





Development of the Topological Trigger for LHCb Run 3

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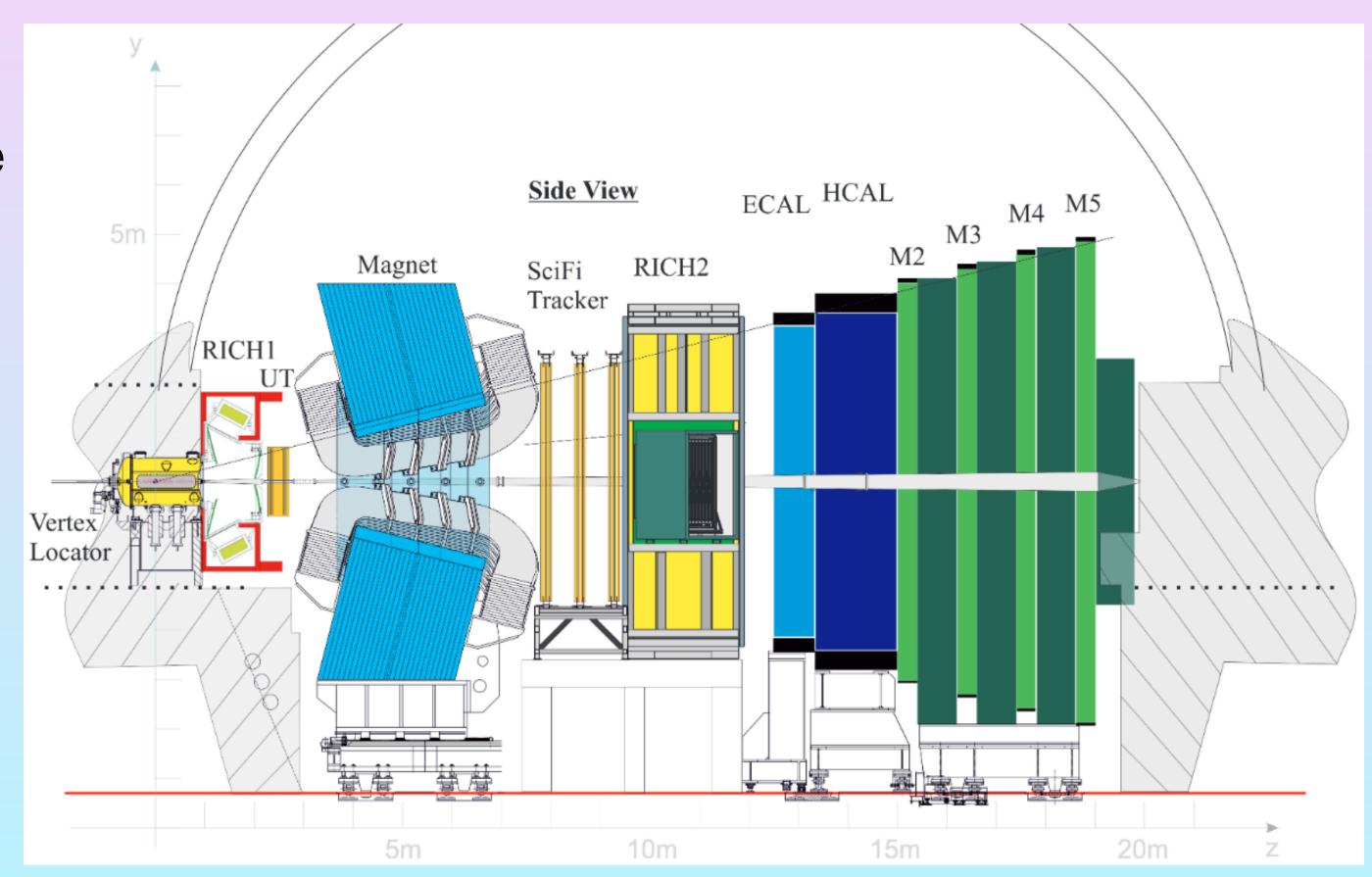
ACAT 2022 - Track 1 Computing Technology for Physics Research



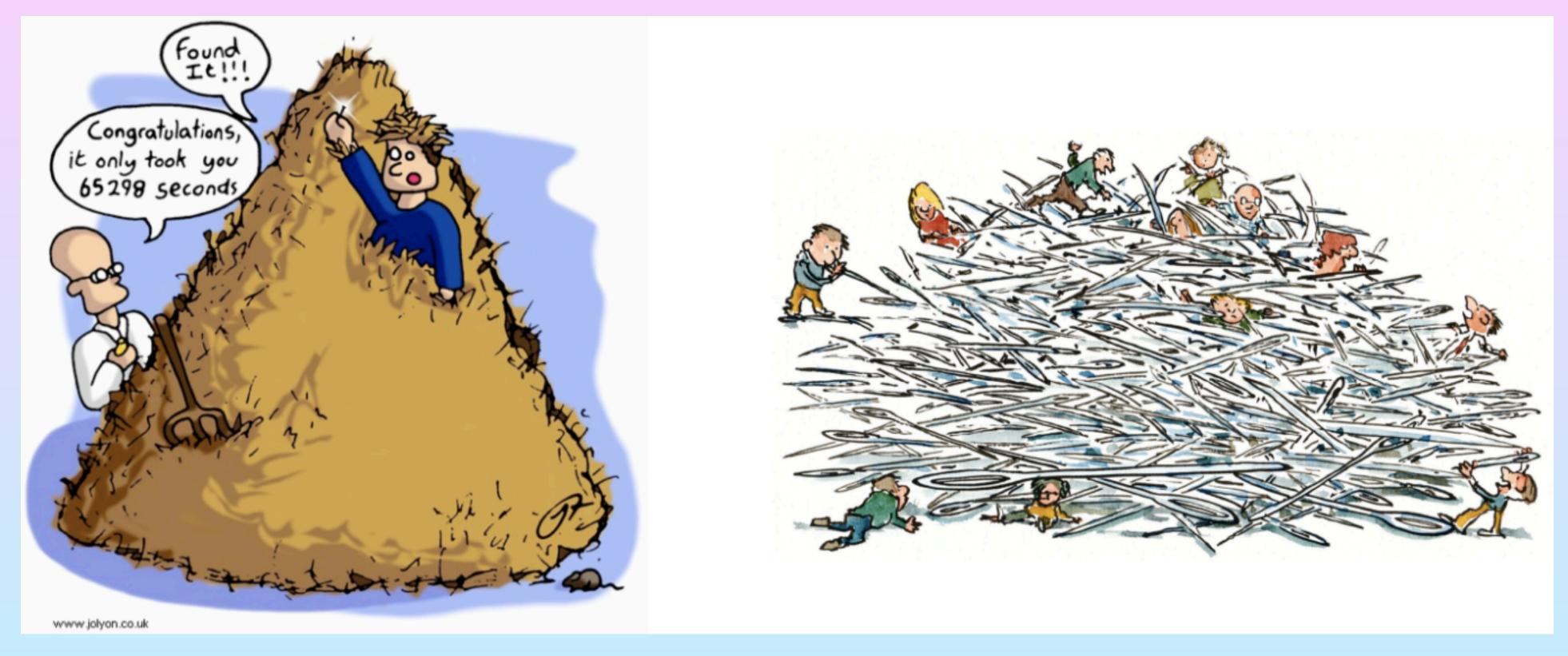


The LHCb Detector in Run 3

- Forward arm spectrometer aiming to measure properties of beauty and charm properties
- Software-only trigger system
- Run 3 Luminosity: $\mathcal{L} = 2 * 10^{33} \text{ cm}^{-2} \text{ s}^{-1}$
 - → 5x higher than in Run 2
- Brand new Tracking system and electronics



Why do we need a Trigger System?

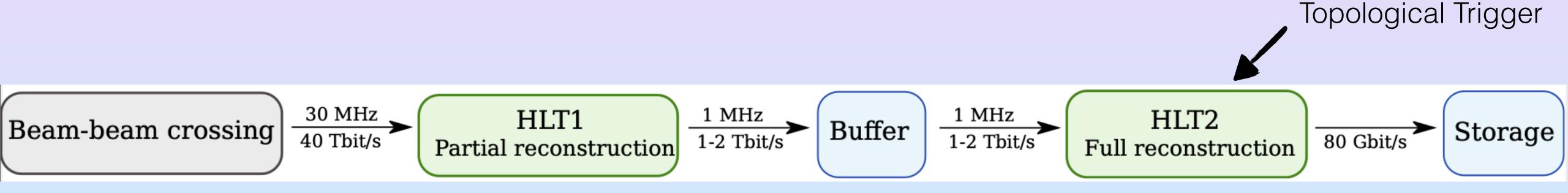


Data acquisition: 30 MHz -> ~495 Ebit or 60 EB for one year* →Real-Time-Analysis (Fast selection of events before storage)

Snowden, P. Thompson, T. Troscianko, Oxford University Press (2006)

LHCb's Trigger System in Run 3

- Remove Hardware Trigger LO
- Conversion to a software-based Trigger System

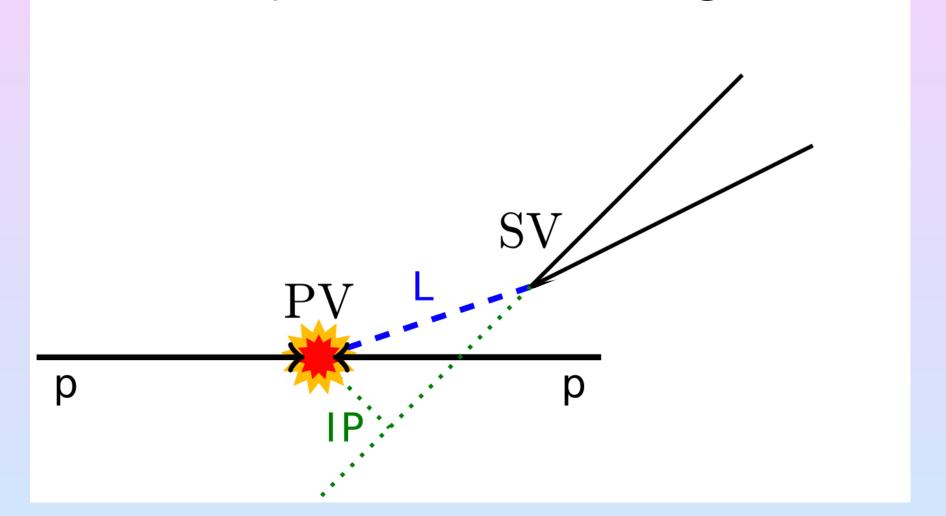


Reduce the incoming data by a factor of 500!

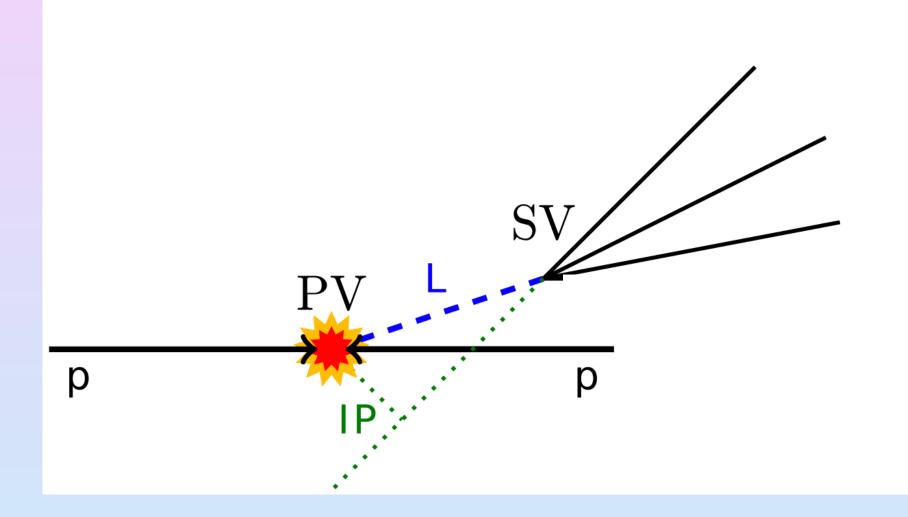
Assuming again 143 days of pp-collisions ~1Ebit or 125PB for storage

HLT2 Reconstruction and Selection

TwoBody Selection Algorithm



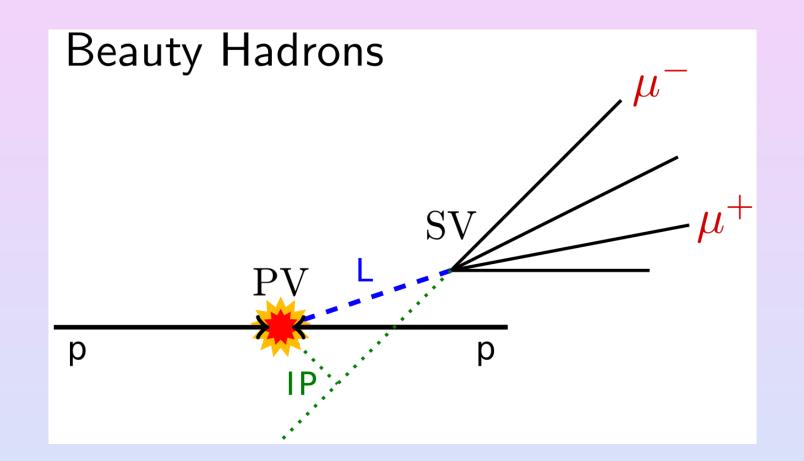
ThreeBody Selection Algorithm

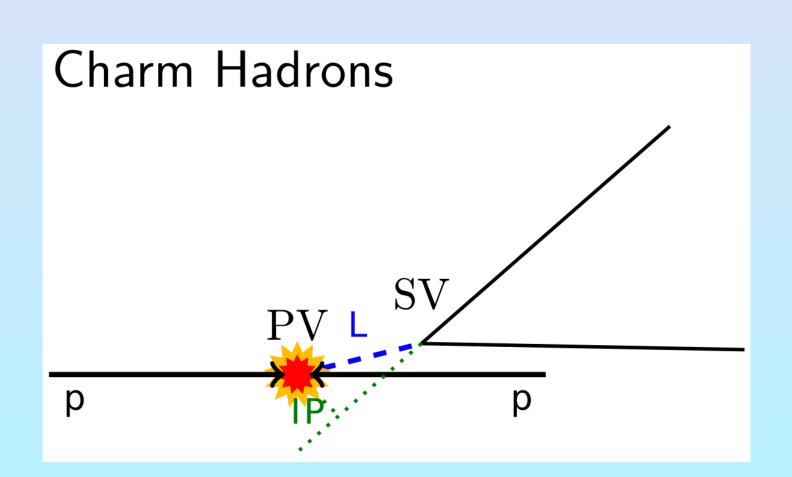


- 1. HLT2 particle reconstruction and identification
- 2. Filter particles using kinematic and spatial properties
- 3. Combine Particles to decay candidates
- 4. Filter decay candidates
- 5. ThreeBody: Repeat Steps 3 and 4
- 6. Trigger Decision

What is the Topological Trigger?

- Inclusive selection targeting any beauty decay
- Discriminate beauty events from charm contributions and soft QCD
- Develop inclusive signature by training on various exclusive beauty samples
 - Each sample provides equal amount of signal to avoid biases
- Based on the topology of beauty decays:
 - Relatively high mass ~5GeV
 - Displaced decay vertex due to lifetime $\sim \mathcal{O}(1 \text{ ps})$
 - Boosted in the forward direction ~travels $\mathcal{O}(1 \text{ cm})$ before decay

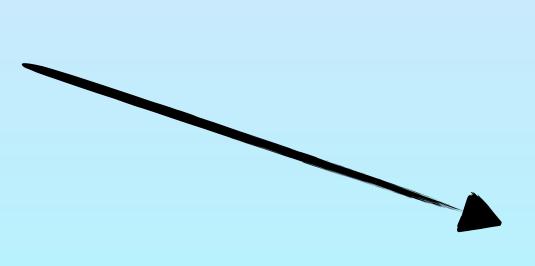




Why is the Topological Trigger important?

- 1. One of the main beauty Triggers for LHCb
- 2. Selection writes directly to storage
- 3. Pre-Filter for other selections

Electrons	Muons
HLT1 in Run I	
TrackAllLO	TrackAllLO
HLT1 in Run II	
TrackMVA	TrackMVA
HLT2 in Run I	
Topo(2,3,4)BodyBBDT	Topo(2,3,4)BodyBBDT
TopoE(2,3,4)BodyBBDT	TopoMu(2,3,4)BodyBBDT
HLT2 in Run II	
Topo(2,3,4)Body	Topo(2,3,4)Body
TopoE(2,3,4)Body	TopoMu(2,3,4)Body



The Topological Trigger: Then vs. Now

Major upgrades in Run 3 for detector and software require innovative improvements for trigger algorithms

Machine Learning Algorithms have improved majorly over the past few years

Run 2: Boosted Decision Tree Algorithm

 Followed same selection strategy as in Run 1 with updated preselection and training of the BDT

Run 3: Neural Network (Monotonic Net)

 Reinvent the selection strategy by moving to a different machine learning algorithm that provides robustness against detector effects and sensitivity to long-lived beauty objects

Monotonic Net - Robustness

Neural Network architecture provides two main features:

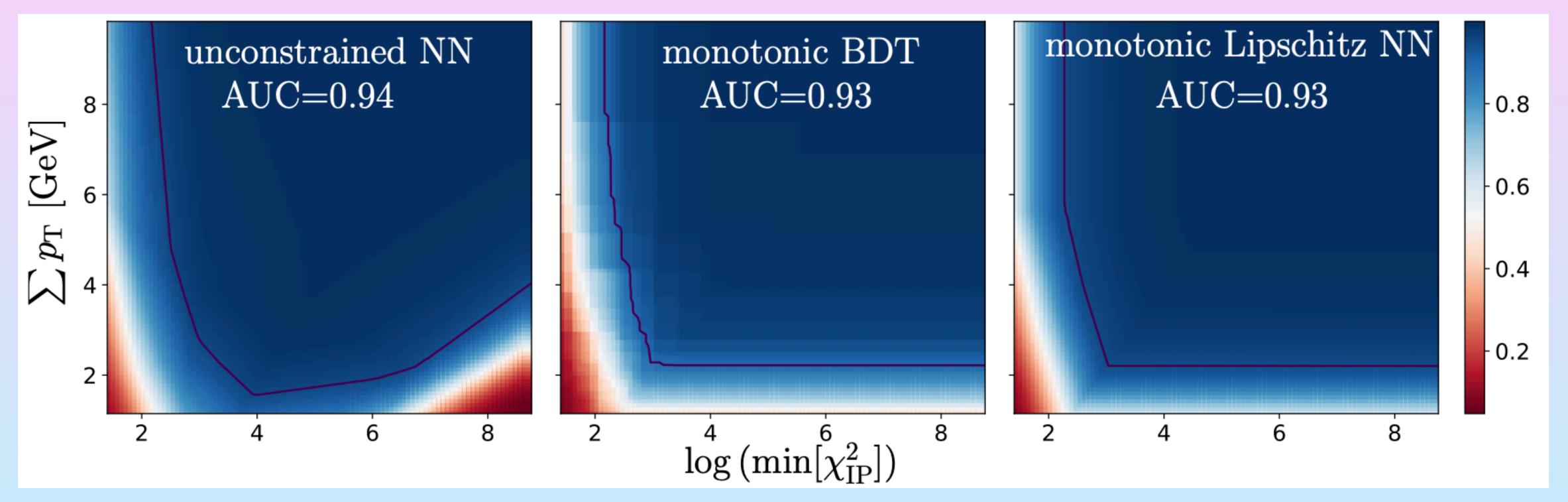
- Robustness: Introduce Lipschitz constant λ:
 - Having inputs x and y and corresponding Prediction f(x) and f(y):
 If(x) f(y)I ≤ λ Ix yI
 - Means: Change of the output is constrained by the change in the input -> Detector effects are reduced -> Robustness ensured

Monotonic Net - Monotonicity

- Monotonicity: Given feature i and inputs x and y:
 - If $x_i < y_i$ ensure that the Prediction follows $f(x_i) < f(y_i)$
 - Chose features where monotonic behaviour could be beneficial
 - Chose direction of the monotonicity: positive or negative

For more details see <u>Ouail Kitouni's talk</u> from Monday

Monotonic Net - An Example

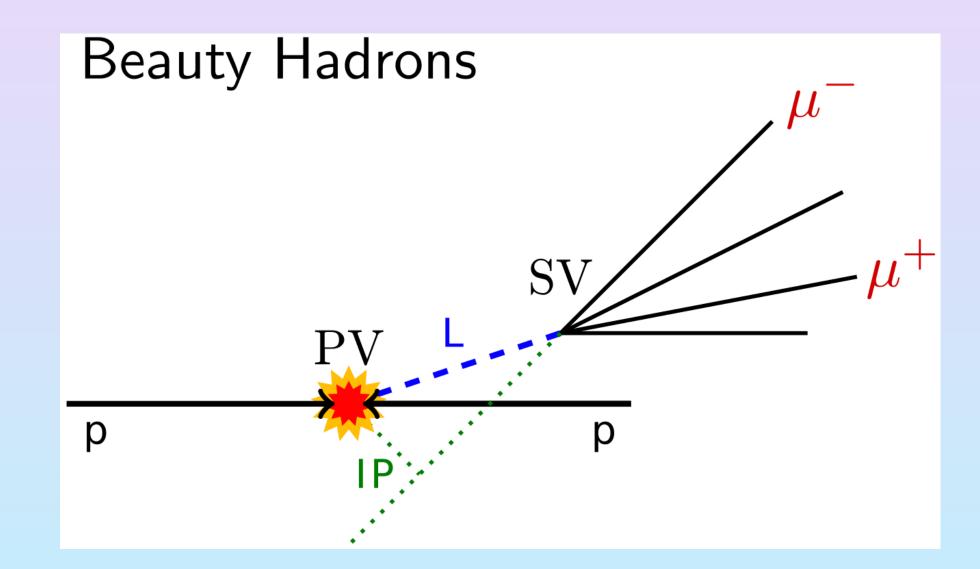


Blue regions: Classified as signal; Red regions: Classified as background; Black Line: Decision Boundary

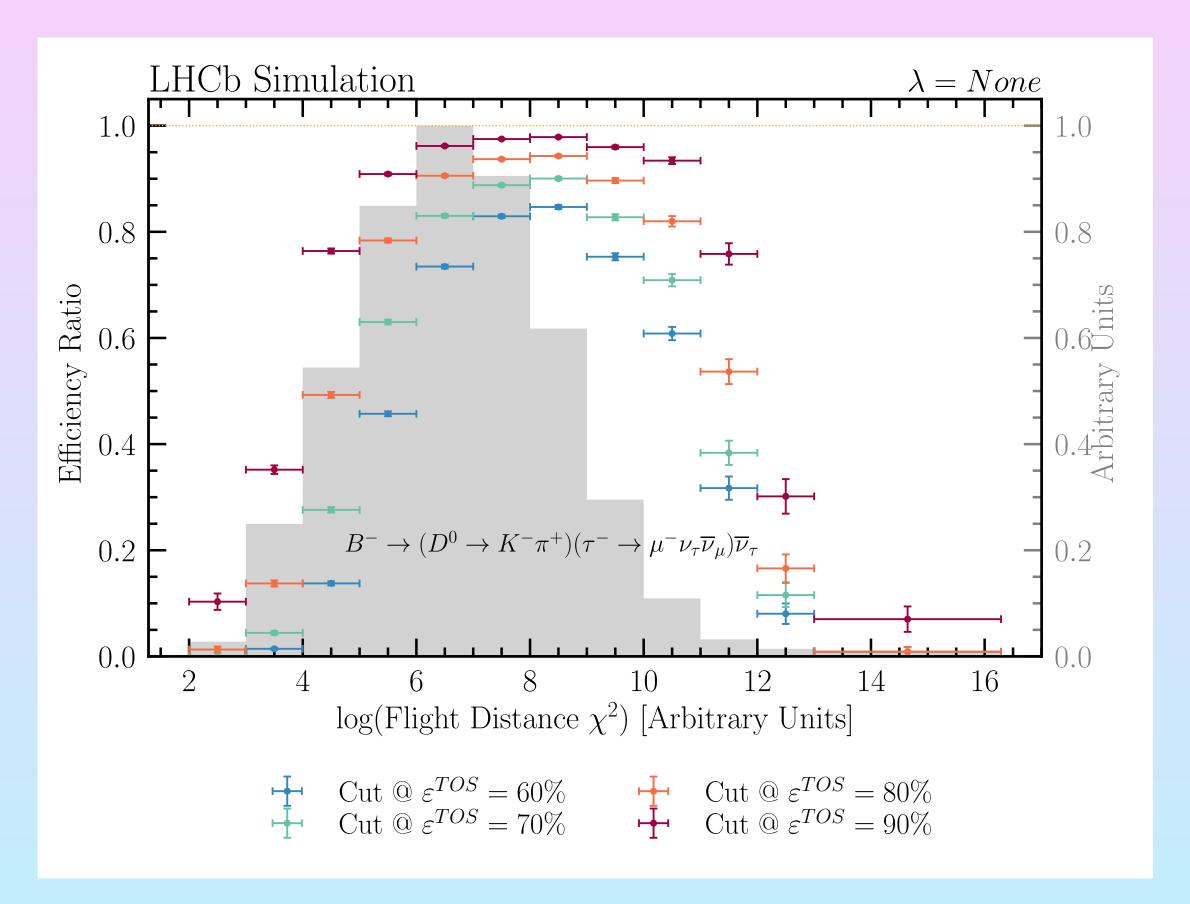
- →Monotonic Net is more sensitive to long-lived signal candidates than unconstrained Neural Networks
- →Monotonic Net provides a smoother decision boundary than other monotonic architectures (Robustness)

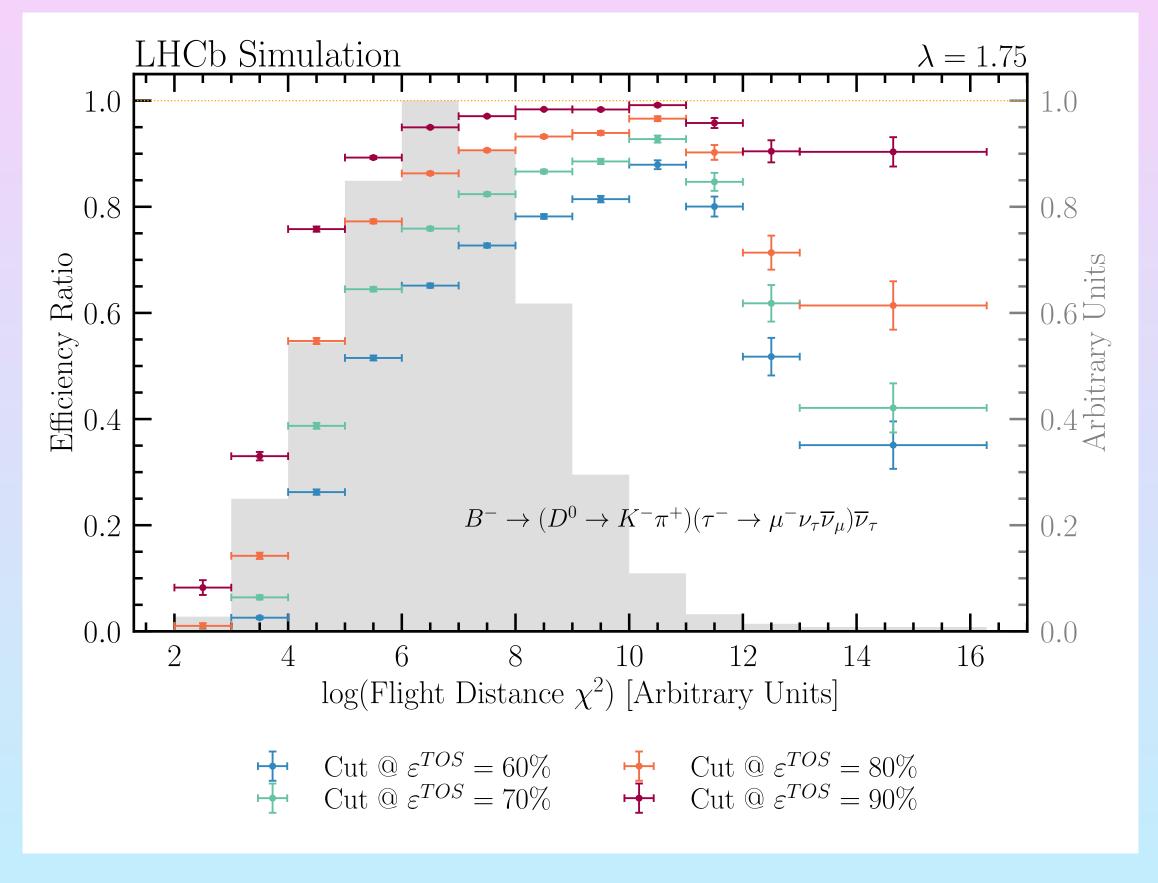
(Monotonic) Features

- Kinematic properties: (transverse momentum p_T)
- Spatial properties: (Flight Distance FD, Impact Parameter IP, Distance of closest approach DOCA)
- Conceptually the **positive monotonic** features are:
 - Transverse Momentum
 - Flight Distance
 - χ^2 of the Impact Parameter (IP)
- → Highly displaced tracks with a high momentum
- → Long-Lived beauty objects



The Monotonic Net in the Topological Trigger



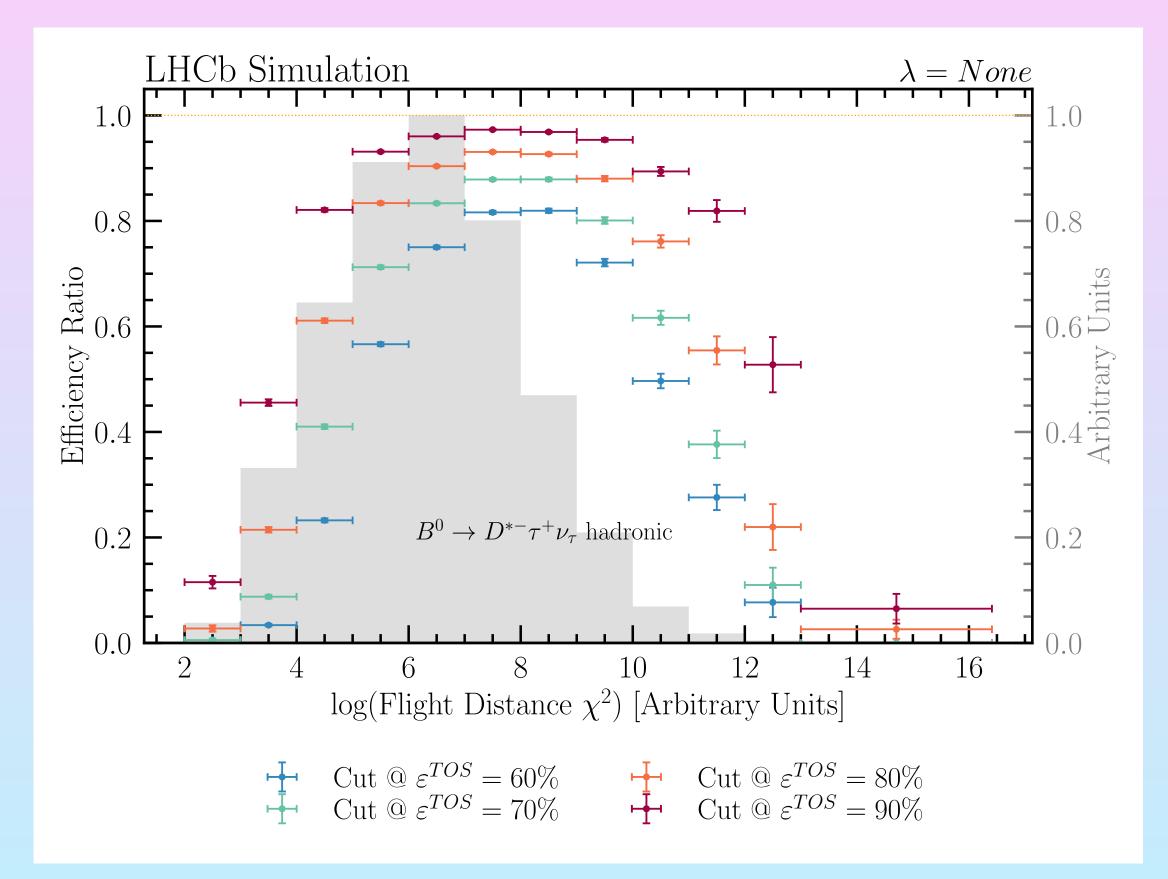


Unconstrained Neural Network (~BDT)

Constrained Neural Network (Monotonic Net)

Monotonic Lipschitz Neural Network provides higher signal efficiency for candidates with a higher flight distance (more displaced tracks)

The Monotonic Net in the Topological Trigger



LHCb Simulation 0.8 0.8 Efficiency Ratio
9.0
9.0 $B^0 \to D^{*-} \tau^+ \nu_{\tau}$ hadronic 0.2 0.2 16 $\log(\text{Flight Distance }\chi^2)$ [Arbitrary Units]

Unconstrained Neural Network (~BDT)

Constrained Neural Network (Monotonic Net)

Run 3 model of the Topological Trigger sensitive to long-lived beauty candidates and therefore also to possible BSM physics

Implementation in the LHCb Software

Bandwidth [MB/s] = Event Size [kB] x Output Rate [kHz]

Limited!

Output Rate

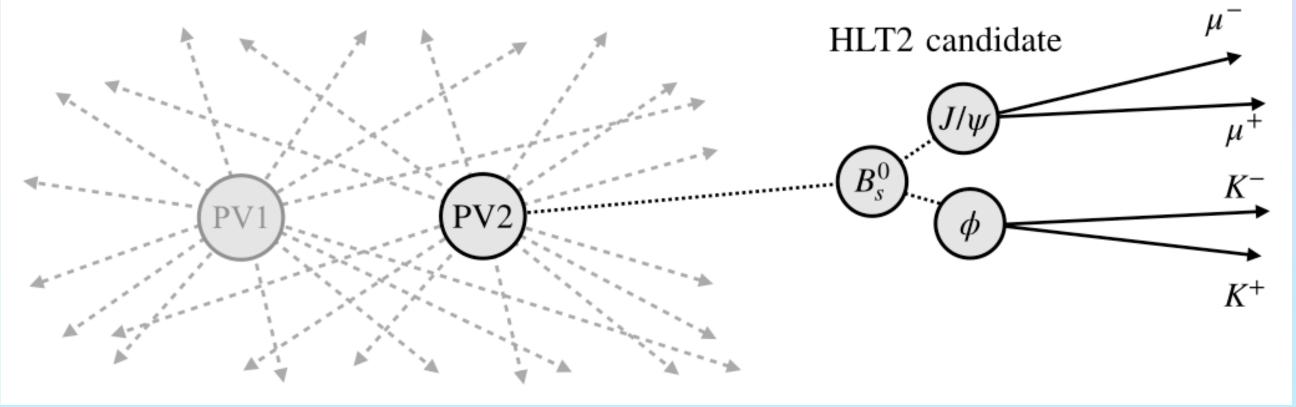
- Control the rate by cutting on the response of the neural network output
- Connected to the efficiency of the selection
 - → Tighter cut on response lower the efficiency
 - → Maximise physics output using minimal computational resources

Implementation in the LHCb Software

Bandwidth [MB/s] = Event Size [kB] x Output Rate [kHz]

Event Size

- Current Implementation:
 Found Signal Candidate in Event?
 - → Save entire event
 - → High event size
- Future **Possibilities**:
 - → Save the parts of the event that correspond to the signal candidate only
 - → Lowered event size



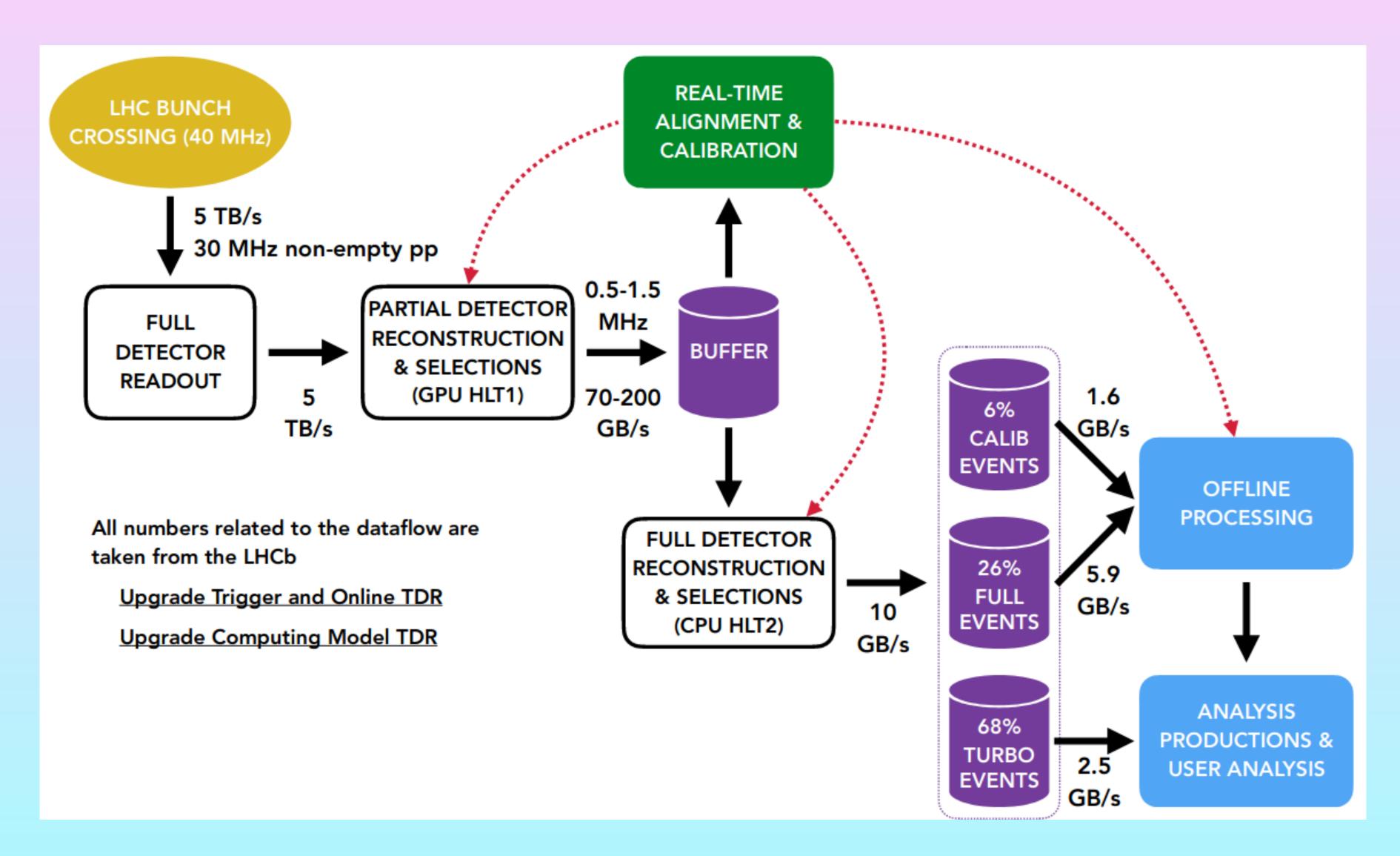
Summary

- Topological Trigger in Run 3 will be based on Neural Networks:
 - Provides robustness against detector effects and sensitivity to long-lived objects
 - Possibility to select BSM physics which would otherwise be disregarded
- Implementing the Neural Networks into the Software requires careful tuning considering limited computational resources

Thank you for your attention!

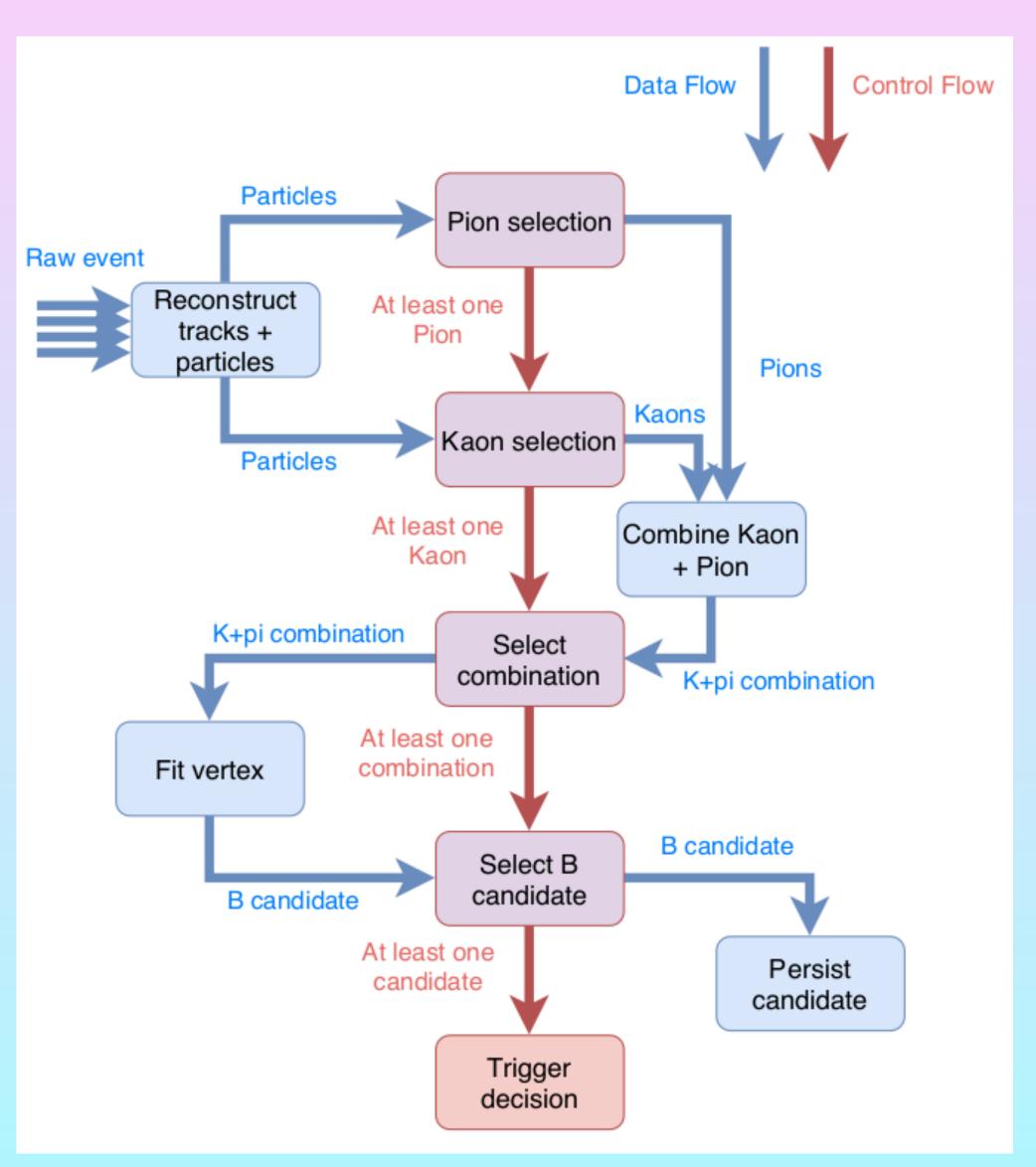
Backup

Data flow in Run 3



Reconstruction Flow in HLT2

- Two important types of algorithms: Particle Filters & Particle Combiners
- Particles and Tracks are reconstructed and then filtered according to certain properties (minimal transverse momentum etc.)
- Particles are then combined to a candidate of a decaying particle
- Combinations and vertices are again filtered according to certain properties
- Combinations are then passed to selection algorithm to obtain Trigger decision



Samples - Detailed Overview

Training Samples

$$\begin{split} A_b^0 &\to (A_c(2625)^+ \to (A_c^+ \to p^+ K^- \pi^+) \pi^+ \pi^-) \mu^- \overline{\nu}_{\mu} \\ B+ &\to (\overline{D}^0 \to K^+ \pi^-) (K_S^0 \to \pi^+ \pi^-) \pi^+ \\ B^0 &\to (D^0 \to K^- \pi^+ \pi^+ \pi^-) (\overline{D}^0 \to K^+ \pi^-) \\ B^+ &\to (\overline{D}^0 \to K^+ \pi^-) \pi^+ \pi^- \pi^+ \\ B_s^0 &\to (D_s - \to K^+ K^- \pi^-) \nu_{\mu} \mu^+ \\ B^+ &\to (D^{\star+}(2010) \to (D^0 \to K^- \pi^+) \pi^+) (\overline{D}^0 \to K^+ \pi^-) \\ B^0 &\to (D^{\star-} \to \pi^- (\overline{D}^0 \to K^+ \pi^-)) (\tau^+ \to \pi^+ \pi^+ \pi^- \overline{\nu}_{\tau}) \nu_{\tau} \\ B_c^+ &\to (J/\psi(1S) \to \mu^+ \mu^-) (\tau^+ \to \mu^+ \nu_{\mu} \overline{\nu}_{\tau}) \nu_{\tau} \\ B^- &\to (D^0 \to K^- \pi^+) (\tau^- \to \mu^- \nu_{\tau} \overline{\nu}_{\mu}) \overline{\nu}_{\tau} \\ B^+ &\to (\overline{D}^{\star 0} \to \pi^0 ((\overline{D}^0 \to K^+ \pi^-)) (\tau^+ \to \pi^+ \pi^- \pi^+ \overline{\nu}_{\tau}) \nu_{\tau} \end{split}$$

$$B^{+} \rightarrow \mu^{+}\mu^{-}\mu^{+}\nu_{\mu}$$

$$\Lambda_{b}^{0} \rightarrow p^{+}\mu^{-}\overline{\nu}_{\mu}$$

$$B^{+} \rightarrow K^{+}\mu^{+}\mu^{-}\mu^{+}\mu^{-}\mu^{+}\mu^{-}$$

$$B_{c}^{+} \rightarrow (J/\psi(1S) \rightarrow \mu^{+}\mu^{-})\mu^{+}\nu_{\mu}$$

$$B_{s}^{0} \rightarrow K^{-}\nu_{\mu}\mu^{+}$$

$$B_{s}^{0} \rightarrow K^{-}(\tau^{+} \rightarrow \mu^{+}\nu_{\mu}\overline{\nu}_{\tau})\nu_{\tau}$$

$$B^{+} \rightarrow \pi^{+}\pi^{-}K^{+}$$

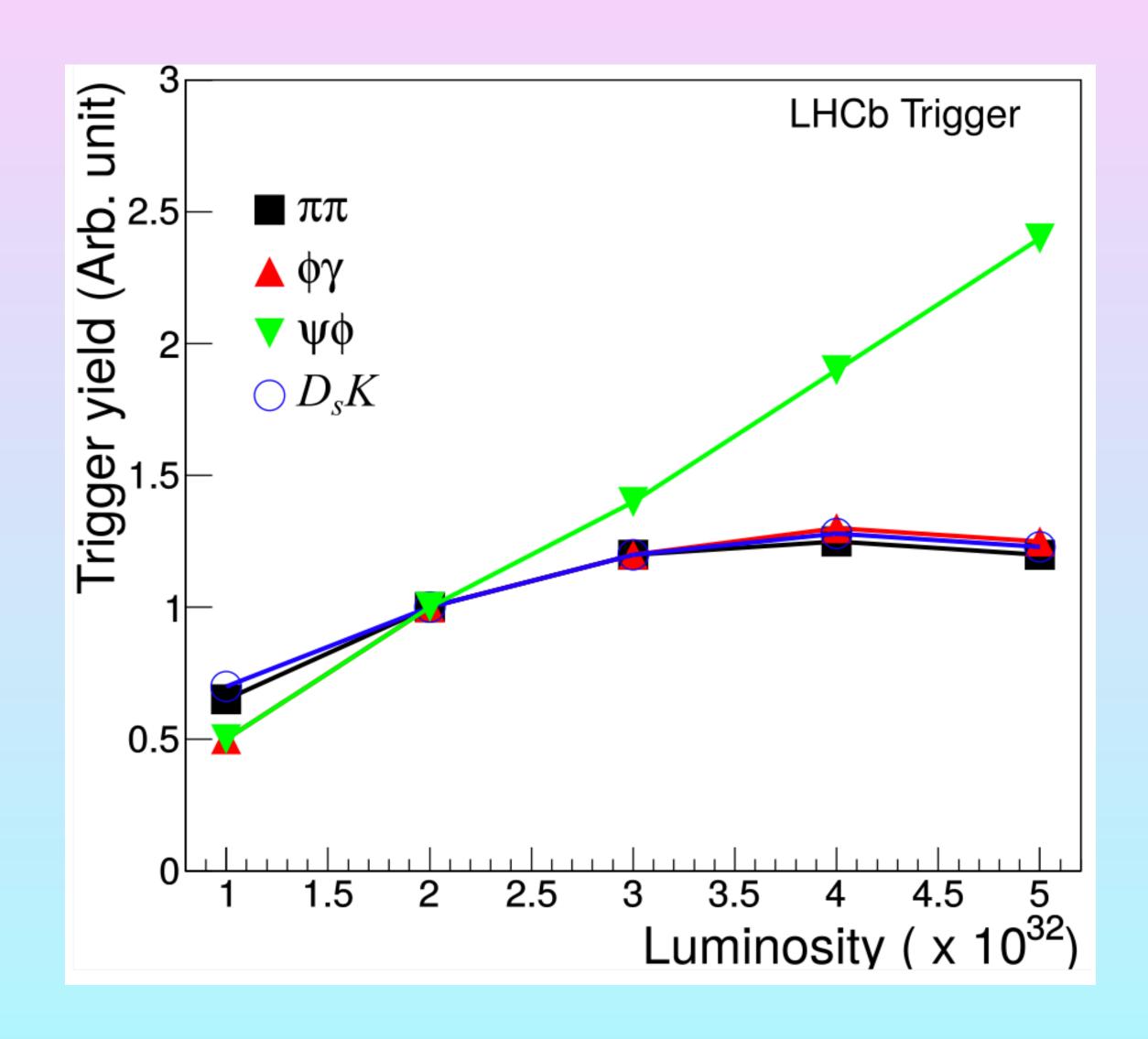
$$B^{0} \rightarrow (K^{*0}(892) \rightarrow K^{+}\pi^{-})\gamma$$

$$B^{-} \rightarrow p^{+}\overline{p}^{-}(\tau^{-} \rightarrow \mu^{-}\overline{\nu}_{\mu}\nu_{\tau})\overline{\nu}_{\tau}$$

$$B^{-} \rightarrow p^{+}\overline{p}^{-}\mu^{-}\overline{\nu}_{\mu}$$

Removal of the Hardware Trigger LO

- L0 used information from the calorimeters and muon chambers
- Based selection on transverse energy information
- Hadronic events required a high threshold in transverse energy to obtain reasonable output rate
- → High threshold cut out high amount of beauty decays
- → Signal yield saturates for increased luminosity
- → Higher luminosity would not provide benefit for hadronic decays



Features - Detailed Table

TwoBody Features ThreeBody Features $\min (p_{T, FS \text{ particles } (1,2)})$ $\min (p_{\mathrm{T, FS particles}} (1,2,3))$ sum ($p_{\rm T, \ FS \ particles}$ (1,2,3)) $p_{\mathrm{T, FS \ particles}}$ (1,2) $p_{\mathrm{T, B-Hadron}}$ $p_{\mathrm{T, B-Hadron}}$ $\log \left(\min(\chi^2_{\text{IP, FS particles }(1,2,3)} \right)$ $\left(\min(\chi_{\text{IP, FS particles }(1,2)}^2\right)$ $\log \left(\max(\chi^2_{\text{IP, FS particles}}(1,2)) \right)$ $\log \left(\max(\chi^2_{\text{IP, FS particles }(1,2,3)} \right)$ $\log \left(\chi_{\rm FD, B-Hadron}^2\right)$ $\log \left(\chi_{\rm FD, B-Hadron}^2\right)$ $\log \left(\chi^2_{\text{Vertex, B-Hadron}}\right)$ $\log \left(\chi^2_{\text{Vertex, B-Hadron}} \right)$ DOCA (B-Hadron) DOCA (B-Hadron) min $(p_{\rm T, FS particles} (1,2)$ sum ($p_{\rm T, FS}$ particles (1,2)) DOCA (TwoBody) $\log \left(\chi_{\rm FD, TwoBody}^2\right)$ $\chi^2_{ m Vertex}$, TwoBody $\log\left(\chi_{\mathrm{IP, TwoBody}}^2\right)$ $p_{\mathrm{T, TwoBody}}$