

# Fast Simulations at LHCb

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ECHEP Workshop

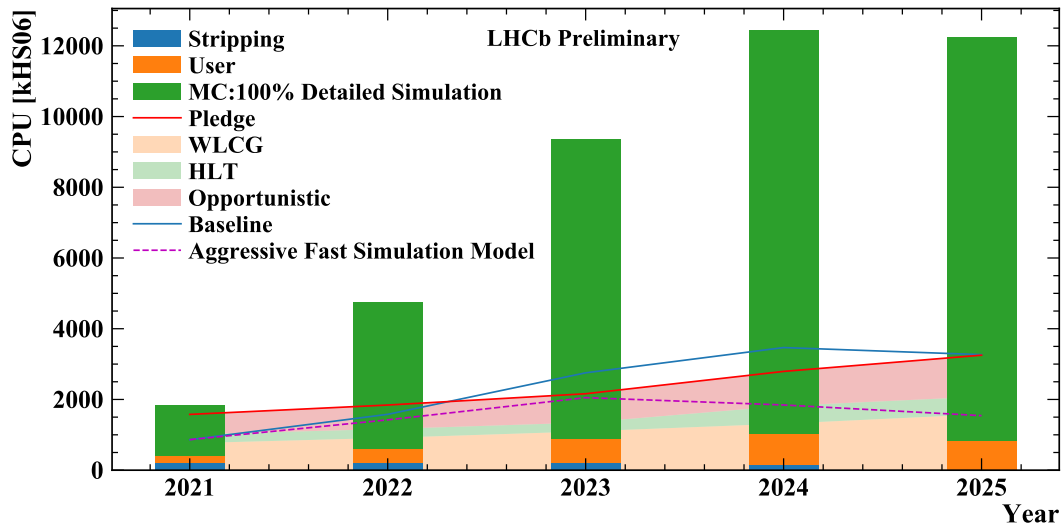
Feb 17, 2020

With many thanks to P. Ilten, M. Kreps



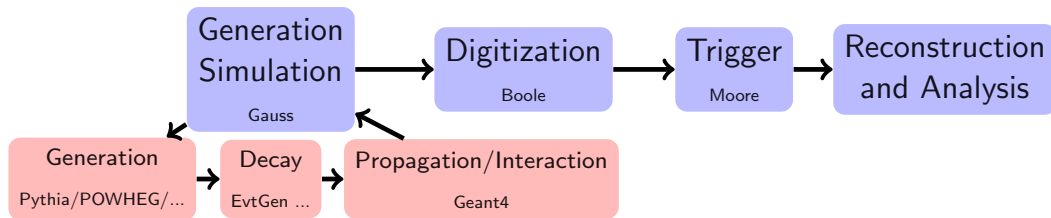
# Future Simulation Needs

[LHCb-FIGURE-2019-018]



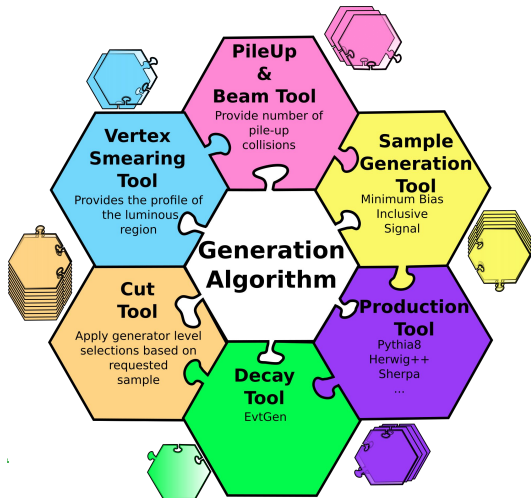
# Simulation Strategy

- ▶ Projected CPU usage is too high for Upgrade and Upgrade II era
- ▶ Must actively pursue other simulation options
- ▶ Reminder: LHCb simulation sequence



- ▶ Efforts to speed up all portions ongoing

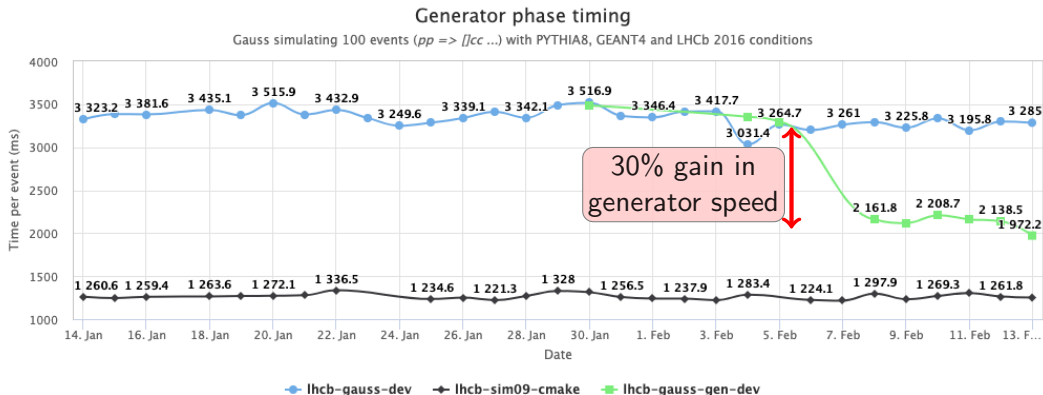
# Generation Phase



- ▶ Highly modular generation phase → support the large physics programme of LHCb
- ▶ Incorporated in LHCb software, which is built around Gaudi Algorithms

# Opportunity: Generation Speedups

- ▶ D. Konstantinov - run Valgrind on Pythia → discovered many issues
- ▶ Many lexical\_cast calls to PDF sets
- ▶ Elimination of these + other improvements give large gain in speed for LHCb Simulation

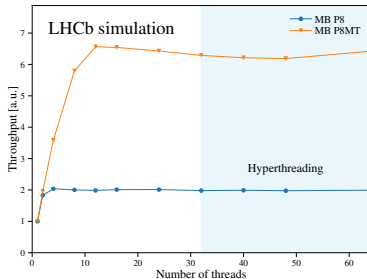
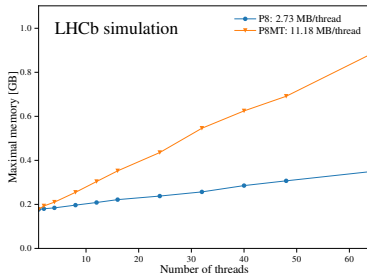


## Opportunity: Forced Hadronization (With thanks to P. Ilten)

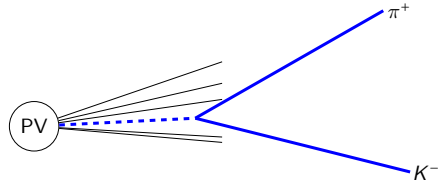
- ▶ At LHCb, most common method is to generate min bias events until signal decay is found
- ▶ Look at all existing signal files, ask how many signal per minimum bias event
- ▶ Using forced hadronization, would get 1/multiplicity speedup
- ▶ On average, could gain  $2.7 \times 10^4$  in timing

PDG ID	name	mult	speedup	decfiles
4224	Sigma_c*++	8.9e-04	1.1e+03	1
4114	Sigma_c*0	8.6e-04	1.2e+03	1
4112	Sigma_c0	5.3e-04	1.9e+03	2
10431	D_s0*+	7.3e-05	1.4e+04	5
10441	chi_c0	5.3e-03	1.9e+02	5
100443	psi(2S)	1.7e-04	6.0e+03	6
4232	Xi_c+	1.3e-03	7.6e+02	7
...				
553	Upsilon(1S)	6.8e-05	1.5e+04	7
3222	Sigma+	4.2e-01	2.4e+00	7
435	D_s2*+	4.0e-05	2.5e+04	8
445	chi_c2	6.5e-03	1.5e+02	8
200553	Upsilon(3S)	1.6e-05	6.2e+04	8
5232	Xi_b0	8.5e-05	1.2e+04	32
310	K_S0	3.2e+00		35
...				
5132	Xi_b-	9.0e-05	1.1e+04	69
411	D+	7.9e-02	1.3e+01	100
431	D_s+	2.8e-02	3.6e+01	106
413	D*(2010)+	5.4e-02	1.8e+01	126
541	B_c+	2.1e-05	4.8e+04	415
5122	Lambda_b0	6.8e-04	1.5e+03	435
531	B_s0	1.8e-03	5.4e+02	676
511	B0	8.0e-03	1.2e+02	856
521	B+	7.9e-03	1.3e+02	984
		current <time>/event [s]	1.7e+02	
		nominal <time>/event [s]	6.4e-03	
		<speedup>	2.7e+04	

- ▶ Future of LHCb simulation: Gauss on Gaussino
- ▶ Core principles of Gaussino
  - ▶ LHCb independent core framework
  - ▶ Build on modularity of Gauss
  - ▶ Incorporate task-based parallelism of Gaudi
  - ▶ Interface to Geant4 and Pythia8
- ▶ First developments show promising results
  - ▶ 2016 LHCb Conditions before previous slide's improvements
  - ▶ Blue: Shared Pythia 8 configuration
  - ▶ Orange: Thread local Pythia8 configuration

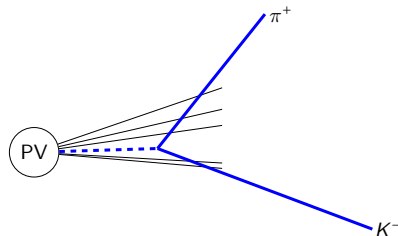


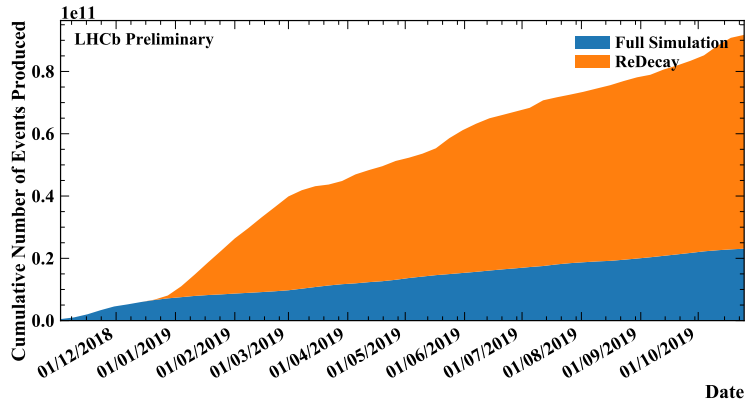
- ▶ Goal: use more efficiently CPU used per event
- ▶ Method: split particles into two groups: those involved in signal process and those from the rest of the event
  - ▶ Generate MC event, store signal origin and momentum
  - ▶ Remove signal and decay products, pass rest of event through the simulation framework
  - ▶ Generate signal decay, merge with rest of the event
  - ▶ Repeat previous step  $N_{\text{ReDecay}}$  times
- ▶ Spend  $\mathcal{O}(90\%)$  time simulating signal
- ▶ Independent of Generator
- ▶ Note: Only useful for generating specific signals





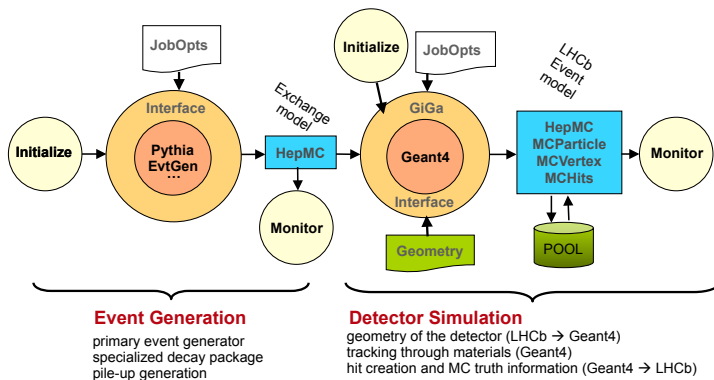
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- ▶ Over the past year, ReDecay has been validated and adopted by Physics WGs
- ▶ 10-50 $\times$  faster  $\rightarrow$  able to generate MC for analyses requiring high statistics samples

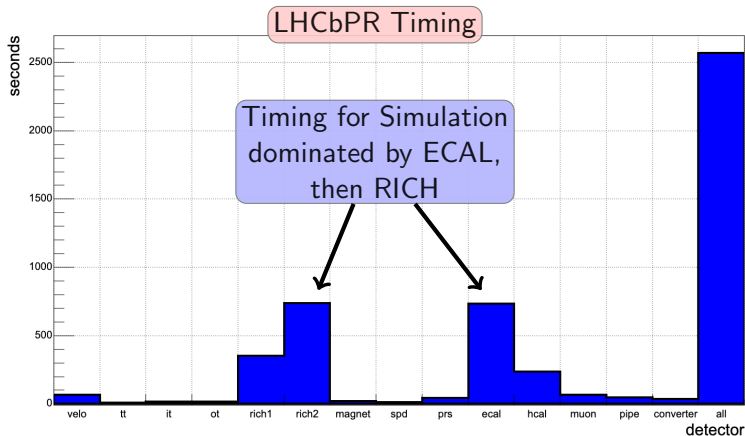
# Simulation Phase



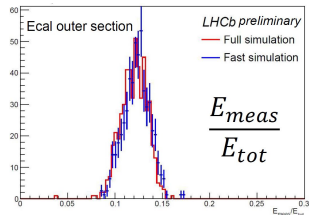
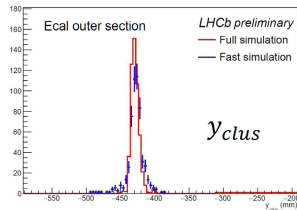
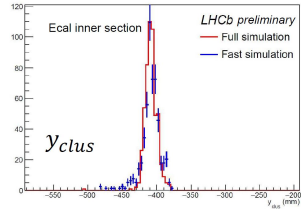
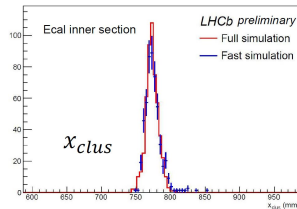
- ▶ After simulation phase, events are passed in the same sequence to the propagation through material

# Timing of Detector Simulation

- Use LHCbPR framework to assess the timing per event for detailed simulation



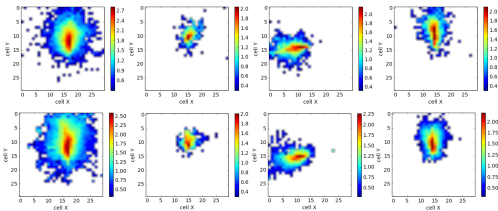
- ▶ Idea: Provide a fast simulation option which automatically replaces Geant4 hits with a hit collection from library, as a function of discrete  $E, \theta, \phi$  “nodes”
- ▶ Not a collection of cell images, but rather energy deposit in smaller points, hence called “point library”
- ▶ Timing for lookup and transformation of points negligible



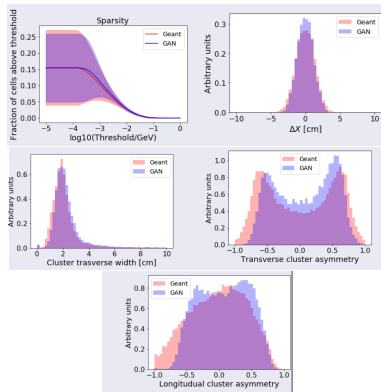
- Use GANs to generate shower image on ECAL face

GEANT Simulated

$\log_{10}(\text{cell energy})$

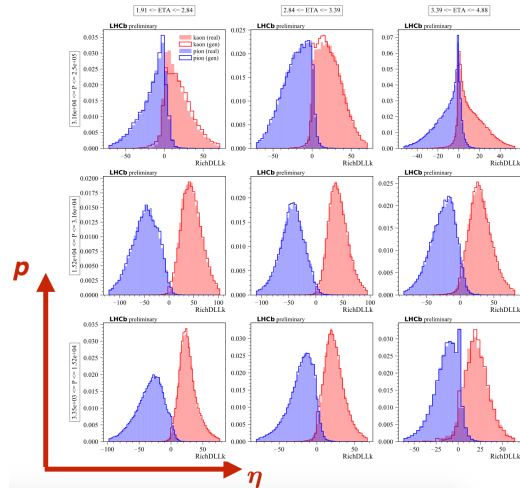
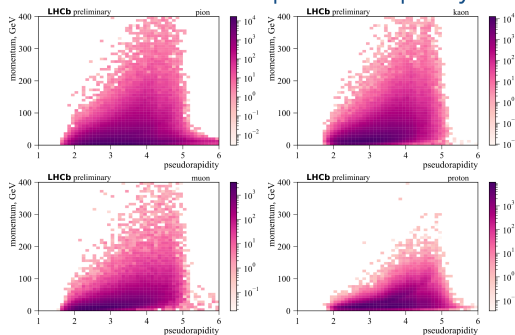


GAN Generated

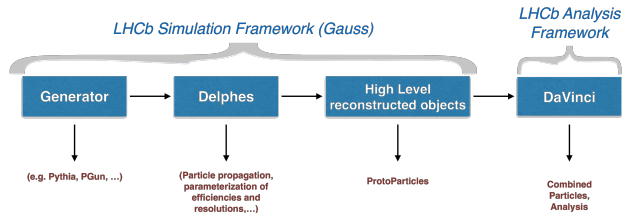


- While input distributions are well reproduced, higher level variables are not necessarily → Model needs to know about these

- ▶ Learn PID response given only particle type and kinematics
- ▶ Based on Cramer GAN, trained on calibration data



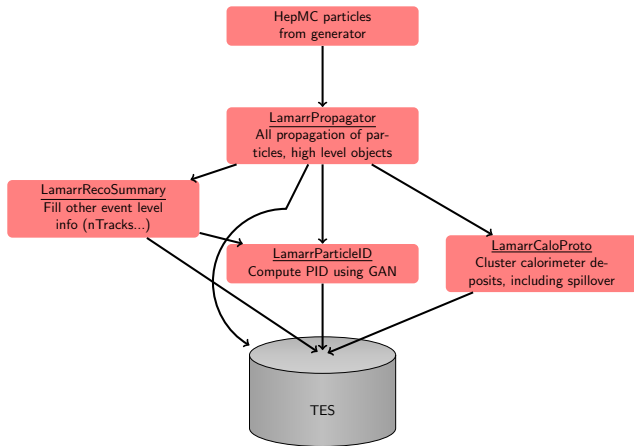
- ▶ “Aggressive” simulation model necessitates 20% ultra-fast parametrized simulation
- ▶ Existing solution: [Delphes](#)
  - ▶ LHCb geometry not naively supported → implemented dipole magnet instead of torroid
  - ▶ Calorimeter segmentation not cartesian → implemented new calorimeter segmentation
  - ▶ Interfacing within Gauss required extra steps



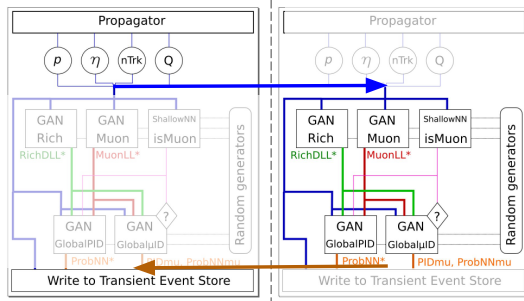


- ▶ During internal review of the adoption of Delphes, we noted that there were many duplications and complications of event processing frameworks
- ▶ The cost/benefit analysis for using Delphes within Gaudi was considered too high compared to using simple parameterization tools (see next slides)
  - ▶ Non-trivial interfacing
  - ▶ External constraints from both sides for data preparation and timing
- ▶ We therefore switch to a fully in-house implementation of parameterizations → Lamarr

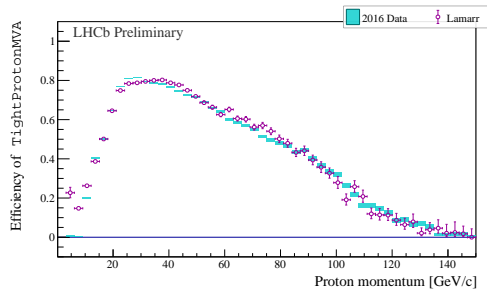
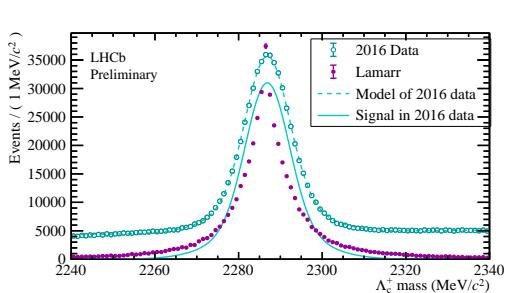
- ▶ Propagator redesigned: Propagate MC particle first to all points of interest, then smear and apply efficiencies.
- ▶ Use Inverse Cumulative Method to sample track info ( $\chi^2_{\text{track}}$ ,  $C_{ij}^{\text{track}}$ , fake track probability...) → mitigates binning dependence of parameterization, and large gain in speed
- ▶ Calorimeter parameterization ported from Delphes Card to simple python lists



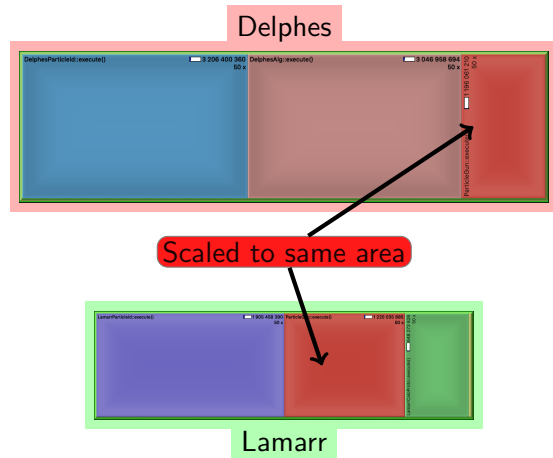
- ▶ Charged particle PID variables in LHCb can vary with occupancy and other event level variables not available at ultra-fast simulation level
- ▶ Solution: sample non-signal variables (e.g. nTracks) with random input
- ▶ Once input defined, use stacked GANs evaluated in TensorFlow to form PID information without calorimeter inputs
- ▶ Limits calls to TensorFlow to once per event



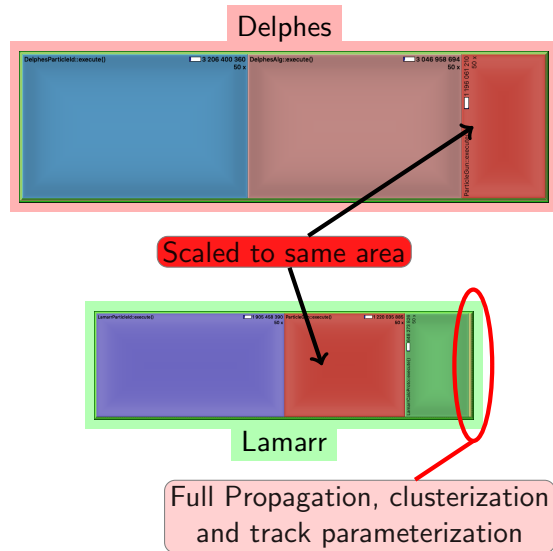
- ▶ Use 2016 Data to validate
- ▶ Example:  $\Lambda_b^0 \rightarrow \Lambda_c^+ \mu^- \bar{\nu}_\mu$ ,  $\Lambda_c \rightarrow p K^- \pi^+$



- ▶ Run Valgrind with Cachegrind on 50  $B^0 \rightarrow K^+ K^- \pi^0 (\rightarrow \gamma\gamma)$  events with both Delphes and Lamarr setup
- ▶ Scale call graph such that ParticleGun is the same
- ▶ With improvements, Propagation and high level particle making is now the sliver on the right hand side of the lowest graph
- ▶ Future improvements focus on:
  - ▶ Internal TensorFlow memory management
  - ▶ Faster random number generators for Calorimeter clusterization



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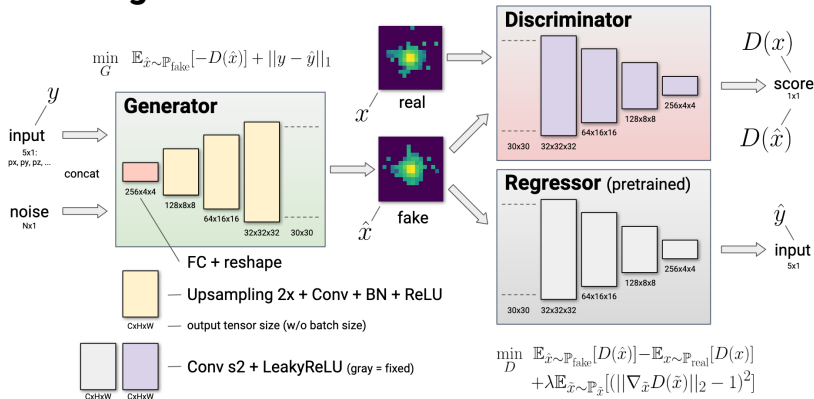
# Synergies and Conclusions

- ▶ Simulation is a necessity to physics experiments
- ▶ The future computing needs of LHCb necessitate fast simulation framework options
- ▶ While LHCb simulation is specialized, synergies exist and are important
  - ▶ Generator level improvements → everyone benefits
  - ▶ Geant4 → mixing of detailed detector simulation with parametric simulation and ML techniques
  - ▶ Ultra-fast simulation is not an HL-LHC problem for LHCb → experience being gained now
  - ▶ Mix and match solutions should exist
- ▶ Exploiting synergies will benefit all

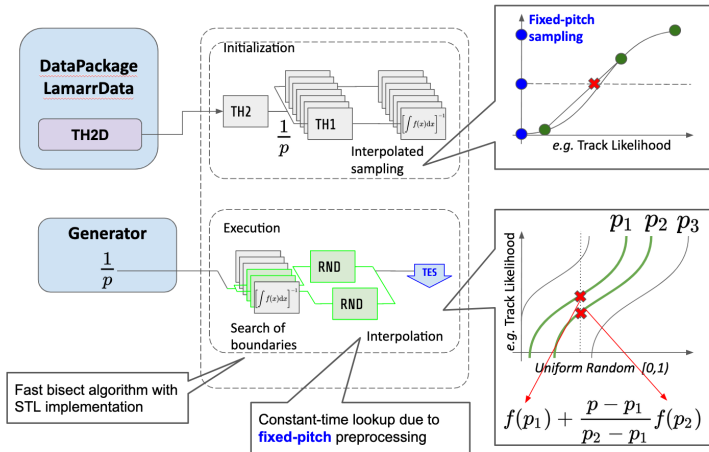
## Backup Slides



## Training scheme



## HistSampler at work



# What LHCb Simulates

- ▶ For LHCb searches, need to generate specific decays for specific searches
- ▶ Example,  $R(D^*)$  needs to generate not only the signal, but also specific backgrounds
- ▶ For some analyses, MC stats is the limiting systematic uncertainty

$R(D^*), \tau \rightarrow \pi\pi\pi(\pi^0)$   
[Phys. Rev. Lett.120\(2018\) 171802](#)

Source	$\delta R(D^{*-})/R(D^{*-})[\%]$
Simulated sample size	4.7
Empty bins in templates	1.3
Signal decay model	1.8
$D^{**} \tau \nu$ and $D_s^{**} \tau \nu$ feeddowns	2.7
$D_s^+ \rightarrow 3\pi X$ decay model	2.5
$B \rightarrow D^{*-} D_s^+ X$ , $B \rightarrow D^{*-} D^+ X$ , $B \rightarrow D^{*-} D^0 X$ backgrounds	3.9
Combinatorial background	0.7
$B \rightarrow D^{*-} 3\pi X$ background	2.8
Efficiency ratio	3.9
Normalization channel efficiency (modeling of $B^0 \rightarrow D^{*-} 3\pi$ )	2.0
Total uncertainty	9.1