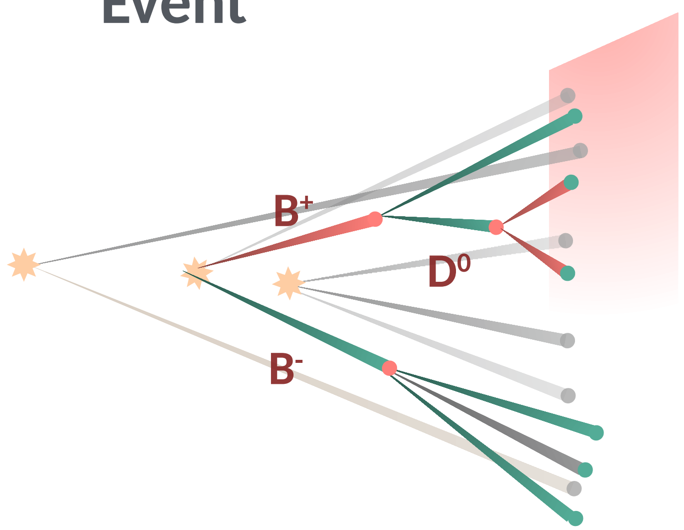
On High-level trigger stage

LHCb HLT 2

reconstructed tracks available

Facing the new era with machine learning

Event



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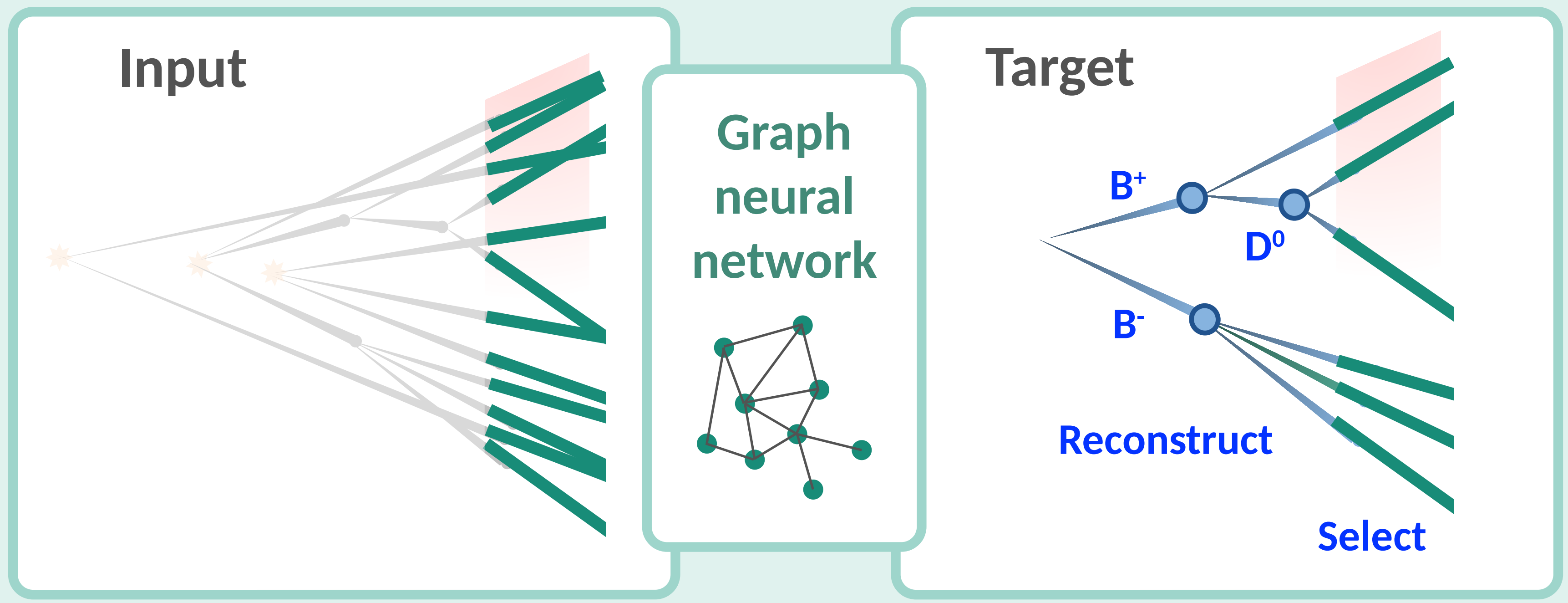


Similar developments in other experiments

Full Event Interpretation algorithm at an e+e- collider
[Comput.Softw.Big Sci. 3 (2019) 1 6], BELLE2-MTHESIS-2020-006].

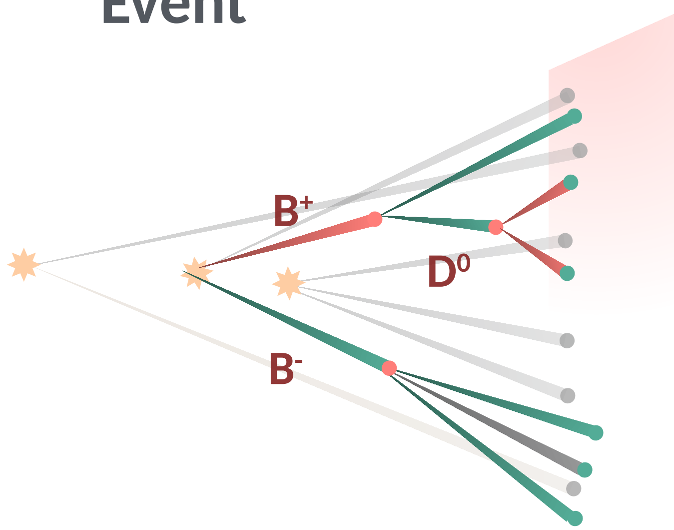
GNNs for trigger purposes
[see e.g. Eur.Phys.J.C 81 (2021) 5, 381, Frontiers in Big Data 3 (2021) 44].

Facing the new era with machine learning

Novel approach proposed

DFEI:
Deep-learning based
Full Event Interpretation

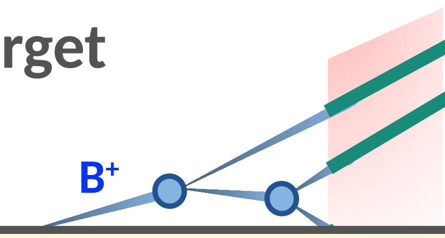
Event



Input

Graph
neural

Target



Belle II case	LHCb case
Only $B^0/B^{\pm}$ hadrons.	All b,c-hadron species.
e+e- environment.	pp environment.
Hermetic detector.	Non-hermetic detector.

ger.

Similar developments ... her experiments

Full Event Interpretation algorithm at an e+e- collider
[Comput.Softw.Big Sci. 3 (2019) 1 6], BELLE2-MTHESIS-2020-006].

GNNs for trigger purposes
[see e.g. Eur.Phys.J.C 81 (2021) 5, 381, Frontiers in Big Data 3 (2021) 44].

Outlook

Performance

The algorithm

First prototype of DFEI for LHCb, focused on b-hadron decays and charged stable particles.

Motivation

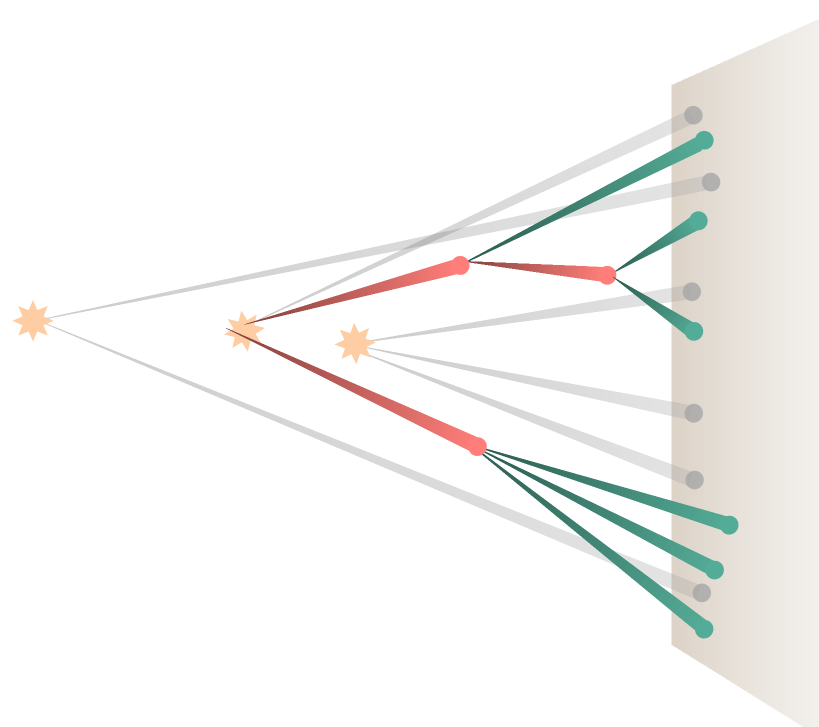
Decays and graph structures

Event

Global: event information
nTracks, …

Nodes: track variables
momentum, (PID), …

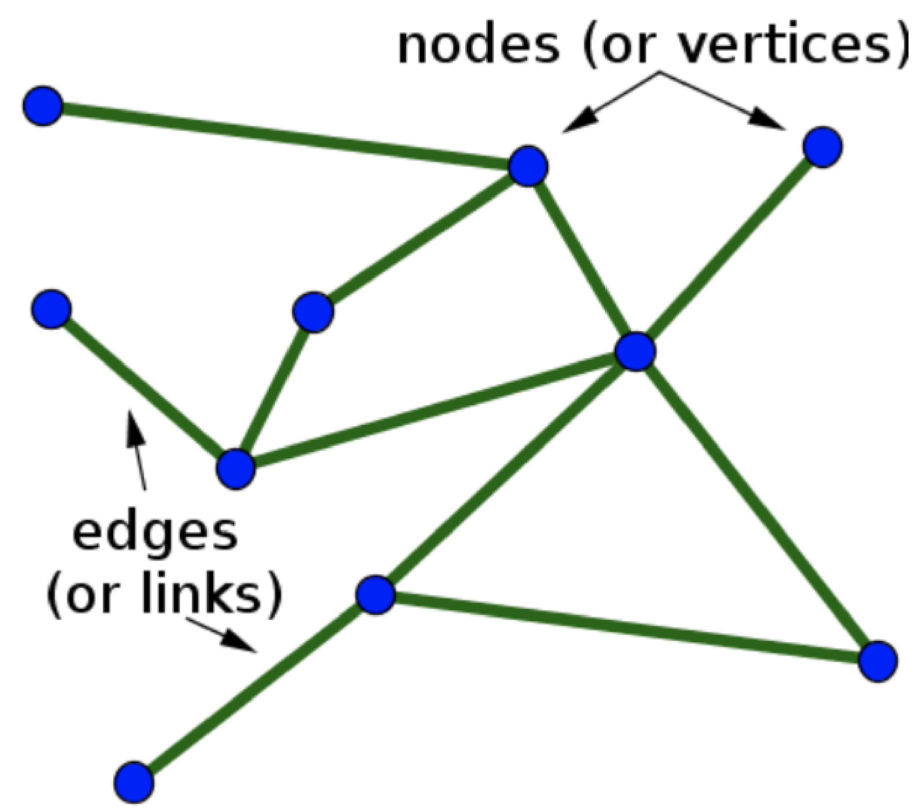
Edges (# nodes^2!): track *relations*
angle, DOCA, …



Graph structures

Representation of objects
with relations

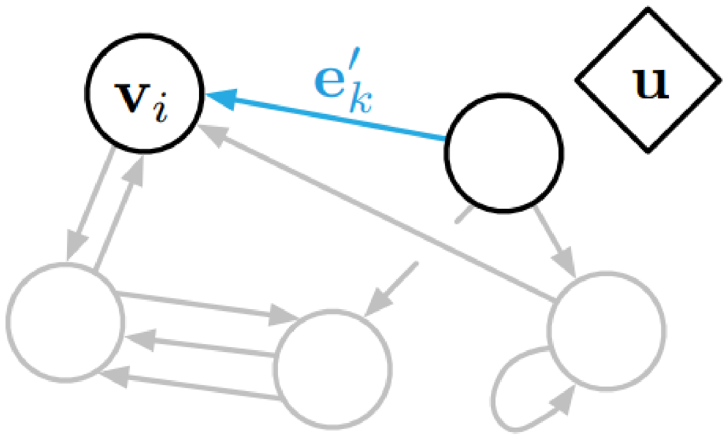
Arbitrary, sparse/dense relations



GNN training

e' v' u' updated (by DNN) e, v, u

BLUE updated by **BLACK** not utilizing **GREY**



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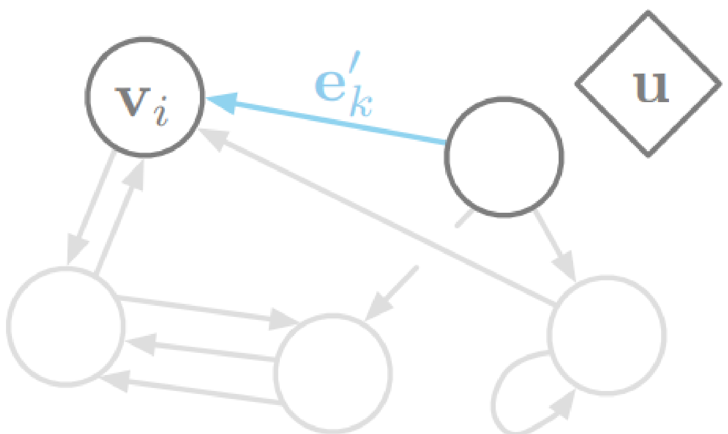

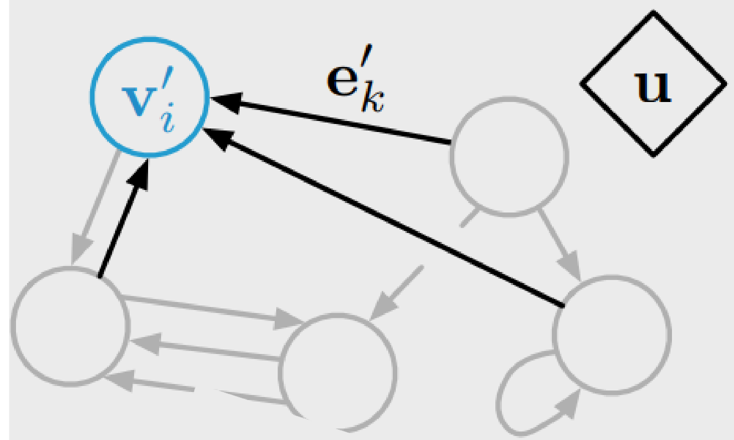

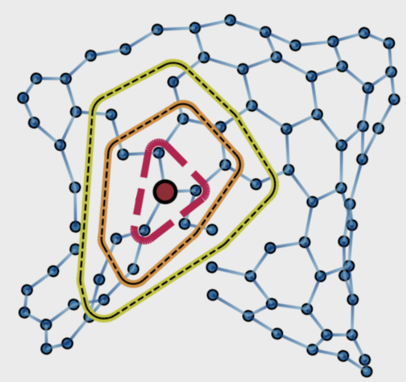

Node Aggregation

Aggregating information from neighbors

GNN training

e' v' u' updated (by DNN) e, v, u

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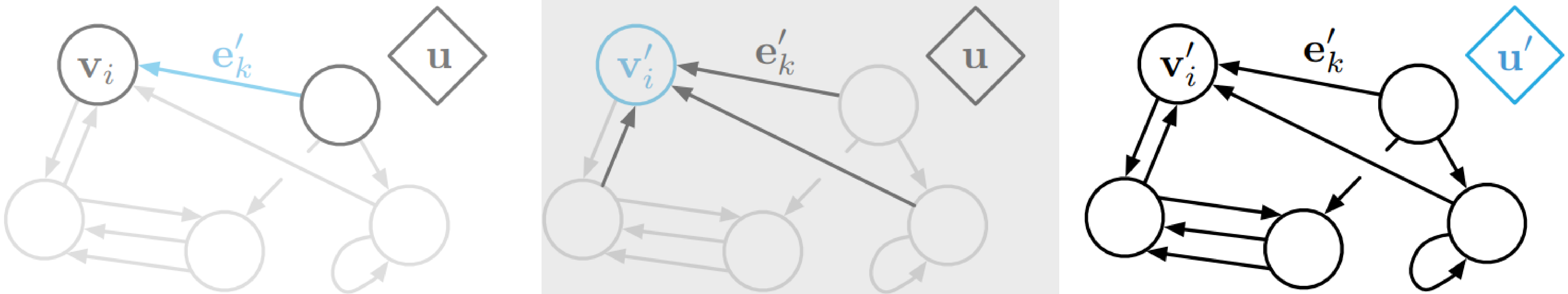

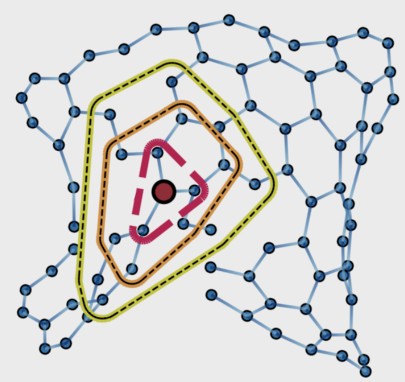

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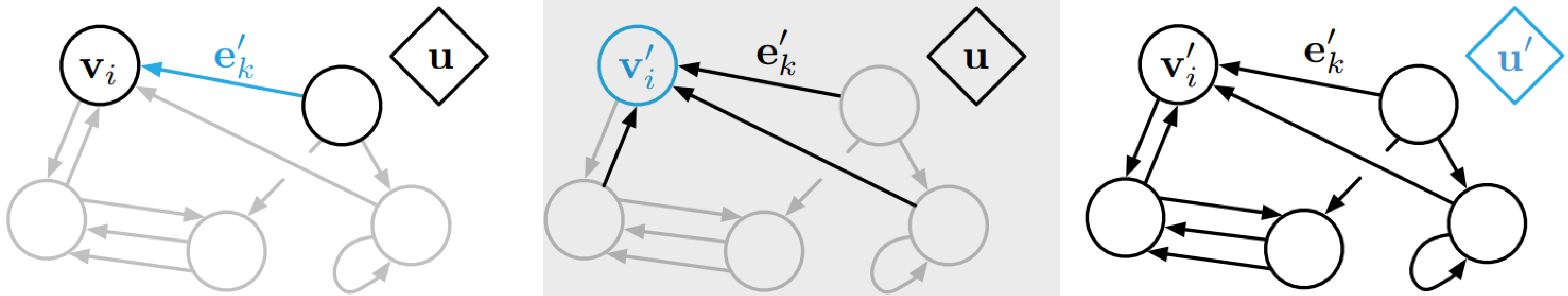

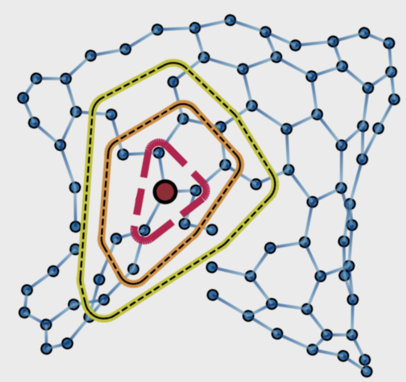

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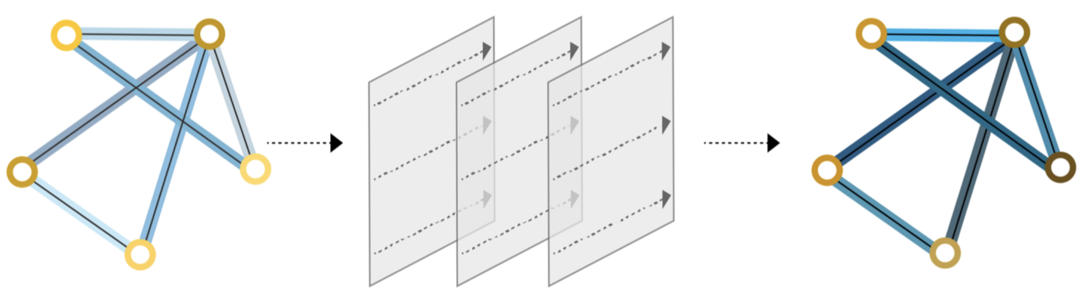


Transformation of graph

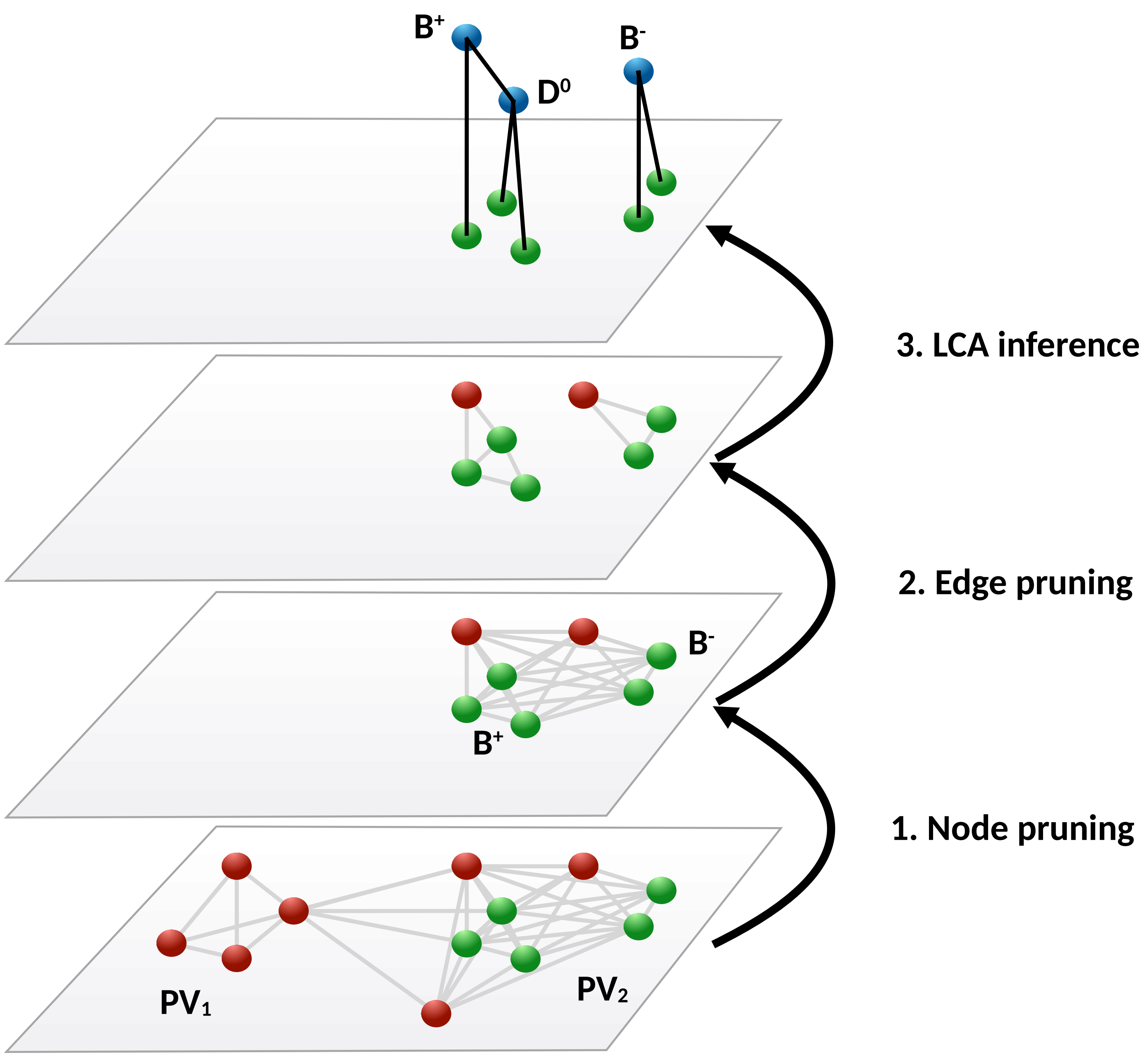
Input graph (X), return graph (y)

Output graph

Node/edge/global features

Different interpretations depending on application

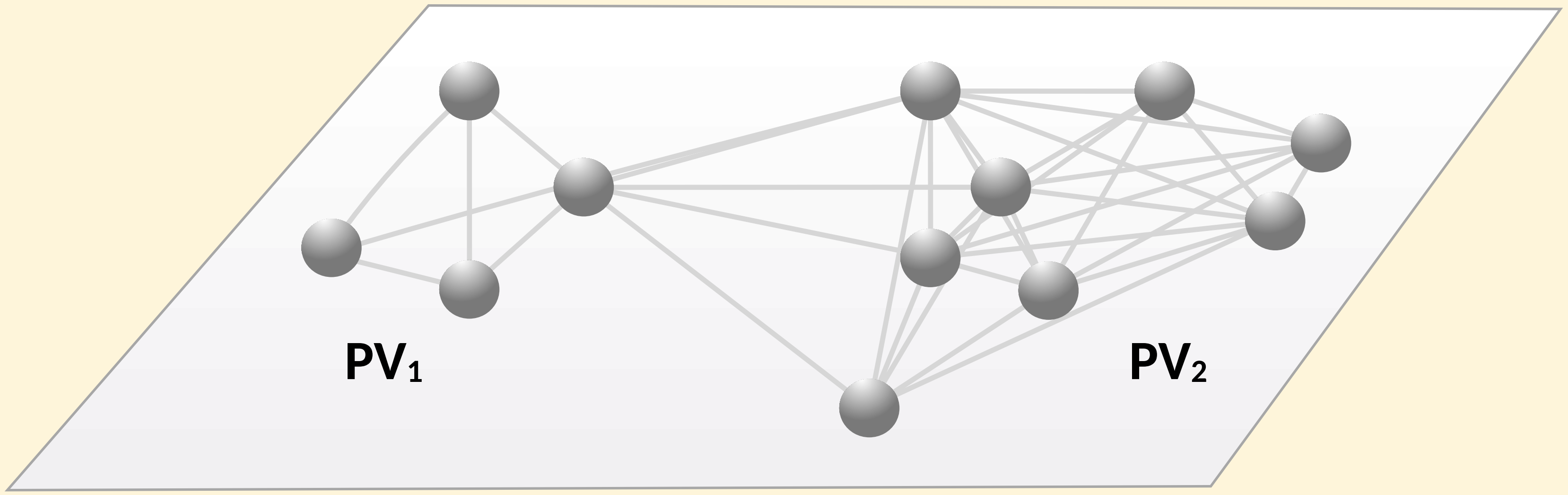
DFEI Overview



Input graph construction

Nodes: all charged particles **~140**

Edges: topologically close particles **~10 000**

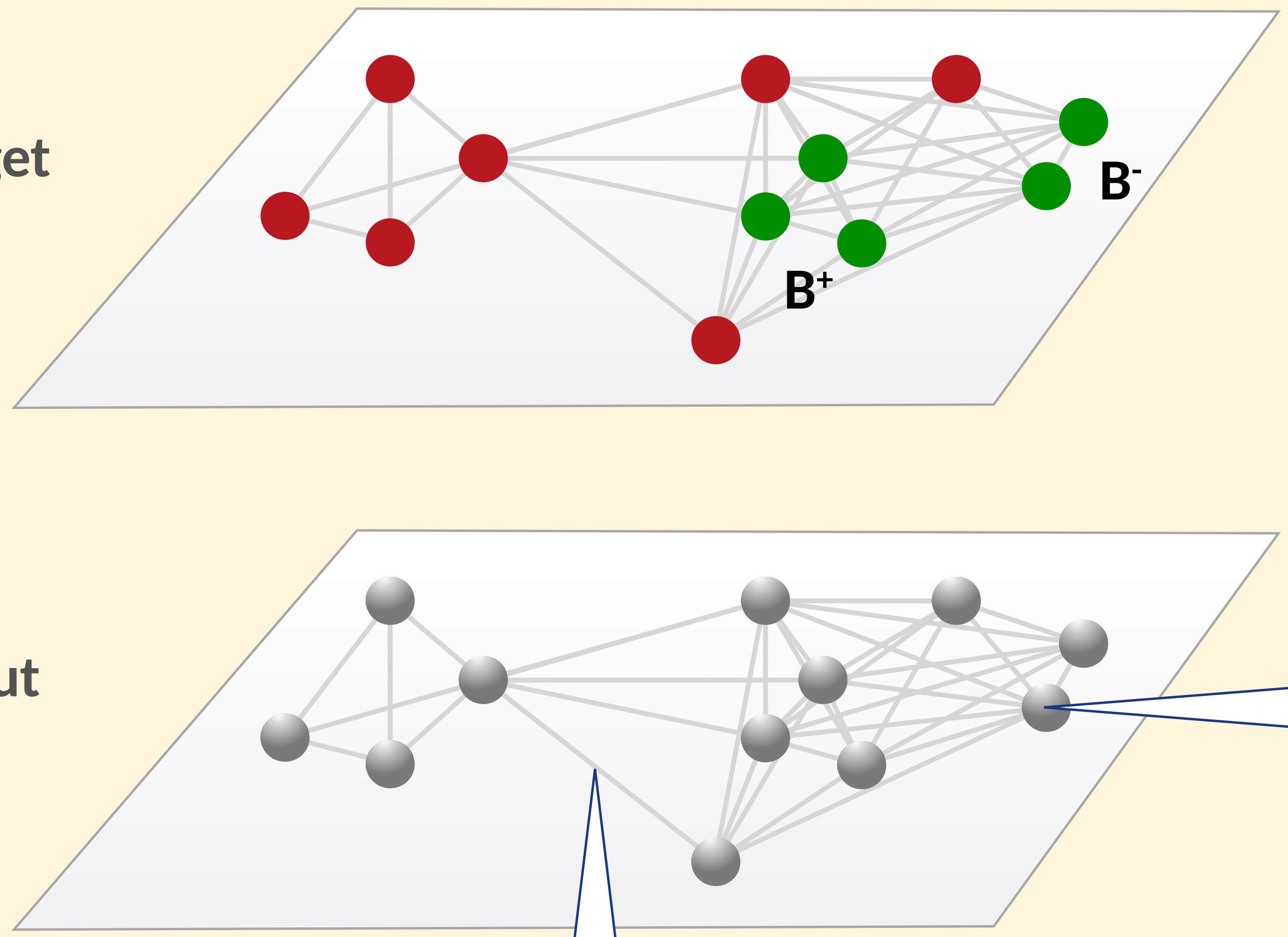


1st module: node pruning

Signal nodes: particles from a b-hadron (any of them)
Background nodes: particles from the rest of the event

Target

Input



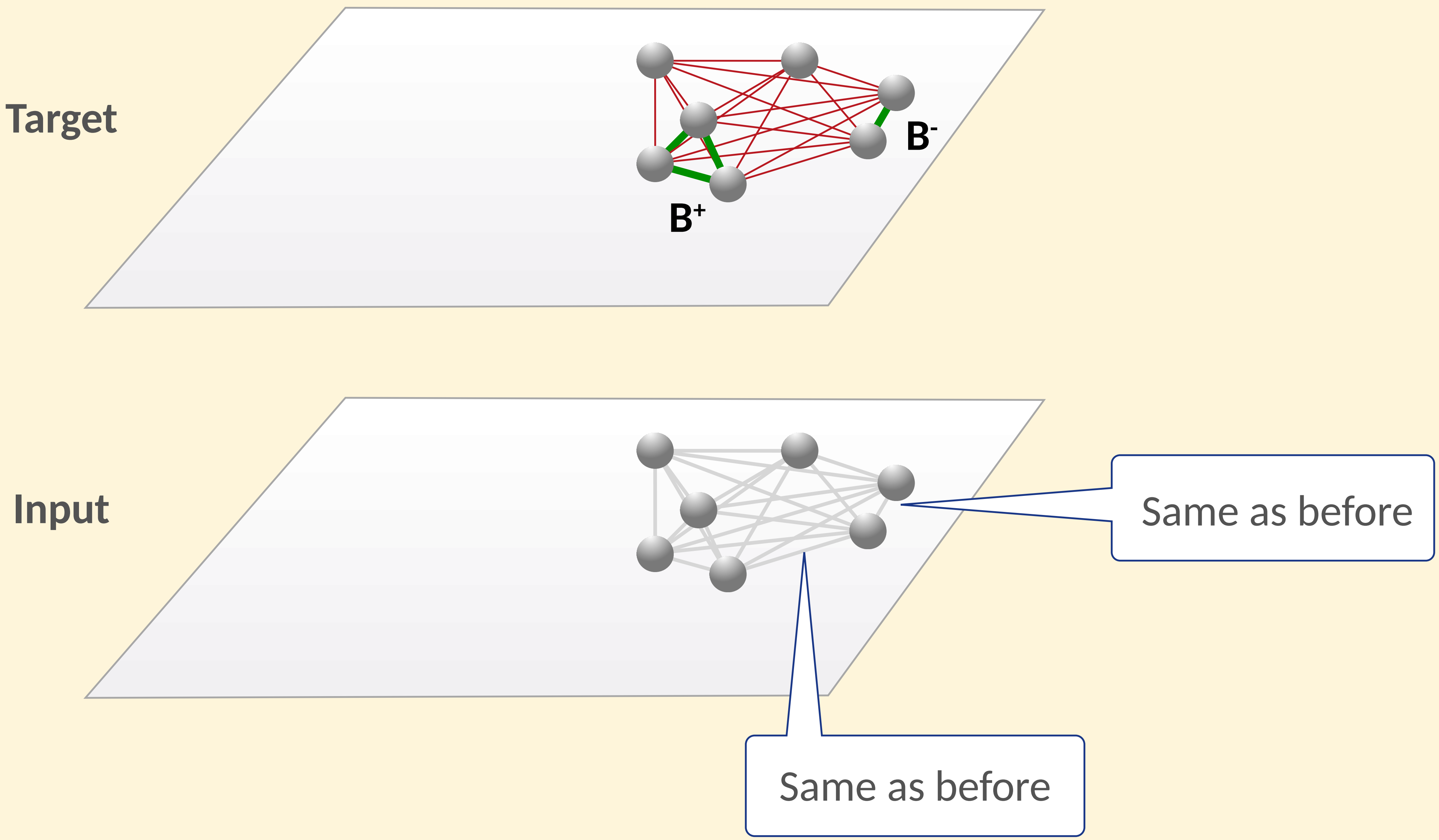
pT: transverse momentum
ETA: pseudorapidity
PV: associated primary vertex
IP: impact parameter with respect to the PV
q: charge

pT, ETA, IP, q

Opening angle, distance (between origins) along the beam axis, “transverse distance” (see backup), from same PV (boolean)

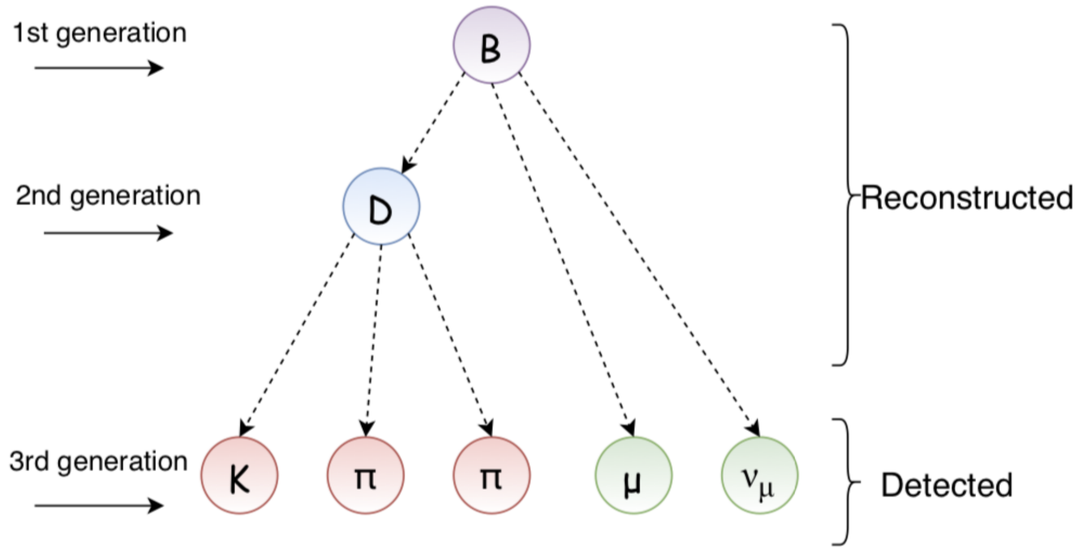
2^{nd} module: edge pruning

Signal edges: pairs of particles with the same b-hadron ancestor
Background edges: any other pair of particles



3rd module: Lowest Common Ancestor (LCA) inference

From [BELLE2-MTHESIS-2020-006]:
(see also [James Kahn et al 2022 Mach. Learn.: Sci. Technol. 3 035012])



Adjacency Matrix

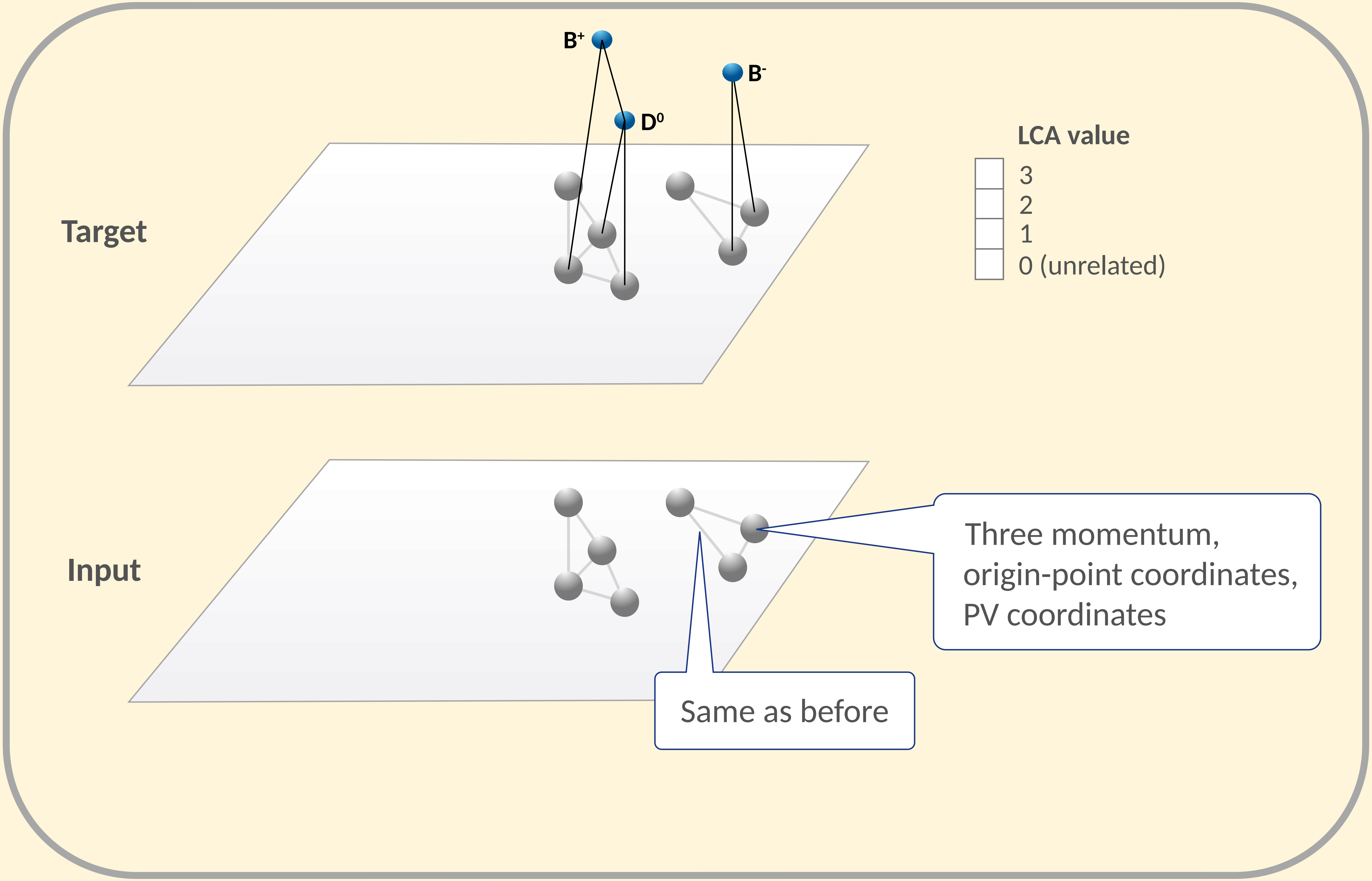
	B	D	K	π	π	μ	ν_μ
B	0	1	0	0	0	1	1
D	1	0	1	1	1	0	0
K	0	1	0	0	0	0	0
π	0	1	0	0	0	0	0
π	0	1	0	0	0	0	0
μ	1	0	0	0	0	0	0
ν_μ	1	0	0	0	0	0	0

LCA Matrix

	K	π	π	μ	ν_μ
K	0	1	1	2	2
π	1	0	1	2	2
π	1	1	0	2	2
μ	2	2	2	0	2
ν_μ	2	2	2	2	0

Problem reduced to **multi-class classification on edges**.

3^{rd} module: Lowest Common Ancestor (LCA) inference



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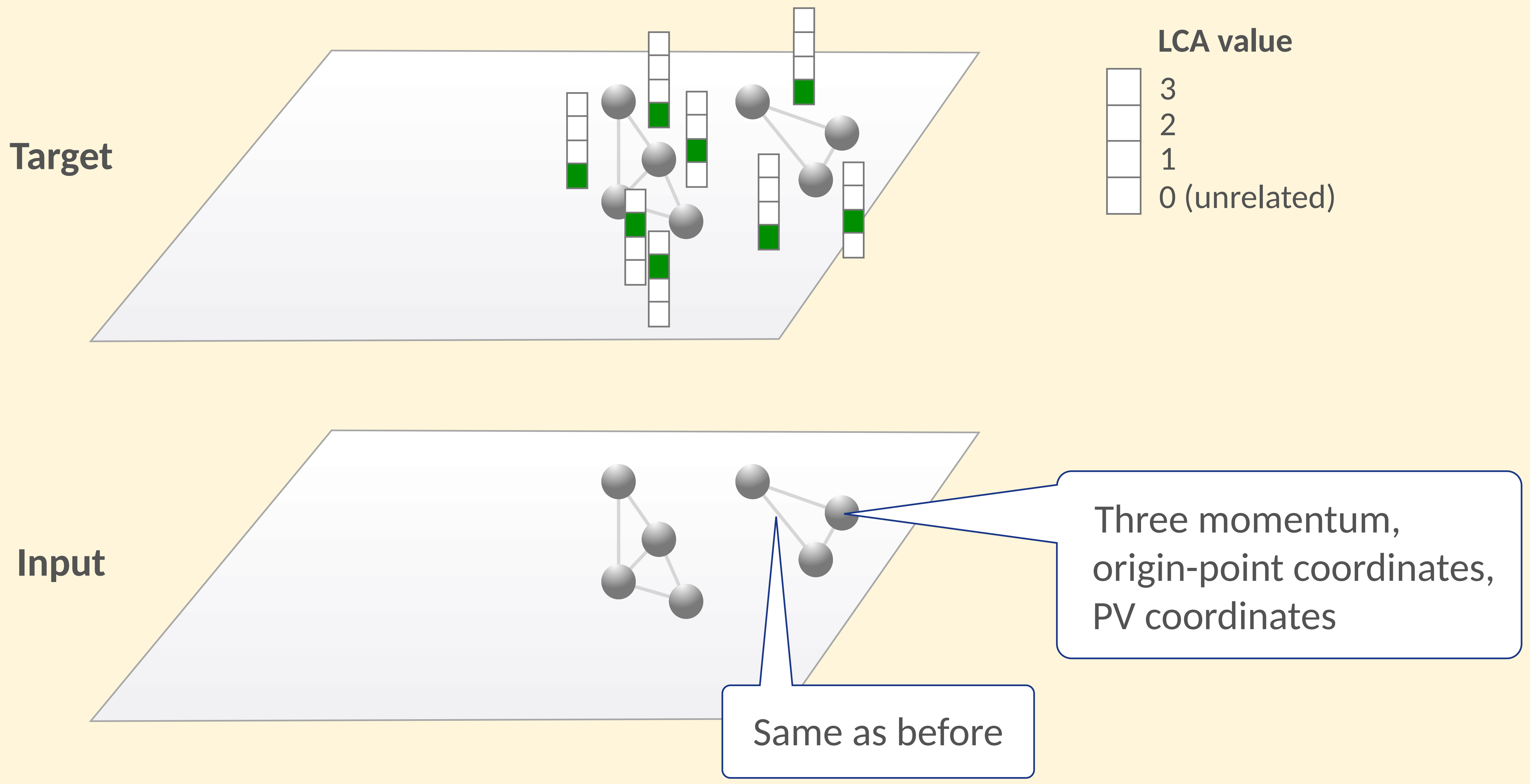
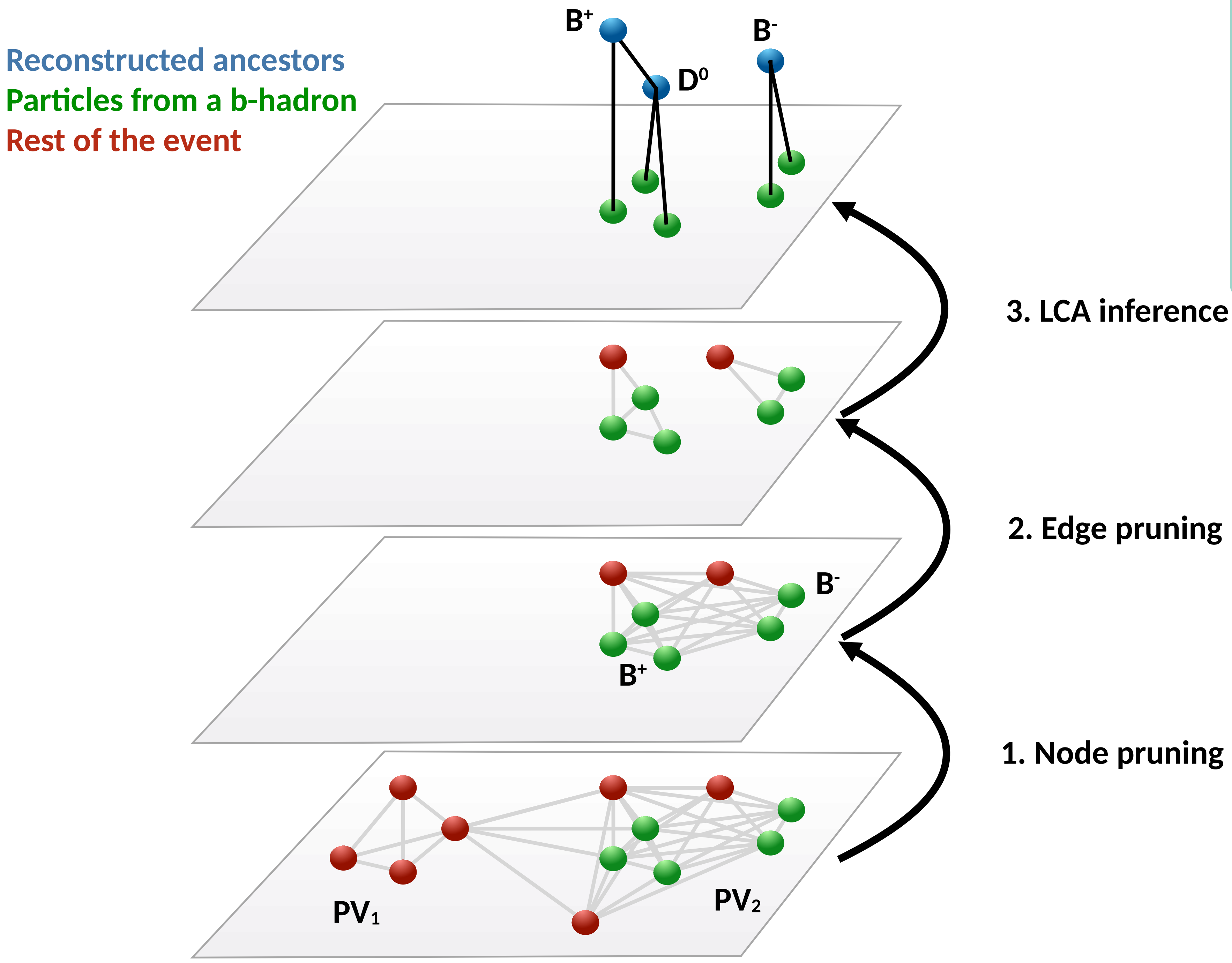


Overview and training

Dataset:

- PYTHIA-based simulation, Run 3-like conditions, approximated emulation of LHCb reconstruction.
- Events required to contain at least one b-hadron (inclusive decay).



Public dataset

for benchmarking & comparison

10.5281/zenodo.7799169

Training

Dataset:
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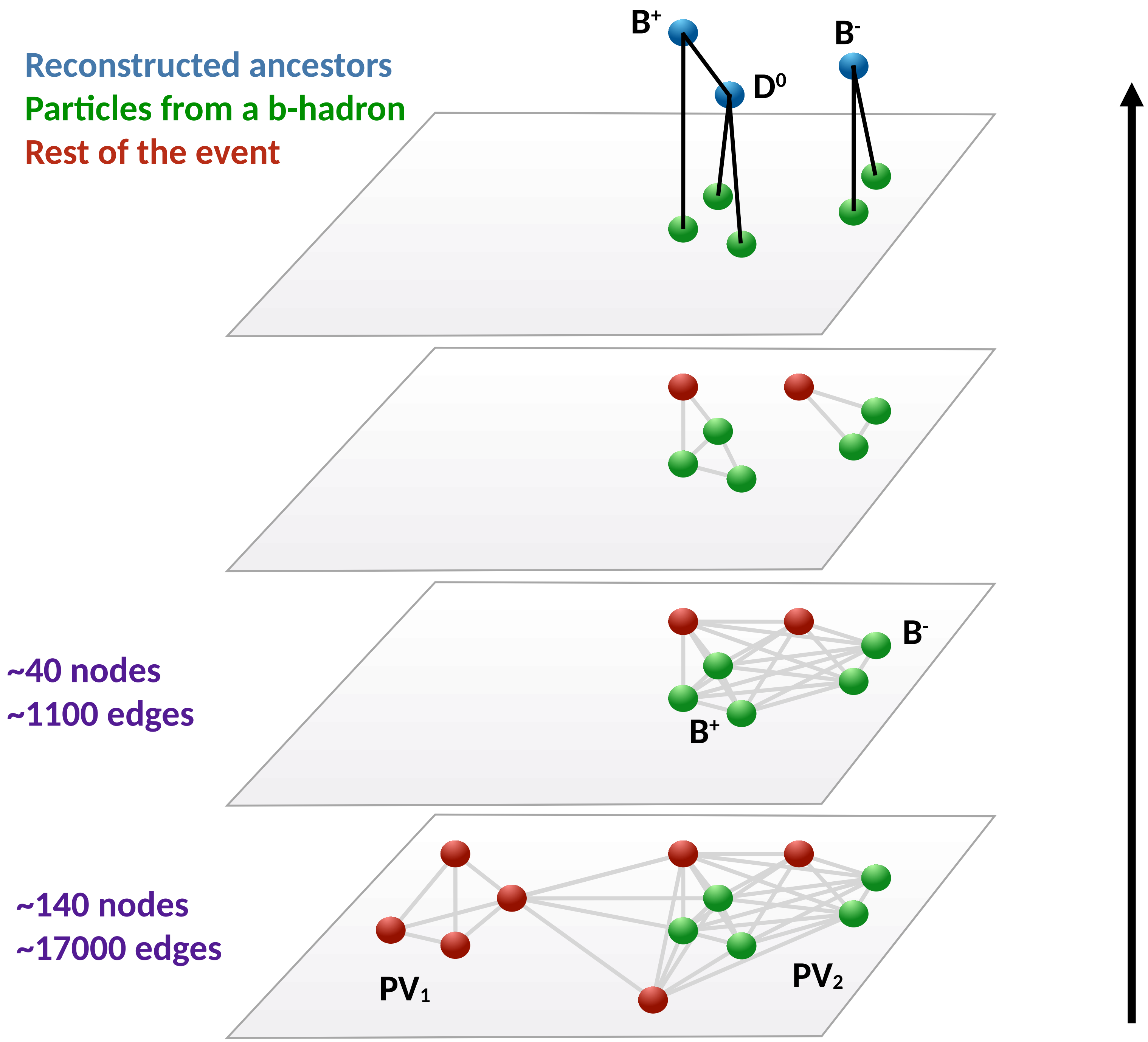

Filter the data

Choose a threshold
(~99% signal eff.)

Train the node-pruning module

Training

Dataset:
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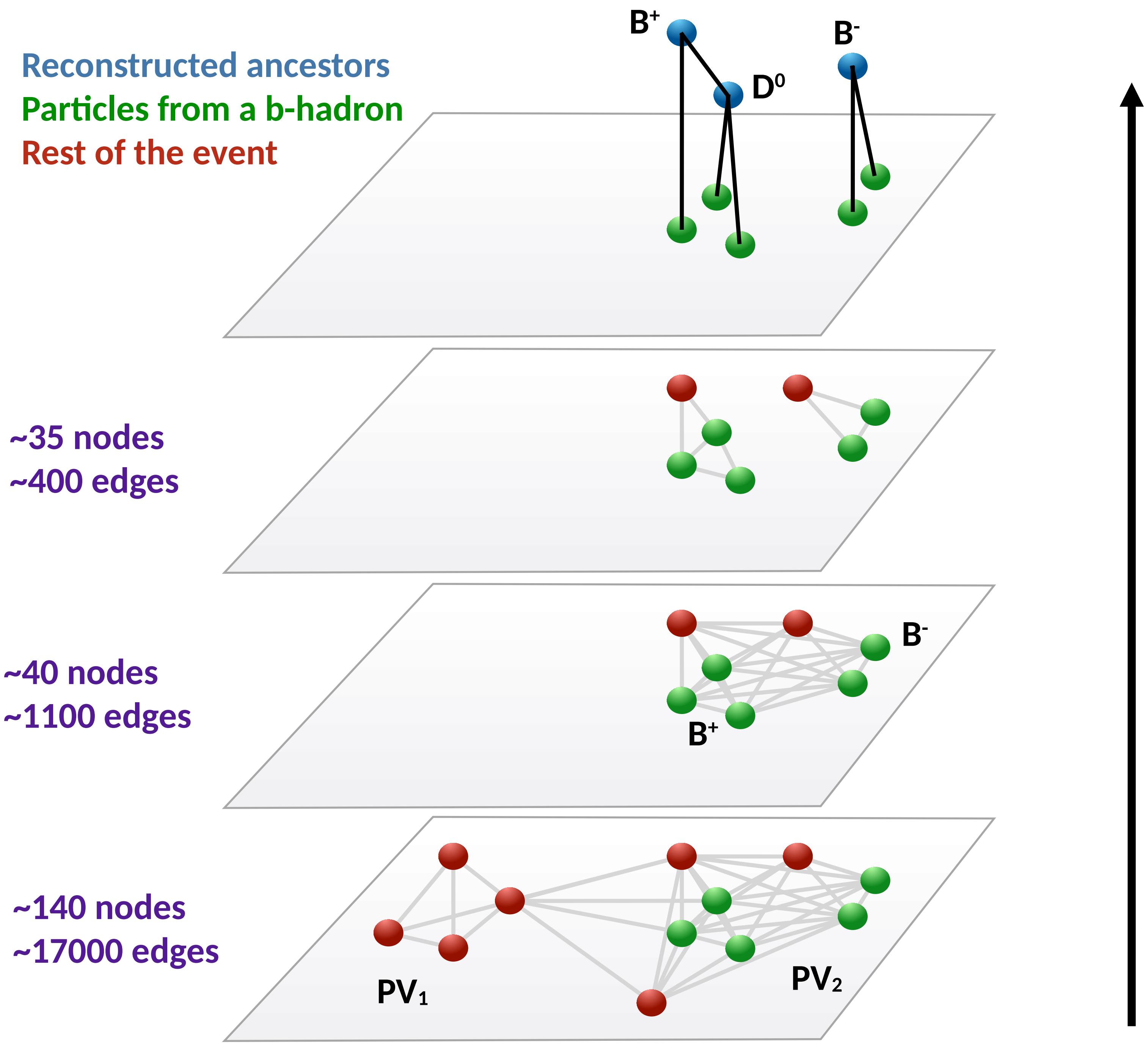

Filter the data

Choose a threshold
(~99% signal eff.)

Train the edge-pruning module

Filter the data

Choose a threshold
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Train the node-pruning module

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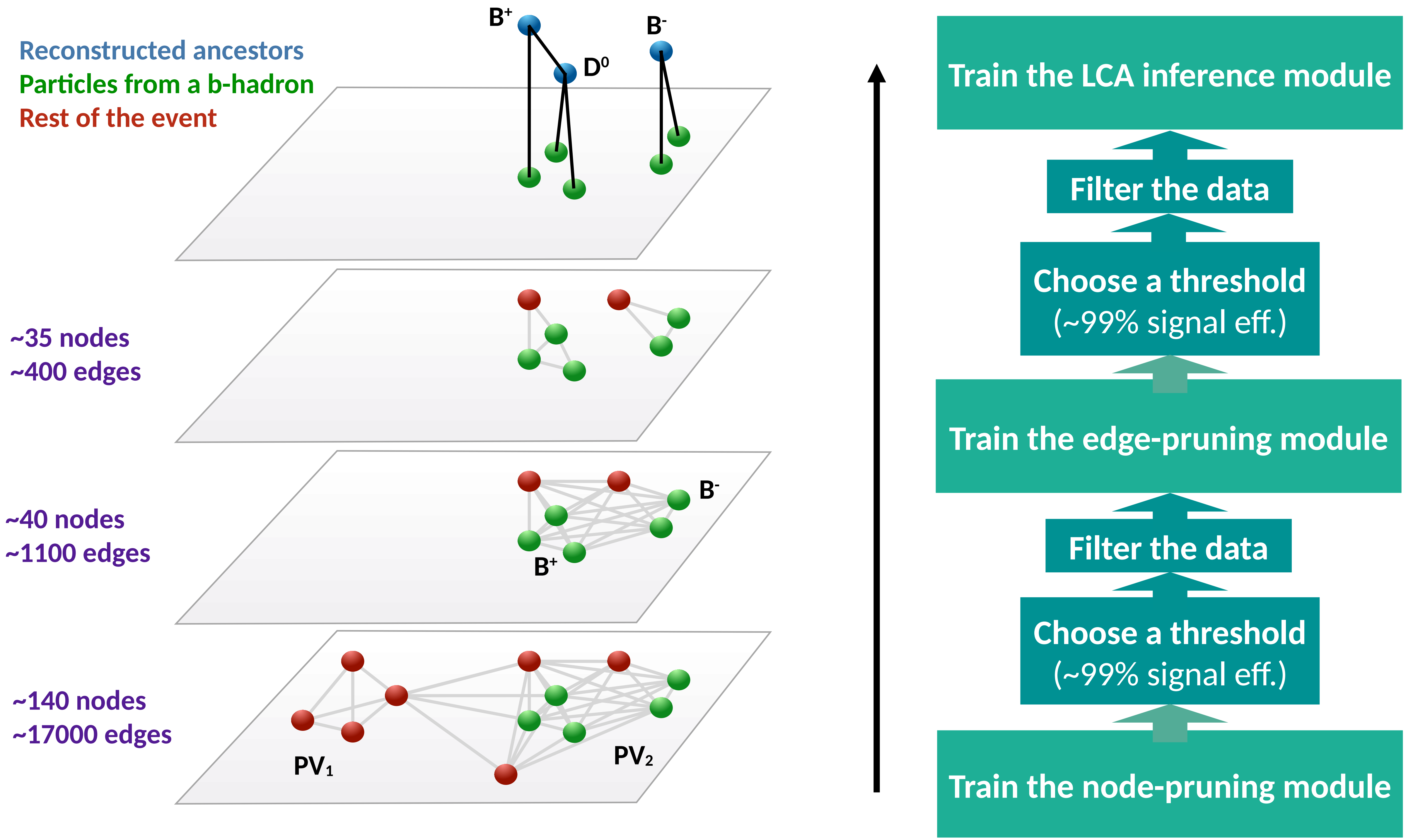


Outlook

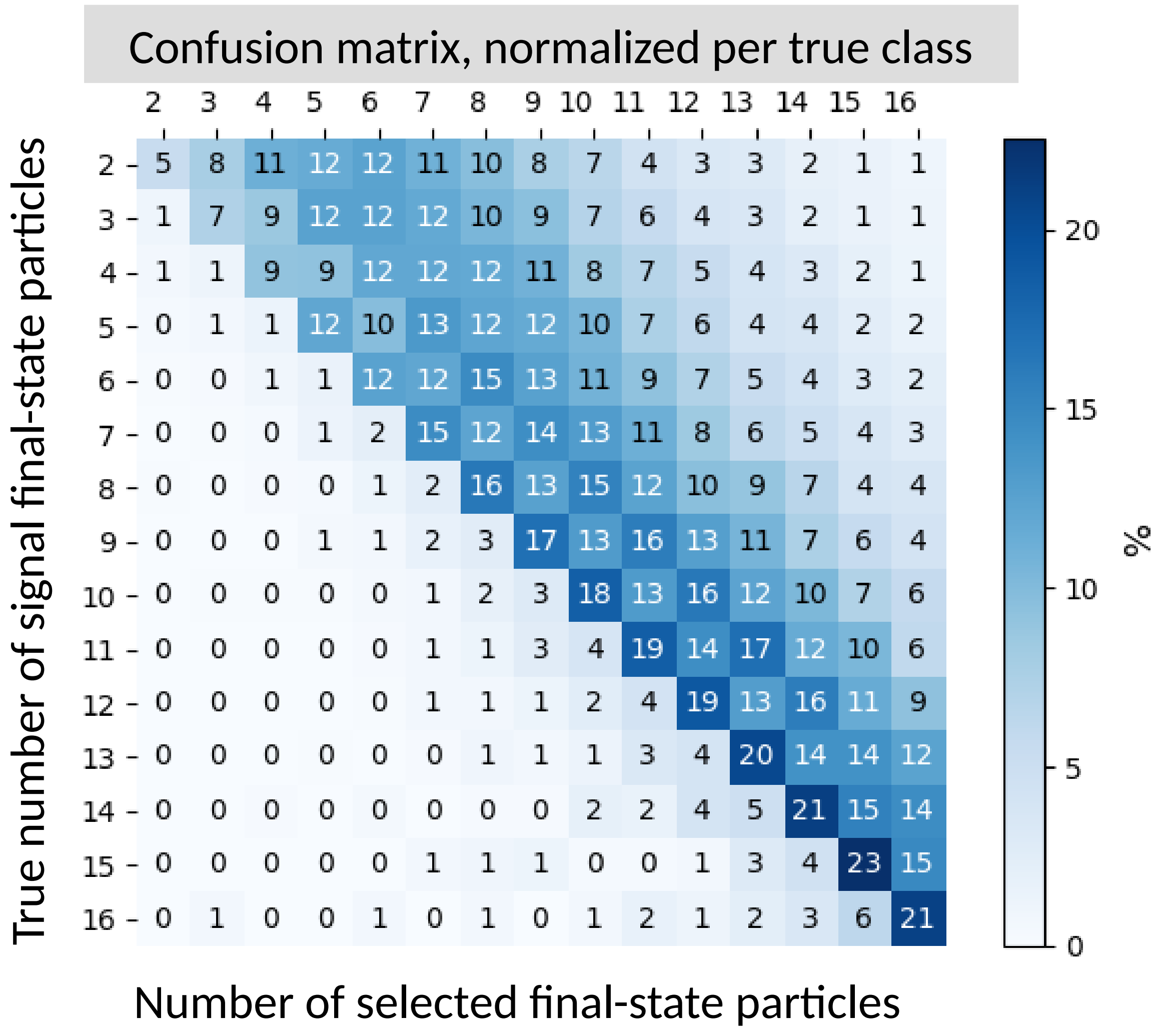
Performance

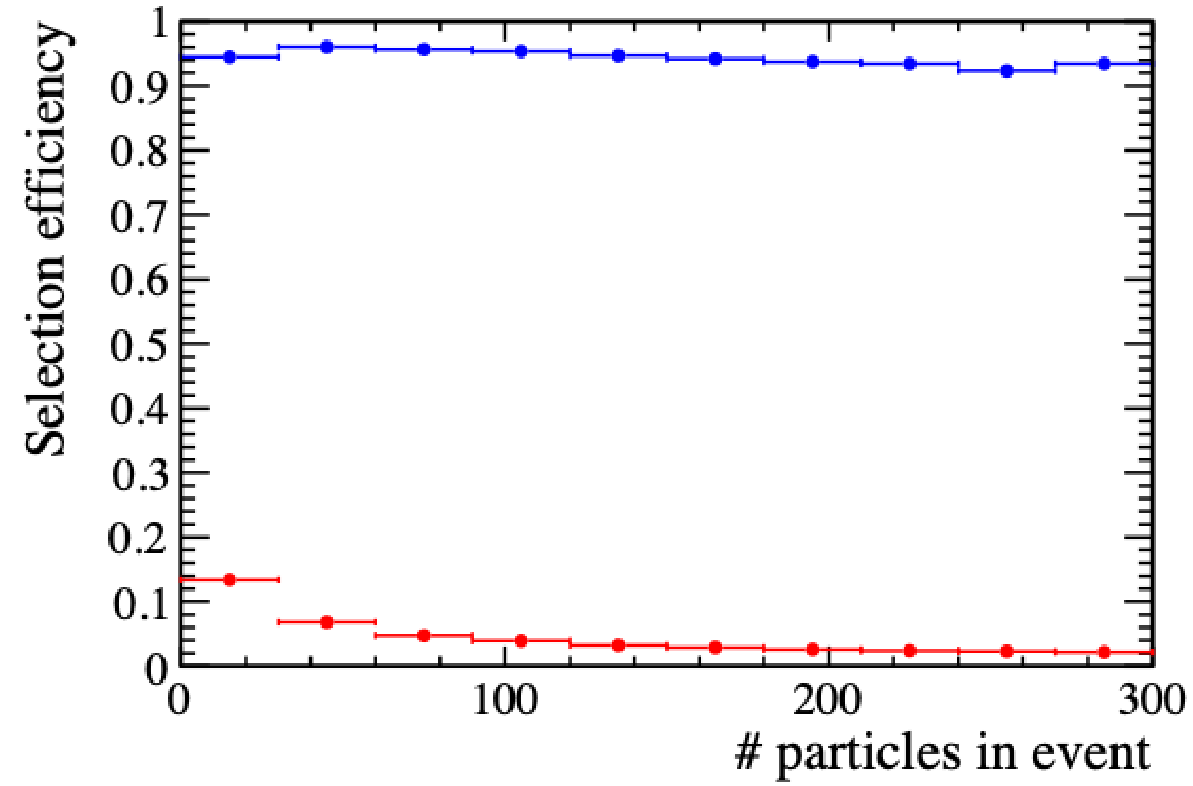
Run3-like conditions

The algorithm

Motivation

Performance: final-state particle filtering





Consistent performance with different number of signals

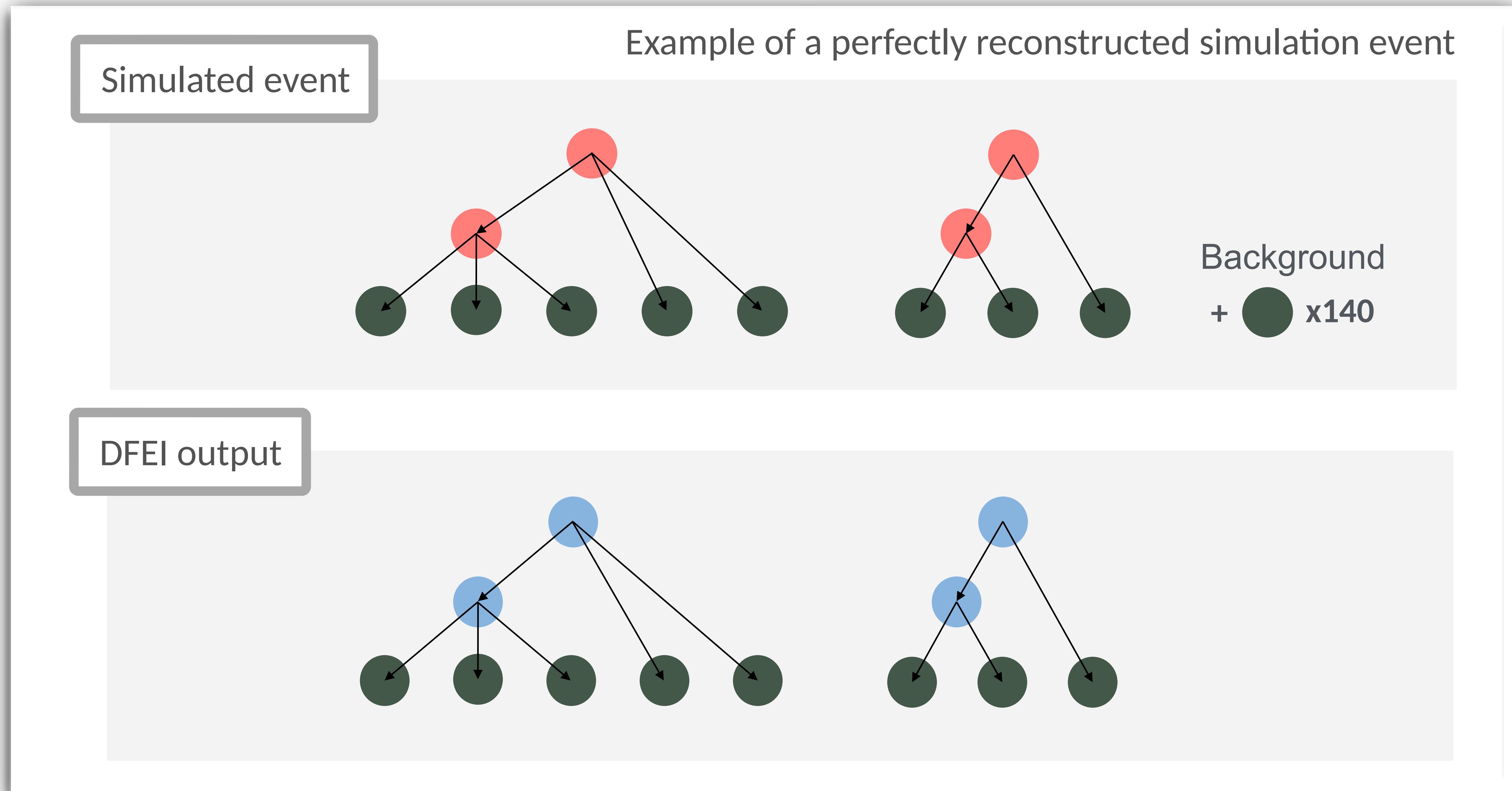
“single-b-hadron-signal” approach performance ***comparable*** to the envisaged nominal LHCb strategy for Run 3 [JINST 14 (2019) 04, P04006]

LHCb: 90% sig eff, 90% bkg rej. power
DFEI: 94% sig eff, 96% bkg rej. power

DFEI selects all of them simultaneously

DFEI capability #1
Powerful event size (~ x14) reduction in a multi-signal environment.

Performance: event reconstruction



- Percent level, comparable to Belle II performance
- Can be easily extended for more target variables!

DFEI capability #2
Automatised and inclusive reconstruction of decay chains.

Outlook

Performance

The algorithm

Motivation

Outlook and Summary

Increased particle multiplicities for **LHCb Upgrades I and II** bring **big challenges, both for trigger and offline analysis**

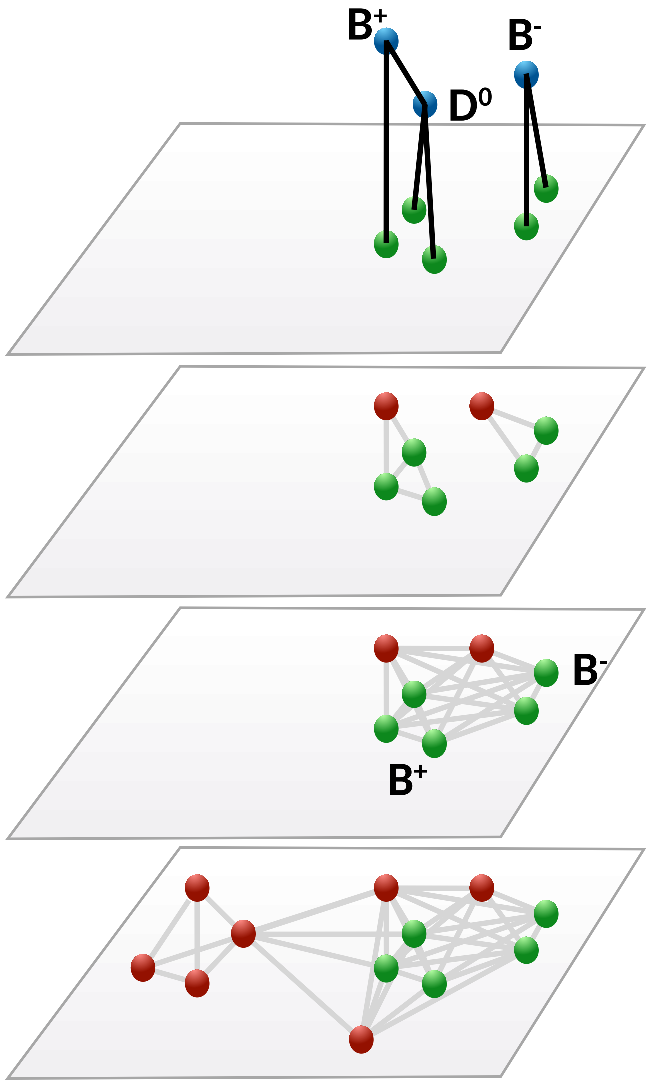
Paradigm change in trigger:
"which *events*?" → "which *parts* of the event?" [DFEI-arXiv:2304.08610]

Online application:

- Safely discard rest of event, minimal loss for analyses
- Hierarchical reconstruction of heavy-hadron decay chains

Offline application:

- Tool for powerful background classification & suppression



First prototype of the DFEI algorithm based on GNN
focused on b-hadron decays and charged stable particles
Very promising performance in realistic conditions!

Backup slides

Next steps

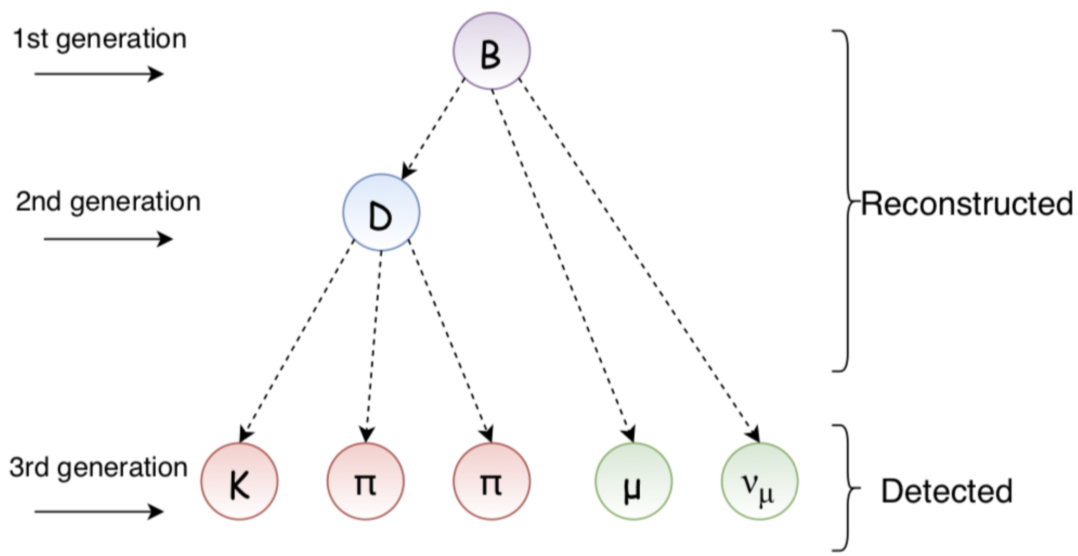
Algorithmic optimization and architectural choices
- Accuracy and useful information (separation, signal channels, …)
- Expansion in functionality (neutrals, PID, …)

Extensive performance studies, crucial for calibration
- In simulation
- In real data

Integration in LHCb trigger
- Export into ROOTs TMVA SOFIE, finishing GNN implementation
- Study usage of hardware accelerators for Upgrade II (FPGA, GPU, …)

3rd module: Lowest Common Ancestor (LCA) inference

From [BELLE2-MTHESIS-2020-006]:
(see also [James Kahn et al 2022 Mach. Learn.: Sci. Technol. 3 035012])



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μ	1	0	0	0	0	0	0
ν_μ	1	0	0	0	0	0	0

↔

LCA Matrix

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π	1	0	1	2	2
π	1	1	0	2	2
μ	2	2	2	0	2
ν_μ	2	2	2	2	0

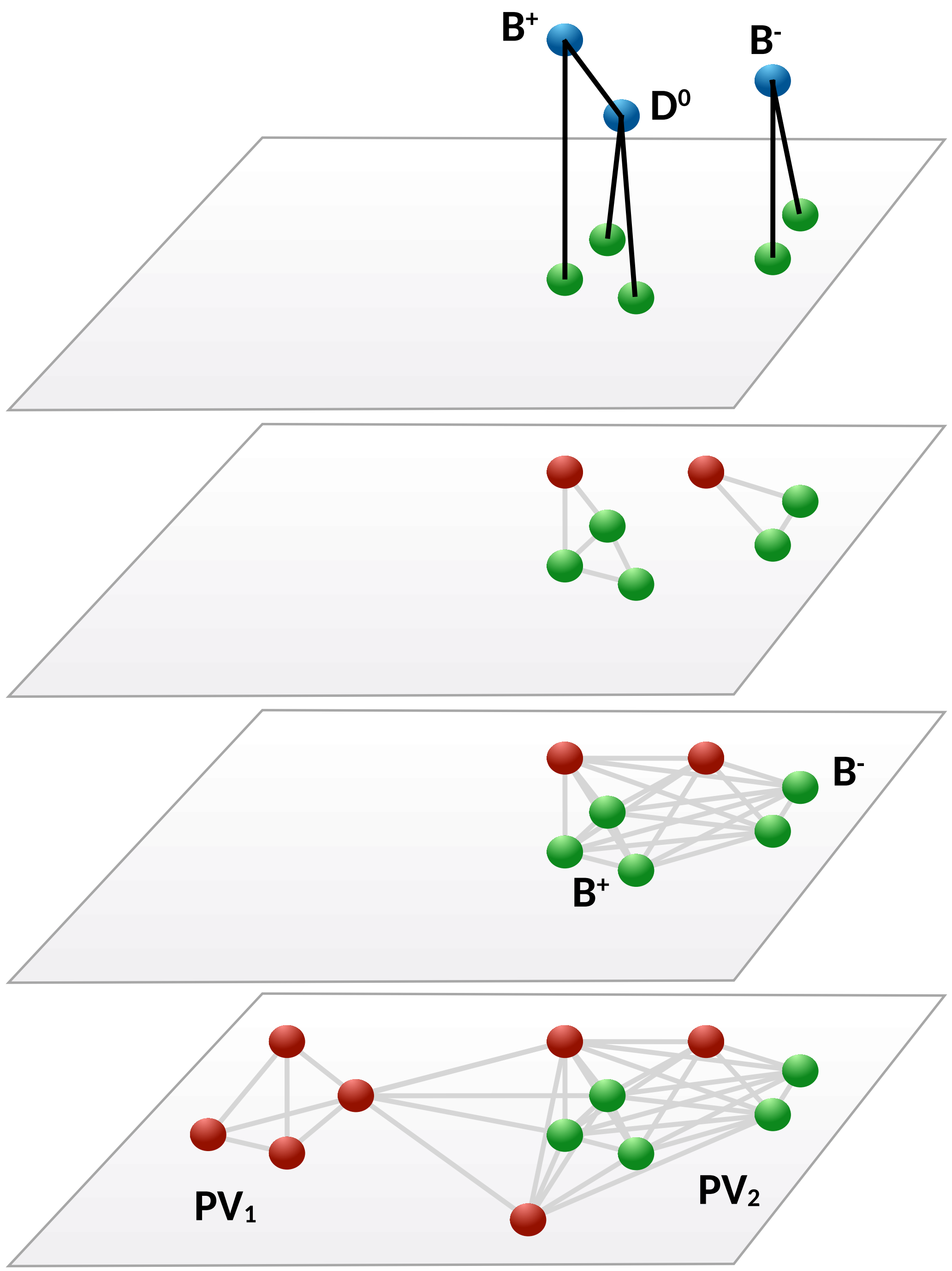
Problem reduced to **multi-class classification on edges**.

For the prototype, use as target a **simplified version of the decay chain, based on the reconstructible vertices**.

- Very-short-lived resonances merged with the previous ancestor.
- Resonances with less than two charged descendants merged with the previous ancestor.

Training

Dataset:
- PYTHIA-based simulation, Run 3-like conditions, approximated emulation of LHCb reconstruction.
- Events required to contain at least one b-hadron (inclusive decay).



PYTHIA configuration

- Proton-proton collisions at 13 TeV
- Avg. number of collisions per event: 7.6

Emulated "Reconstruction"
using public Run 3 expectations

- Within LHCb acceptance
- Origin point of the tracks
- Three-momentum of the tracks.
- Position of the primary vertices.

Dataset published
for benchmarking & comparison

10.5281/zenodo.7799169

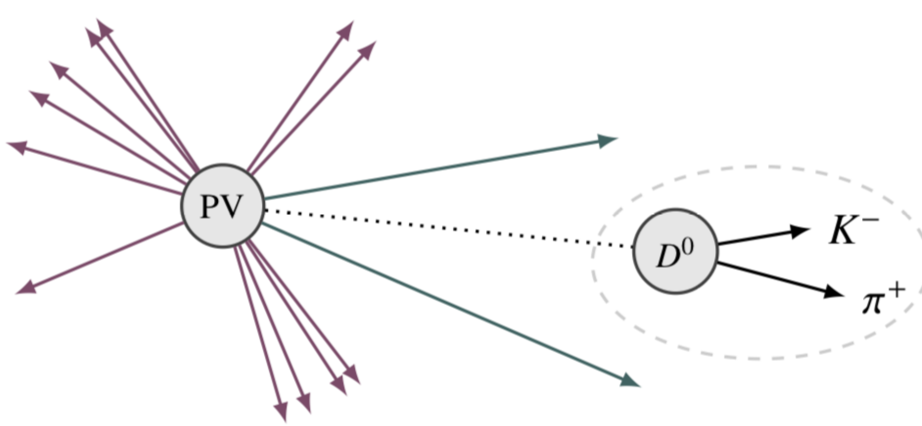
Signal-based trigger vs Full Event Interpretation (FEI)

Signal based

The current LHCb trigger is an **OR between many decay-mode selection lines**.

Since Run2, to reduce the event size, some lines **store only parts of the event which are related** to the specific signal. [JINST 14 (2019) 04, P04006]

E.g.: store the signal + the tracks in the same primary vertex (PV).



FEI

New proposal: try to **reconstruct the b- and c- hadron decay chains in the event**, in a hierarchical-clustering manner (cluster → unstable particle), **and discard the rest**.

Advantages:

- **Exploit extra correlations** between objects in the event.
- **Bandwidth oriented**: focus on storing as much “useful” information as possible.
- Case of several signals per event as an integral part of the approach.
- Establishment of a basis for an expanded functionality of the trigger: inclusive selections, study of anomalous events …

Training dataset: emulating Run3 conditions

Particle collision&decay

The training and performance studies are currently done using **PYTHIA**, with the following configuration:

- Proton-proton collisions at 13 TeV.
- Average number of collisions per event: 7.6.
- Selecting **events with at least one b-hadron produced (inclusive decay)**.

"Detection and reconstruction"

We require all the tracks and the b-hadrons to be **inside the LHCb geometrical acceptance**.

In addition, we **emulate the reconstruction of the following quantities**, using publicly available expectations for the LHCb performance in Run3 (see backup):

- **Origin point of the tracks** (first measurement in the Vertex Locator).
- **Three-momentum of the tracks**.
- **Position of the primary vertices**.

Decay-level performance

Decay mode	Perfect (%)	Wrong hierarchy (%)	Not iso. (%)	Part. reco. (%)
Inclusive H_b decay	4.6 ± 0.1	5.9 ± 0.1	76.0 ± 0.2	13.4 ± 0.1
$B^0 \to K_0^*[K\pi]\mu^+\mu^-$	35.8 ± 0.7	19.2 ± 0.6	44.9 ± 0.7	<0.02
$B^0 \to K^+\pi^-$	38.0 ± 0.7	$-$	54.7 ± 0.7	7.2 ± 0.4
$B_s^0 \to D_s^-[K^-K^+\pi^-]\,\pi^+$	32.8 ± 0.7	7.1 ± 0.4	53.7 ± 0.8	6.4 ± 0.4
$B^0 \to D^-[K^+\pi^-\pi^-]D^+[K^-\pi^+\pi^+]$	22.7 ± 0.6	22.4 ± 0.6	54.9 ± 0.8	<0.02
$B^+ \to K^+K^-\pi^+$	35.7 ± 0.7	10.2 ± 0.4	46.4 ± 0.7	7.7 ± 0.4
$\Lambda_b^0 \to \Lambda_c^+[pK^-\pi^+]\,\pi^-$	21.7 ± 1.0	8.9 ± 0.7	36.8 ± 1.2	32.6 ± 1.1
$B_s^0 \to J/\psi[\mu^+\mu^-]\,\phi[K^+K^-]$	26.9 ± 0.6	20.5 ± 0.5	52.5 ± 0.6	<0.02

Different types of decay reconstruction
- wrong hierarchy: correct tracks but wrong hirarchy
- Not isolated: additional tracks that do not belong to the decay
- missing tracks of the true decay

Fraction of perfect signal reconstruction is approximate the tag side efficiency for FEI at Belle (II) (order few per cent for semileptonic decays and few per mille for hadronic decays.)

Example of decay-tree simplification used in the prototype

Original chain of ancestors:

$\pi^+ \leftarrow$ ρ(770)^0 $\leftarrow$ φ(1020) $\leftarrow$ D$^+$ $\leftarrow$ B^0 $\leftarrow$ B^{*0}

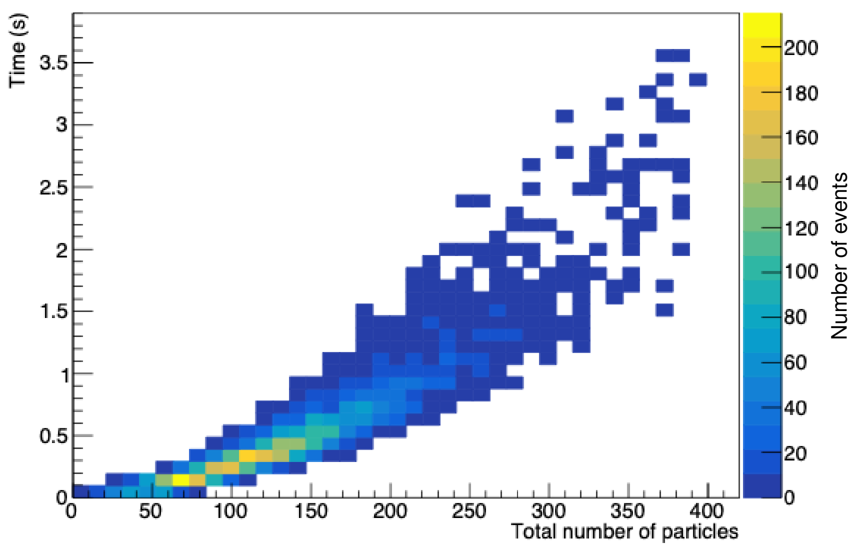
Simplified chain of ancestors (based on reconstructible vertices):

$\pi^+ \leftarrow$ D$^+$ $\leftarrow$ B^0

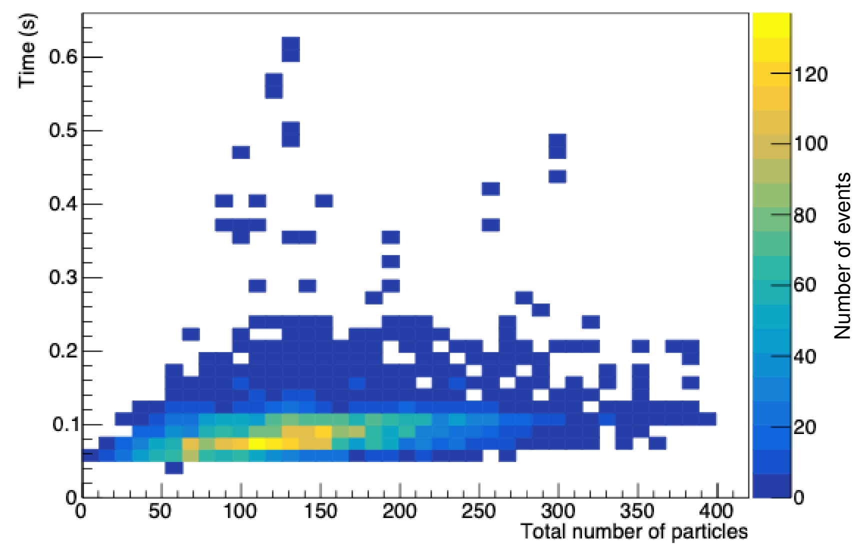
Performance: timing

Simplistic study (no parallelisation, no hardware accelerators*, algorithm to be further optimised), to **understand which are the slowest parts of the algorithm and how they scale with the total number of particles per event.**

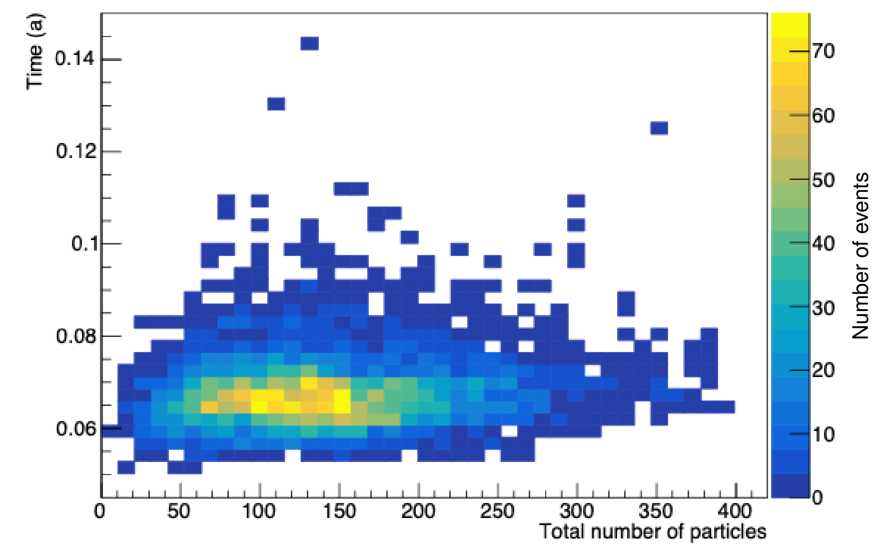
Node pruning



Edge pruning



LCA reconstruction



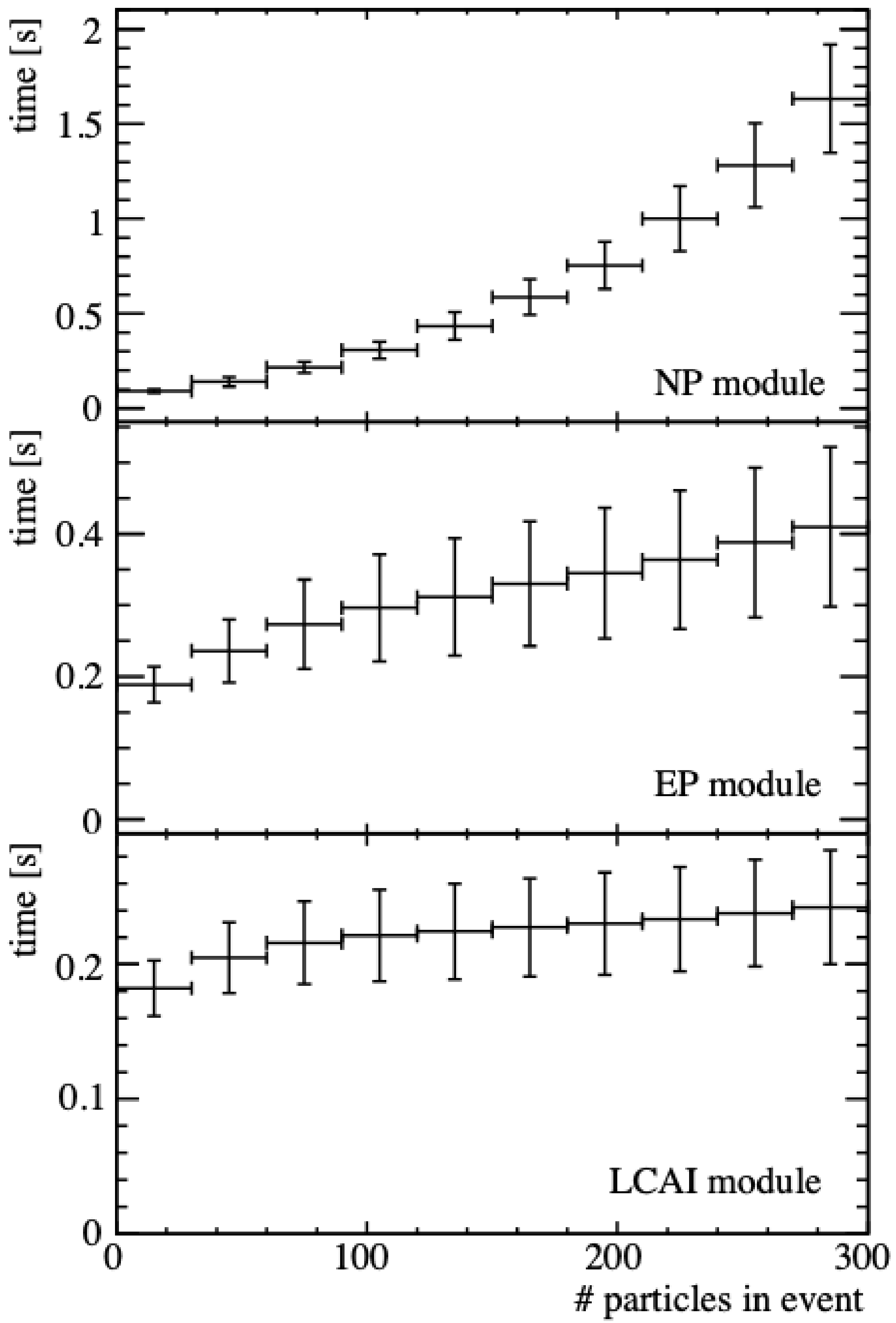
The slowest part is the node pruning, which also has the strongest dependency on the number of particles. → Many possible ways of optimisation.

The processing time of the subsequent algorithms is quite stable regarding changes in event complexity.

(*) Study done on a darwin-x86_64 architecture with a 2.8 GHz Intel Core i7 processor.

Timing studies

Scaling
current implementation



Implementation

Currently Python & TensorFlow
flexible for experimenting

TMVA SOFIE implementation (WIP)
Fast Inference System

Possible speed improvements

Simplification of layers, especially first
Approximate convolutions etc.

Hardware accelerators such as FPGA, GPU,...
WIP for GNNs in general