

## Bachelor's Colloquium

# **Electron track reconstruction studies and improvement for LHCb's real-time analysis trigger**

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## Motivation

tracking at LHCb:

- tracks of charged particles are reconstructed with efficiencies well above 90%
- **but:** electron tracking underperforms significantly compared to other particles
- due to bremsstrahlung
- interesting for e.g. LFU, FCNC<sup>1</sup>

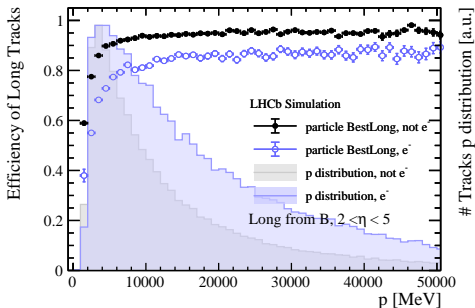


Figure: Long track efficiency of electrons and other particles.

<sup>1</sup>doi: 10.1088/1742-6596/1271/1/012009

## The LHCb Experiment

- major experiment at LHC, located at CERN
- tracking system: Vertex Locator (VELO), Upstream Tracker (UT), magnet, Scintillating Fibre (SciFi) tracker

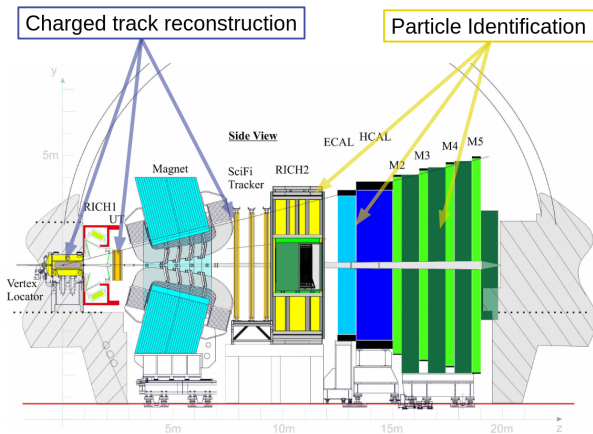


Figure: Side view of the LHCb detector in the non-bending  $y$ - $z$  plane [1].

## Electrons at LHCb

electrons lose energy while traversing the LHCb detector:

- energy loss dominated by bremsstrahlung
- typically lose 30% of energy upstream
- bremsstrahlung obstructs reconstruction

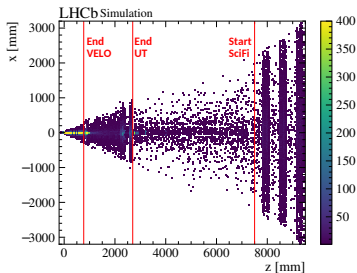


Figure: Bremsstrahlung vertices.

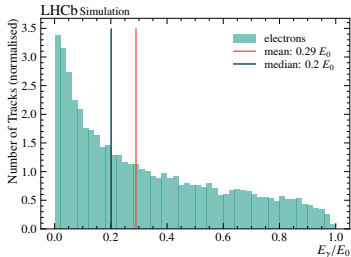


Figure: Energy loss of electrons in the LHCb detector upstream of the magnet.

## Tracking of Charged Particles

- **Long tracks** offer the highest relative momentum resolution with 0.5%
- main type used in physics analyses
- reconstructed via two methods: Forward tracking and **Track Matching**

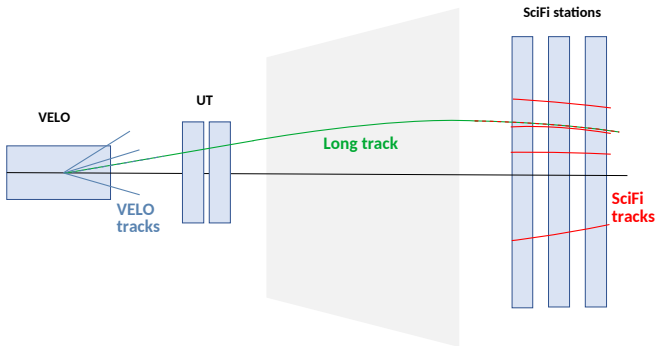


Figure: Sketch of LHCb track types in the  $x$ - $z$  plane.

# The Matching Algorithm

Basic idea:

- combine VELO and SciFi tracks of same particle to create Long tracks
- quantify the level of agreement, *i.e.* a match, between reconstructed VELO and SciFi tracks
- evaluate a neural network to obtain a combined estimate of being correctly matched

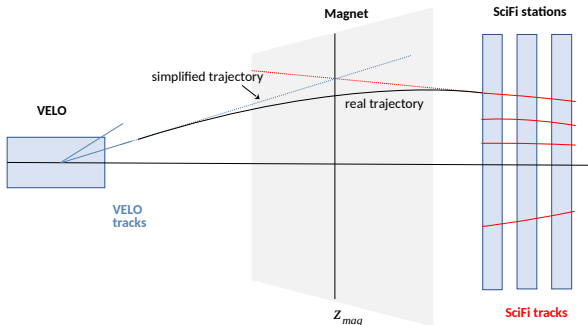


Figure: Basic idea of the Matching algorithm in the  $x$ - $z$  plane.

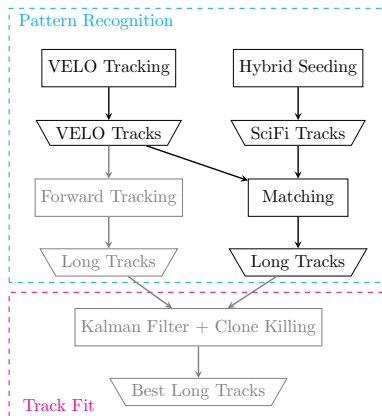
## Track Reconstruction in the Real-Time Analysis Trigger

LHCb's software trigger is split into two stages:

**HLT1:** partial event reconstruction

**HLT2:** full event reconstruction

→ **Matching**



**Figure:** Dataflow in the Long-track reconstruction in HLT2.

## Improving Electron Reconstruction

currently, all track-reconstruction tuning explicitly excludes electrons

solution:

- run dedicated electron reconstruction sequence in HLT2
- expand the Matching algorithm
  - must be computationally cheap to be run in the trigger
- tune parameterisations to electrons
- upper limit for the efficiency around 93%, determined by the VELO and SciFi track efficiency



## Track Preselection

electron track identification:

- tracks can be extrapolated to the ECAL and matched to EM-showers:
- VELO tracks to the EM-showers of photons emitted by electrons in the VELO
- SciFi tracks to the EM-showers of the electrons directly

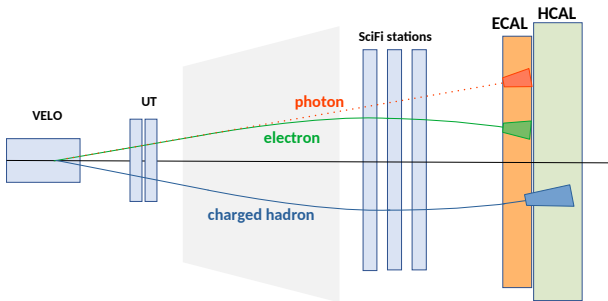


Figure: Different particle signatures in the LHCb detector.

## Parameter Tuning

tune all approximations and estimates used during the reconstruction to electrons, e.g. the  $z_{\text{mag}}$  parameter in the optical model

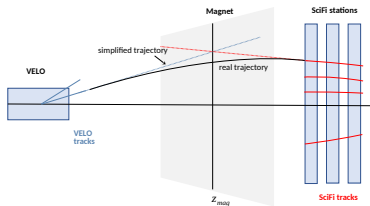


Figure: Magnet kick position  $z_{\text{mag}}$ .

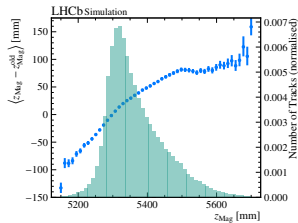


Figure: Old  $z_{\text{mag}}$  parameterisation for electrons.

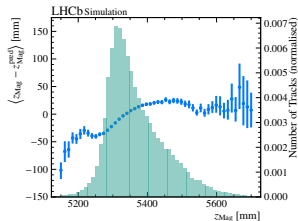


Figure: New  $z_{\text{mag}}$  parameterisation for electrons.

## Input Variables I

- six input variables
- most important:

$$D_x = |x_V(z_{mag}) - x_T(z_{mag})|$$

$$D_y = |y_V(z_{mag}) - y_T(z_{mag})|$$

- used to define a combined quality measure, mimicking a  $\chi^2$ :

$$\chi_{\text{match}}^2 = \frac{D_x^2}{\sigma_x^2} + \frac{D_y^2}{\sigma_y^2} + \frac{|\Delta t_y|^2}{\sigma_{t_y}^2}$$

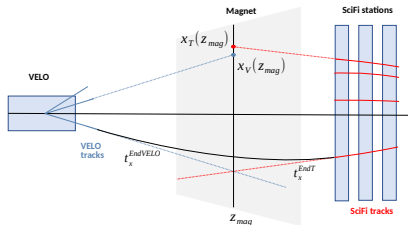


Figure: Bending plane.

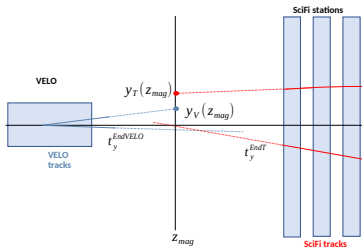


Figure: Non-bending plane.

## Input Variables II

- six input variables

- most important:

$$D_x = |x_V(z_{\text{mag}}) - x_T(z_{\text{mag}})|$$

$$D_y = |y_V(z_{\text{mag}}) - y_T(z_{\text{mag}})|$$

- used to define a combined quality measure, mimicking a  $\chi^2$ :

$$\chi_{\text{match}}^2 = \frac{D_x^2}{\sigma_x^2} + \frac{D_y^2}{\sigma_y^2} + \frac{|\Delta t_y|^2}{\sigma_{t_y}^2}$$

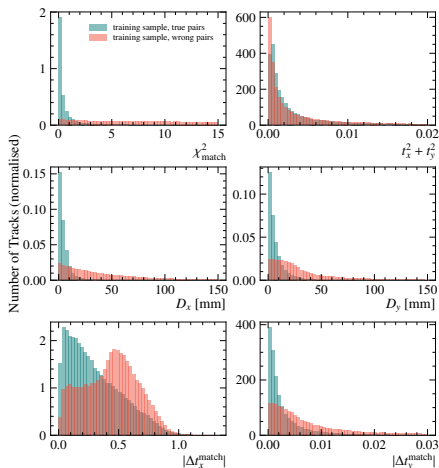


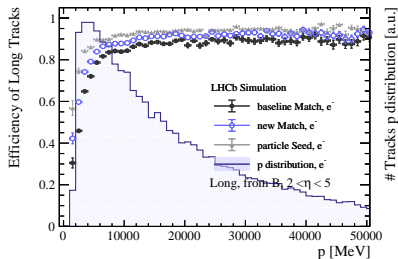
Figure: Matching MLP input variables.

## Test on Electron Tracks

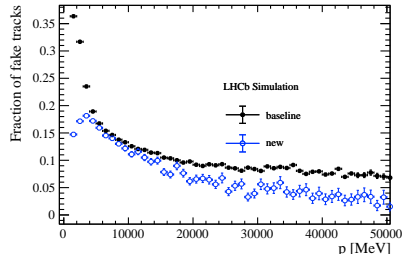
- new Matching algorithm tested on electron tracks
- efficiencies above 90%
- low ghost rate

**Table:** Integrated Matching efficiencies using perfect track selection.

Category	Old $\varepsilon$ [%]	New $\varepsilon$ [%]
Long	64.02	78.33
<b>Long from B</b>	83.46	87.90
$p > 5 \text{ GeV}$	86.22	90.05
$p > 3 \text{ GeV } p_T > 0.5 \text{ GeV}$	87.72	90.86
Ghost rate	17.44%	13.96%



**Figure:** Efficiency using perfect selection.



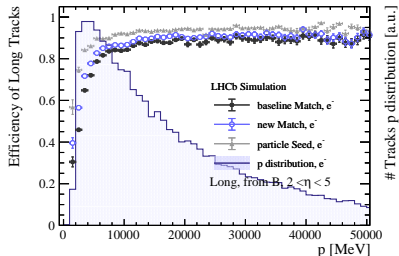
**Figure:** Ghost rate.

## Improved Electron Reconstruction Efficiency

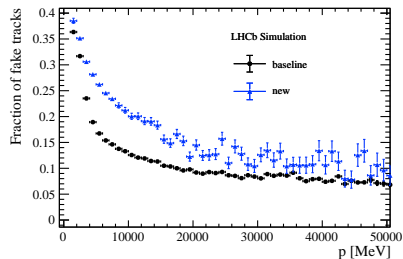
- new Matching algorithm + ECAL filter
- efficiency increases by 3%
- higher ghost rate
- reduce with ECAL filter

**Table:** Integrated Matching efficiencies using ECAL filtered tracks.

Category	Old $\varepsilon$ [%]	New $\varepsilon$ [%]
Long	64.02	76.41
<b>Long from B</b>	83.46	86.40
$p > 5 \text{ GeV}$	86.22	88.55
$p > 3 \text{ GeV } p_T > 0.5 \text{ GeV}$	87.72	89.39
Ghost rate	17.44%	26.37%



**Figure:** Matching efficiency using ECAL filter.



**Figure:** Ghost rate.

## Conclusion

summary:

- significant improvement in electron reconstruction
  - ECAL filter to control ghosts
  - improved Matching for electrons
- efficiencies around 90%

next steps:

- ECAL filter ready for SciFi tracks, but not for VELO tracks
- use the electron track-finding for real data
- optimise track fit
  - improve ghost rate

## References I

- [1] P. A. Günther, “Track reconstruction development and commissioning for LHCb’s run 3 real-time analysis trigger.”
- [2] A. Hoecker et al., “TMVA - toolkit for multivariate data analysis.”
- [3] V. Gligorov et al., “RTA and DPA dataflow diagrams for run 1, run 2, and the upgraded LHCb detector,”
- [4] A. Scarabotto, “Track matching for LHCb upgrade 2.”
- [5] C. A. Beteta et al., “Calibration and performance of the LHCb calorimeters in run 1 and 2 at the LHC.”





# BACKUP

# Multi-layer Perceptron (MLP)

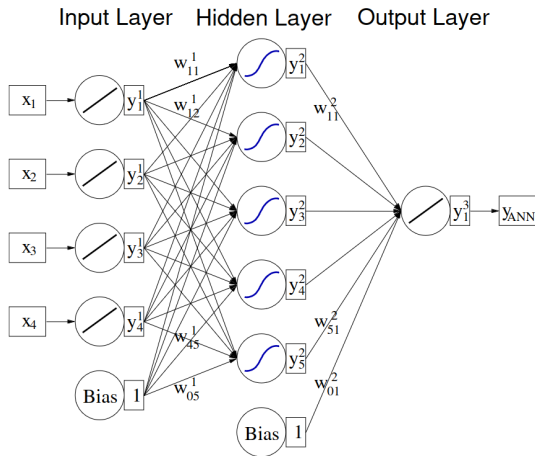
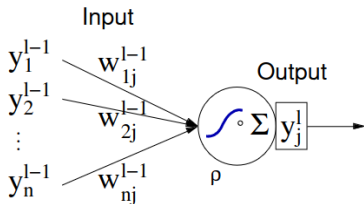


Figure: MLP with one hidden layer [2].

# Multi-layer Perceptron (MLP)



**Figure:** Single neuron  $j$  in layer  $l$  with  $n$  input connections [2].

Variable	Definition
$x_i$	inputs
$w_i$	synaptic weights
$b$	bias weight
$y_i$	predicted values
$d_i$	target values
$n$	# input connections
$\eta$	learning rate

neuron activation function

$$\rho = \text{ReLU} = \max \left( 0, b + \sum_{i=1}^n y_i w_i \right)$$

error function

$$E = \text{CE}(y_i, d_i)$$

$$= -\frac{1}{n} \sum_{i=1}^n (d_i \log(y_i) + (1 - d_i) \log(1 - y_i))$$

→ gradient descent of weight ( $\eta \equiv 0.02$ )

$$w'_i = w_i + \Delta w_i$$

$$= w_i - \eta \frac{\partial E(w_i)}{\partial w_i}$$

## Magnetic field at LHCb

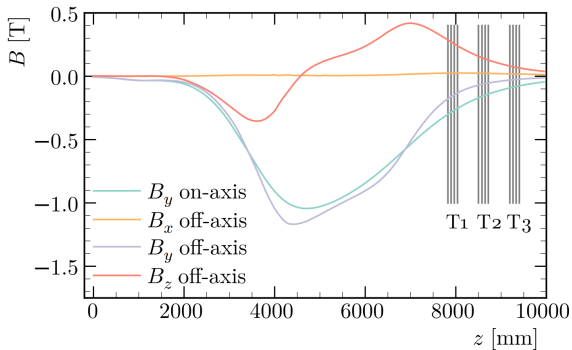
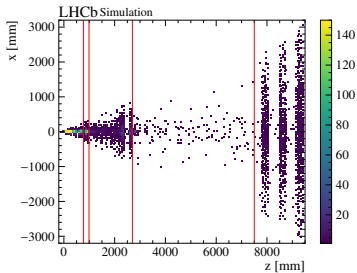
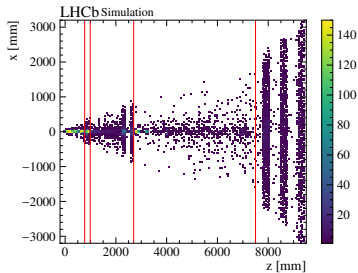


Figure: Magnetic field strength components as functions of  $z$  [1].

# Bremsstrahlung



a: Found.



b: Lost.

Figure: Distributions of bremsstrahlung vertices.

## Performance Metrics

two main performance indicators for track reconstruction with pattern recognition:

1. *track finding efficiency*  $\varepsilon = \frac{N_{\text{reconstructible \& MC-matched}}}{N_{\text{reconstructible}}}$
2. *fake track fraction or ghost rate*  $r_{\text{fake}} = \frac{N_{\text{fake}}}{N_{\text{total}}}$

where  $N$  is the number of tracks in the corresponding category

reconstructible:

- must pass through at least three VELO layers, leaving hits in each
- must leave a hit in at least one  $x$  and one stereo layer in each of the SciFi stations

MC-matched:

- more than 70% of the hits must originate from the same simulated particle

## RTA Trigger

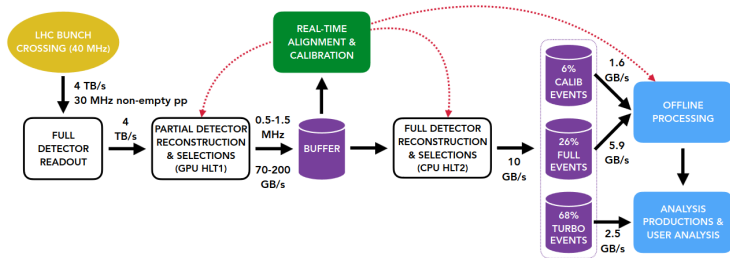
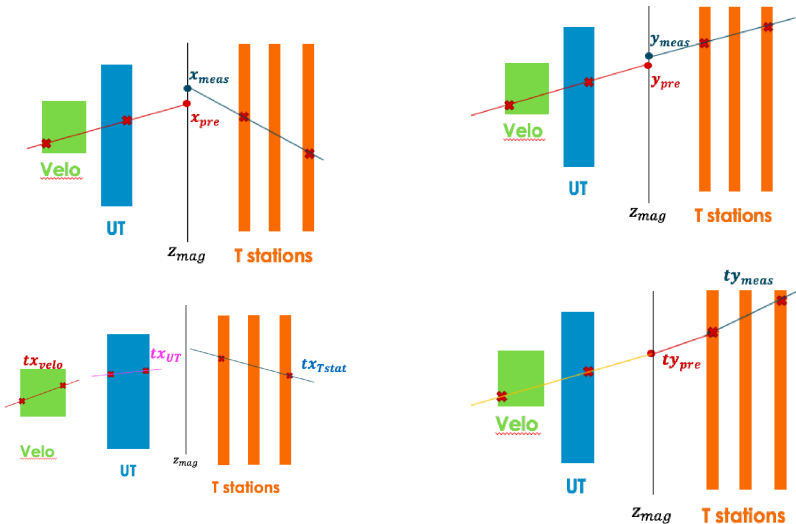


Figure: LHCb software trigger dataflow [3].

# Matching Variables

input variables of the Matching algorithm [4]:





## Matching Variables

$$t_x^2 + t_y^2 = \left(\frac{dx}{dz}\right)^2 + \left(\frac{dy}{dz}\right)^2$$

$$|\Delta t_x| = |t_x^{\text{EndT}} - t_x^{\text{EndVelo}}|$$

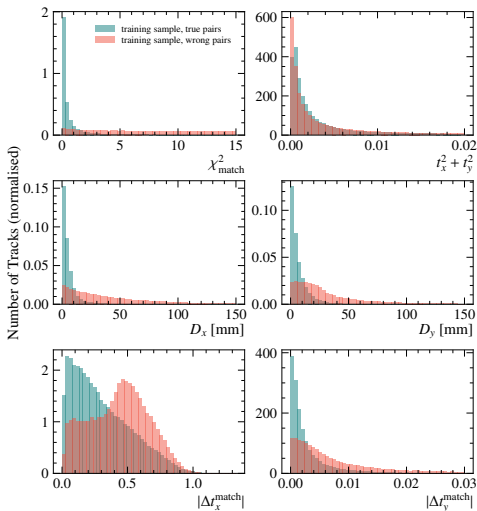
$$|\Delta t_y| = |t_y^{\text{EndT}} - t_y^{\text{EndVelo}}|$$

$$D_x^{\text{match}} = \left| x_T - x_V + t_x^{\text{EndT}}(z_{\text{mag}}^{\text{match}} - z_T) - t_x(z_{\text{mag}}^{\text{match}} - z_V) \right|$$

$$D_y^{\text{match}} = \left| y_T - y_V + t_y^{\text{EndT}}(z_{\text{match}} - z_T) - t_y(z_{\text{match}} - z_V) - y_{\text{corr}}^{\text{match}} \right|$$

$$\chi_{\text{match}}^2 = \frac{D_x^2}{\sigma_x^2} + \frac{D_y^2}{\sigma_y^2} + \frac{|\Delta t_y|^2}{\sigma_{t_y}^2}$$

# Input Variables



some changes in the preselection cuts yield better performance

Table: Matching MLP input variables for electron Matching.

Variable	Preselection
$\chi^2_{\text{match}}$	< 15
$D_x$	< 300 mm
$D_y$	< 300 mm
$ \Delta t_x^{\text{match}} $	< 2.0
$ \Delta t_y^{\text{match}} $	< 0.15
$t_x^2 + t_y^2$	

Figure: Matching MLP input variables.

# Electron Matching Tuning

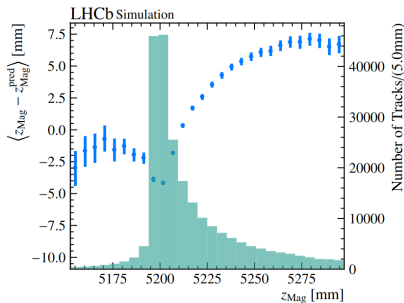


Figure: Old  $z_{mag}$  parameterisation for not-electrons [1].

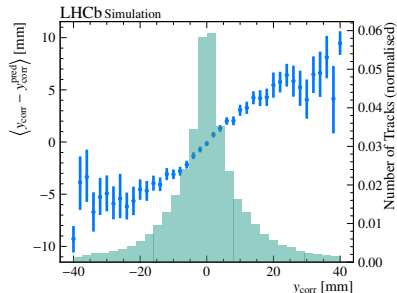
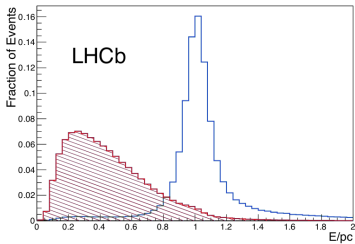


Figure: New  $y_{corr}$  parameterisation for electrons.

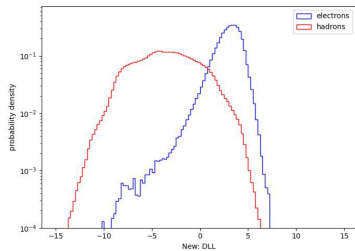
$$\begin{aligned}
 z_{mag}^{match} &= c_0 + c_3 t_x^2 + c_7 t_y^2 + |\Delta t_x| (c_1 + c_{11} t_y^2) + \Delta t_x^2 (c_5 + c_{10} |x_T|) \\
 &\quad + \Delta t_x t_x (c_4 + c_9 |x_T|) + |x_T| (c_2 + c_8 t_x^2 + c_6 |x_T| + c_{12} |x_T|^2) \\
 y_{corr}^{match} &= t_y (c_2 |\Delta t_x|^2 + c_3 |\Delta t_y|^2) + \Delta t_y (c_0 + c_1 \Delta t_x)
 \end{aligned}$$

# ECAL Filter

SciFi track selection using the already implemented ECAL filter



a:  $E/p$  using Run 1 data.



b: DLL using simulated events.

**Figure:** Distributions of variables in the ECAL for electrons (blue) and hadrons (red). Figure (a) adopted from Ref. [5], and (b) from (M. V. Veghel, personal communication, January 2024).

## Electron Matching Response

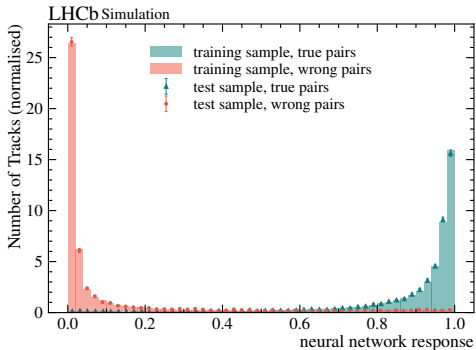


Figure: Output of the electron Matching MLP.

# BestLong Efficiency

Table: Integrated Best Long efficiency of electrons and other particles.

BestLong	$e^\pm$ $\varepsilon$ [%]	not- $e^\pm$ $\varepsilon$ [%]
From $B$	82.04	91.08
$p > 5 \text{ GeV}$ from $B$	84.15	94.00
Ghost rate	10.26%	

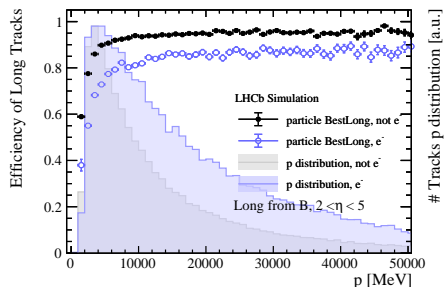
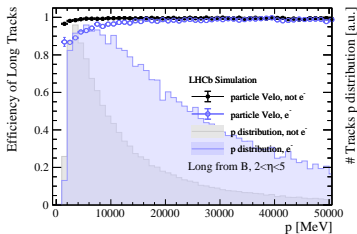
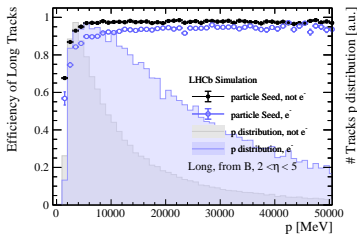


Figure: Long track efficiency of electrons and other particles, after Kalman filter.

# VELO and T Track Efficiencies



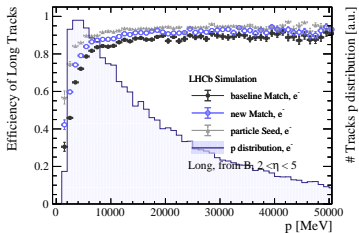
a: VELO Tracking efficiency.



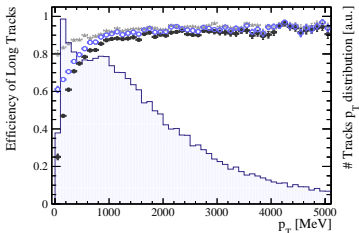
b: Hybrid Seeding efficiency.

Figure: Reconstruction efficiency of (a) the VELO Tracking, and (b) the Hybrid Seeding algorithm.

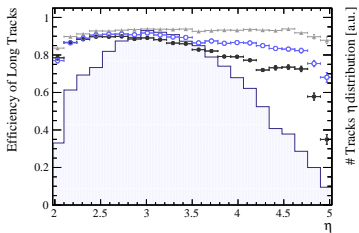
# New Matching Efficiencies - Long from $B$



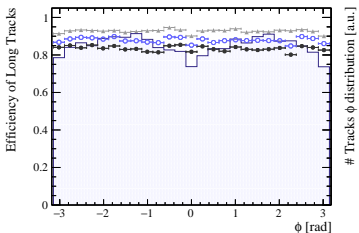
a: Efficiency vs. momentum.



b: Efficiency vs. transverse momentum.



c: Efficiency vs. pseudorapidity.

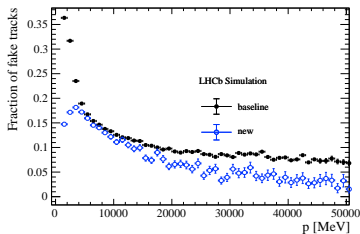


d: Efficiency vs. azimuthal angle.

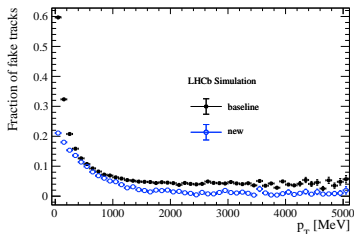
Figure: Track finding efficiency of the electron Matching algorithm.



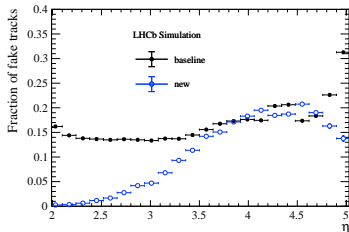
## New Matching Fake Track Fraction - Long from $B$



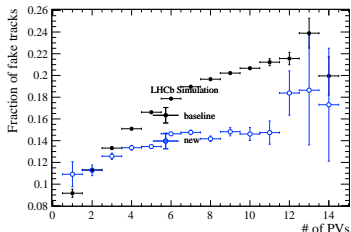
a: Fake track fraction vs. momentum.



b: Fake track fraction vs. transverse momentum.

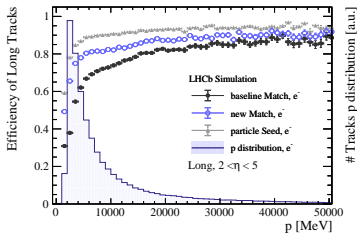


c: Fake track fraction vs. pseudorapidity.

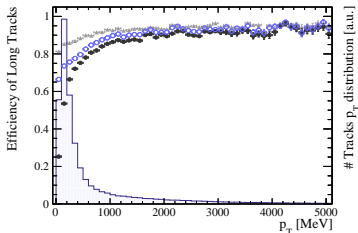


d: Fake track fraction vs. number of primary vertices.

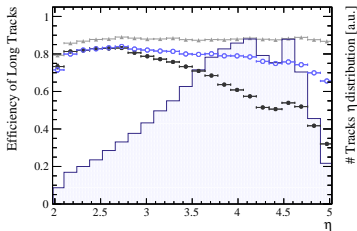
# New Matching Efficiencies - Long



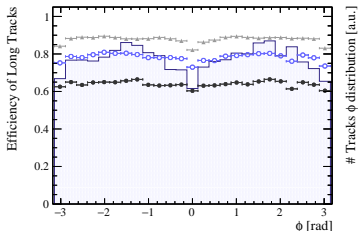
a: Efficiency vs. momentum.



b: Efficiency vs. transverse momentum.



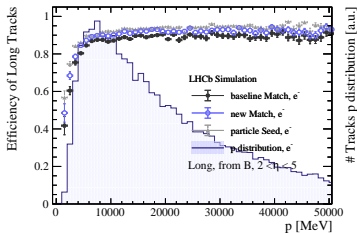
c: Efficiency vs. pseudorapidity.



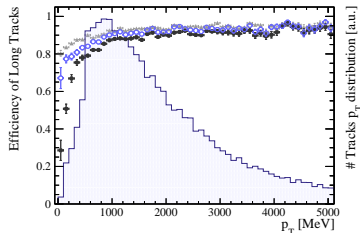
d: Efficiency vs. azimuthal angle.

Figure: Track finding efficiency of the electron Matching algorithm.

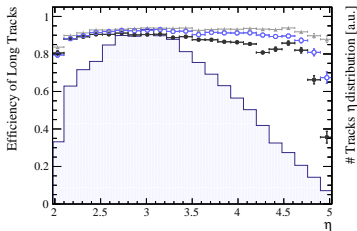
# New Matching Efficiencies - Long from $B p > 3 \text{ GeV } p_T > 0.5 \text{ GeV}$



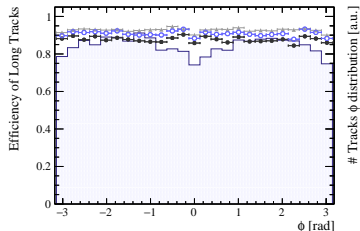
a: Efficiency vs. momentum.



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d: Efficiency vs. azimuthal angle.

Figure: Track finding efficiency of the electron Matching algorithm.