

Ghost rate study of the tracking at LHCb

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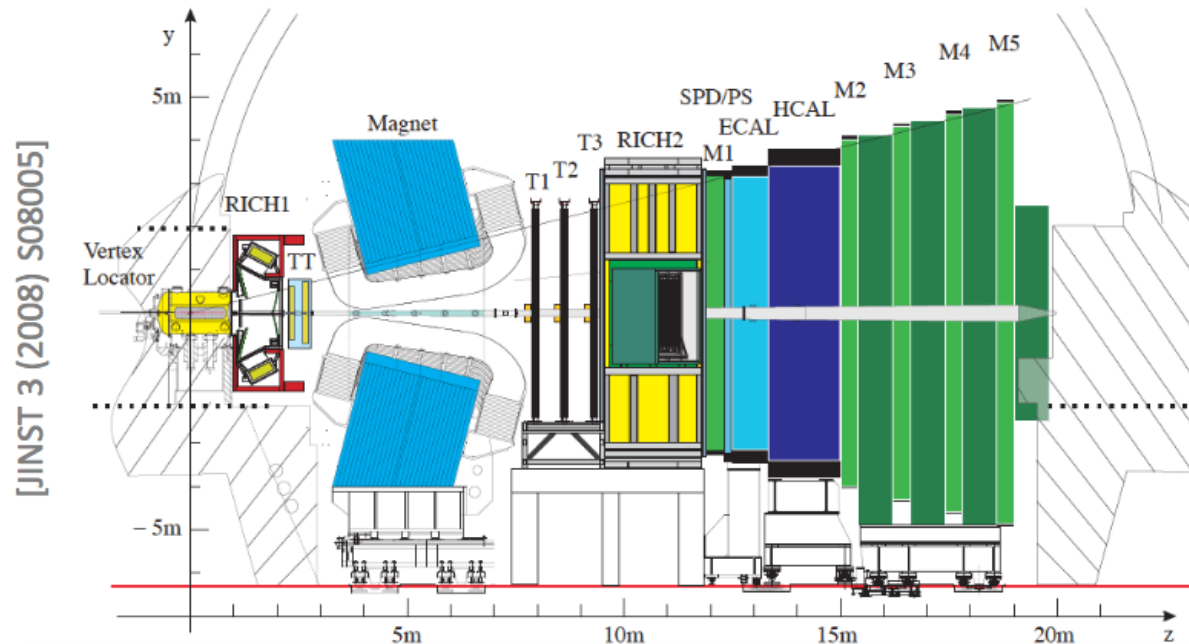


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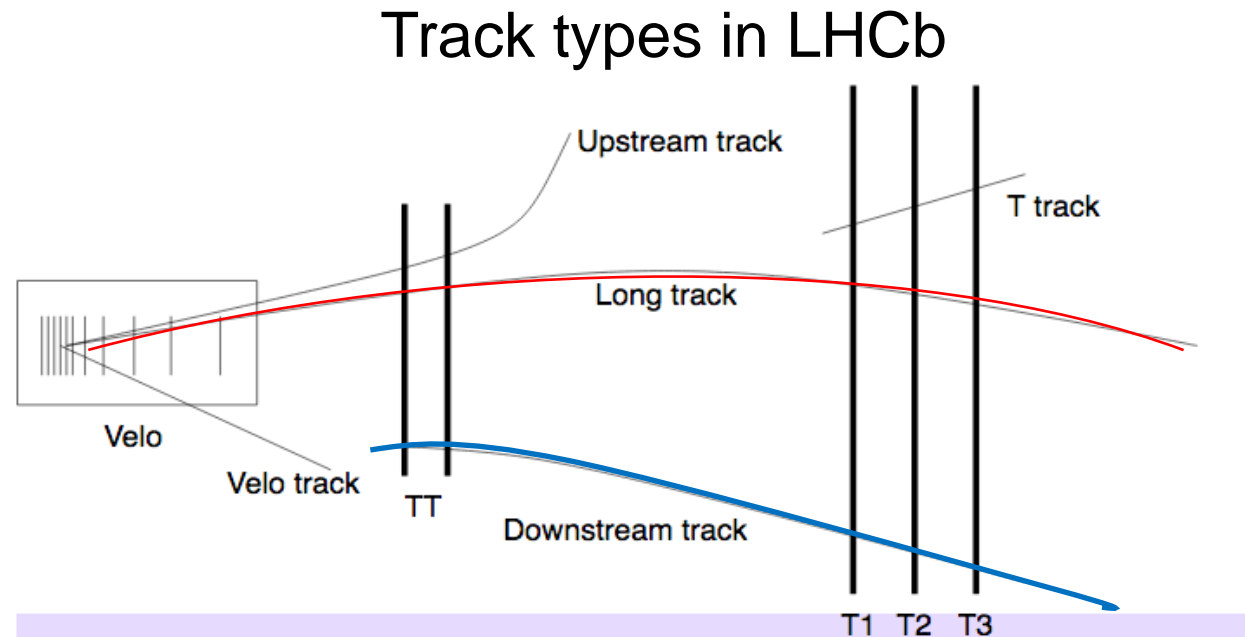
LHCb detector

- ❑ Single arm forward spectrometer: $2 < \eta < 5$
- ❑ Excellent detector performance:
 - **Track reconstruction efficiency >95%**
 - Momentum resolution $dp/p \sim [0.5\%-1\%]$
 - Excellent particle identification



Track reconstruction

- The LHCb tracking reconstruction contains two steps:
 - Track finding - pattern recognition algorithm
 - Track fitting - Kalman filter

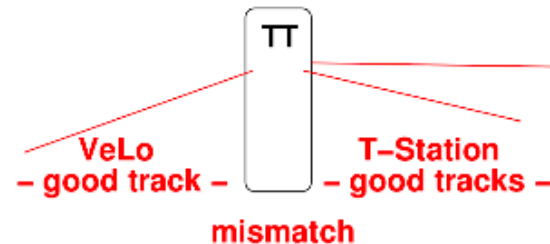
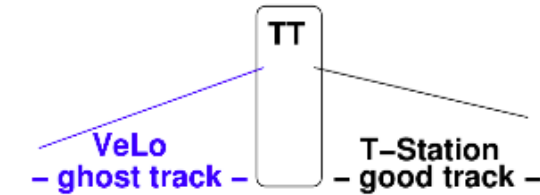
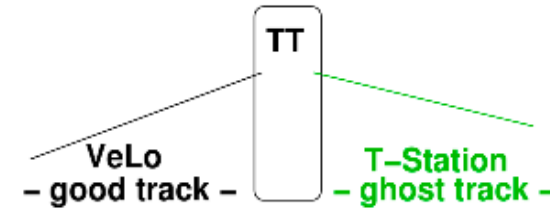


Ghost sources

□ Long tracks mis reconstruction:

- **Hadronic interaction with detector material– main sources**
- T station segments mis-reconstruction
- VELO segments mis-reconstruction
- mis-matched between VELO and T-station

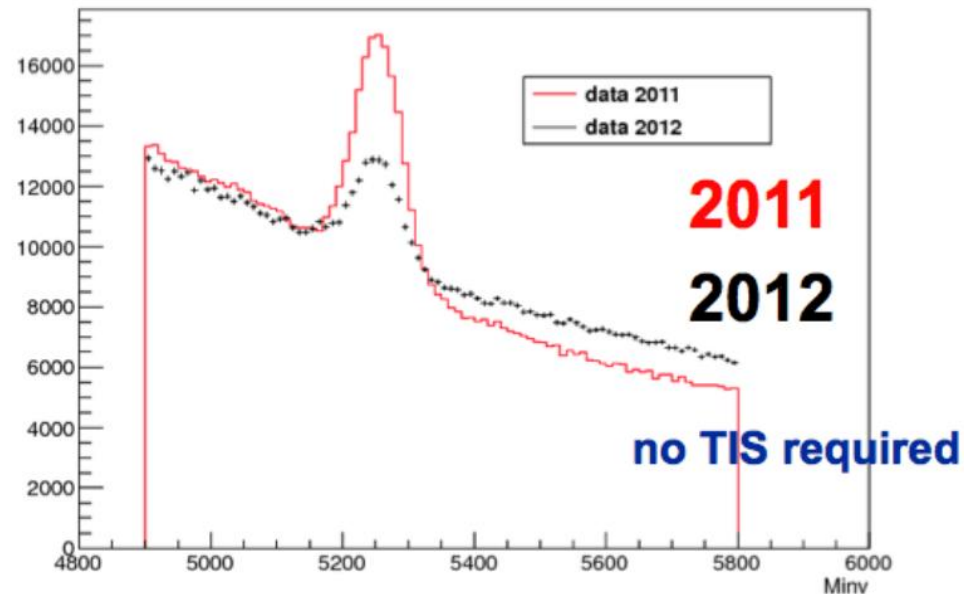
□ Ghost tracks: Low probability of χ^2 from the track fit and missing hits



Motivation

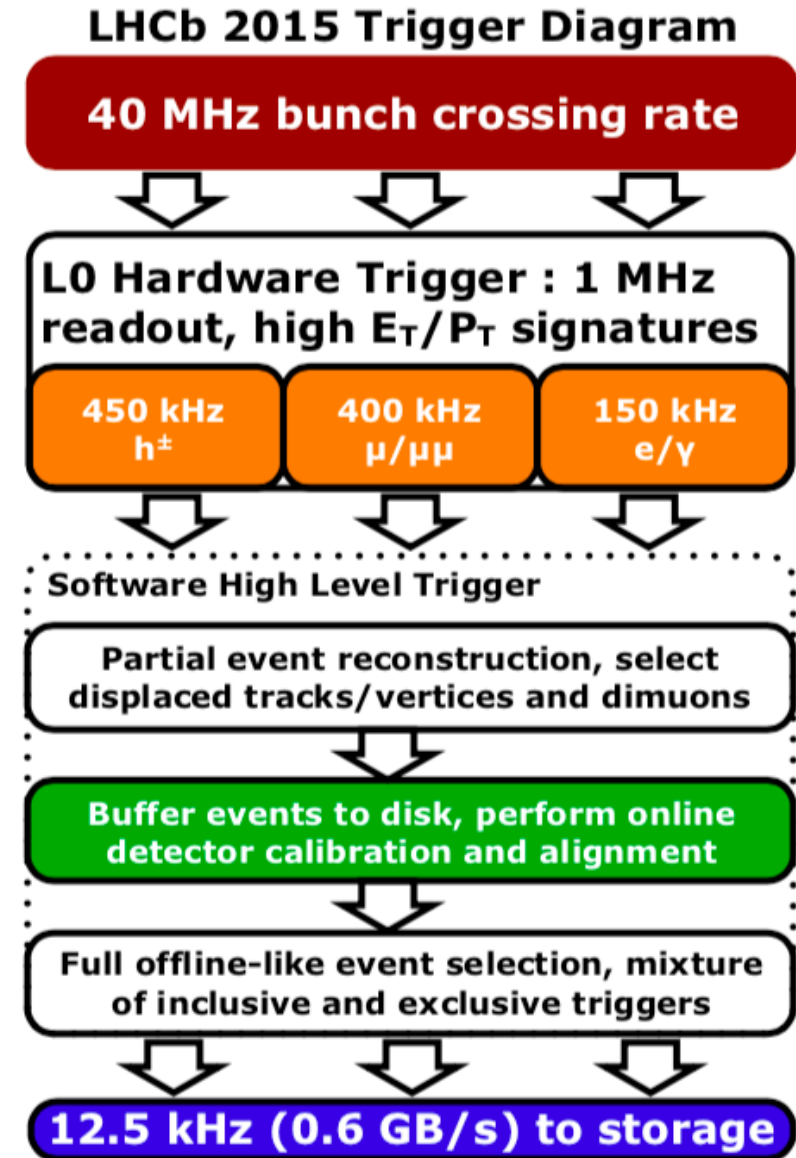
- 2012 data show more noise and worse S/B ratios than 2011 data
 - Problems are caused by background from mis-reconstructed **ghost tracks**

$B \rightarrow hh$ channel in $B_s \rightarrow \mu\mu$ analysis



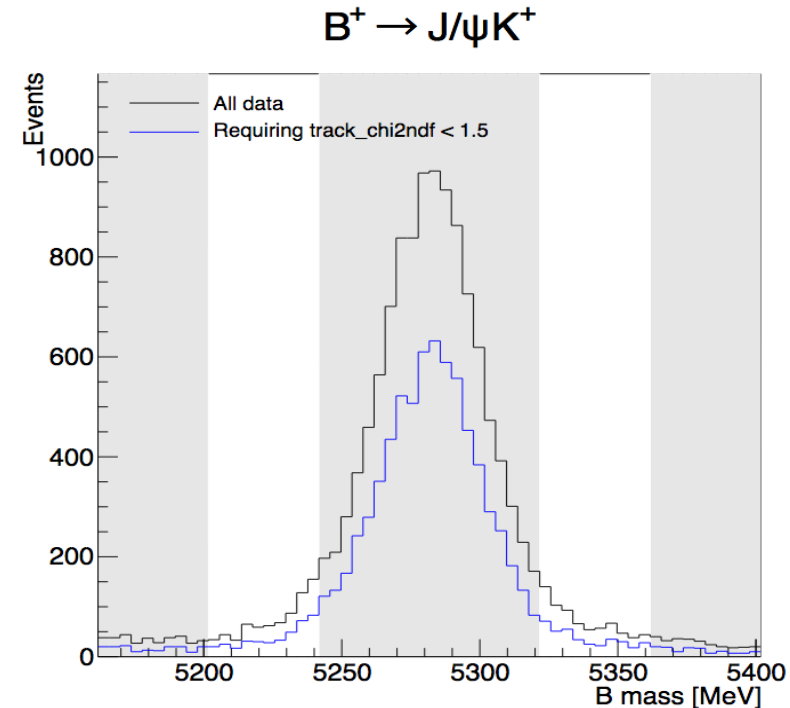
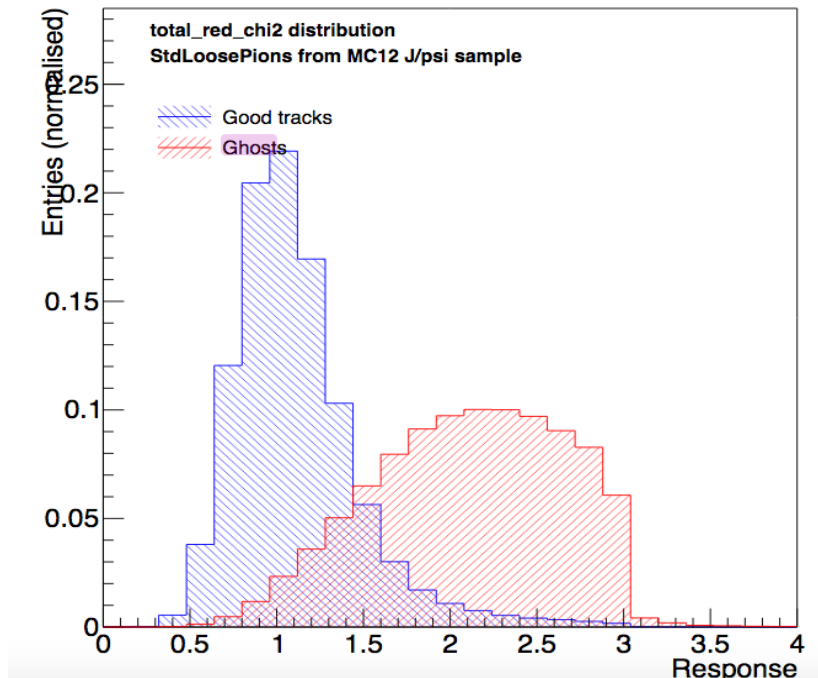
Reconstruction sequence in Run II

- Reconstruction in two stages
 - Fast stage (HLT1): partial event reconstruction
 - Full stage (HLT2): full offline-like selection
- The majority of the LHCb tracking algorithms efficiency is limited by the ghost rate
- **Need to make track reconstruction faster with low fake rate**



Previous method: cut on track χ^2/ndof

- ❑ Quality of fit is generally worse for ghosts than for real tracks
- ❑ Require $\chi^2/\text{ndof} < 1.5$
 - Improves S/B ratio
 - Signal efficiency of 63.3%, **Signal loss too high**
- ❑ Need something more efficient to suppress ghost tracks



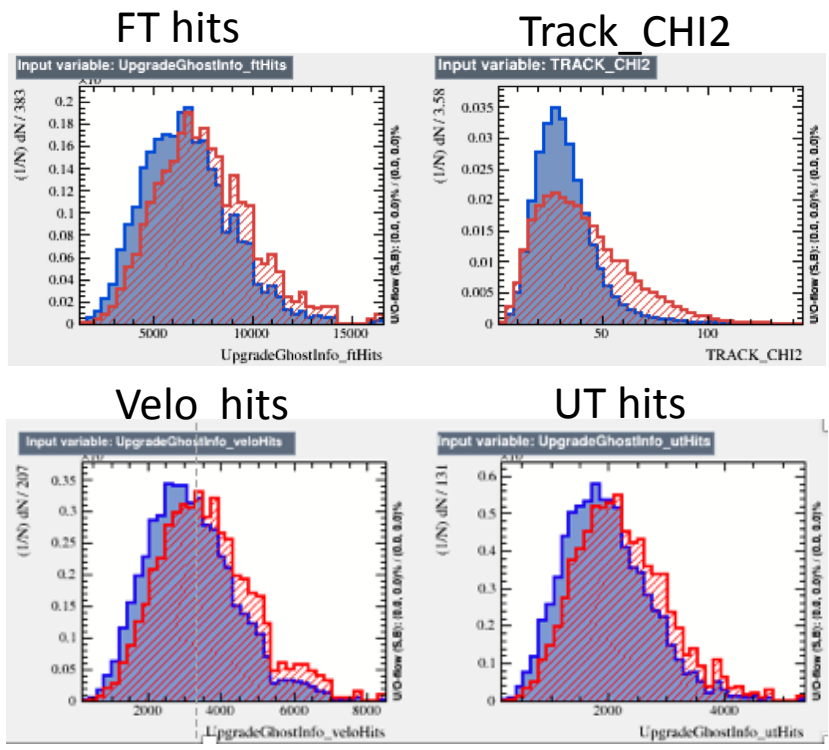
New algorithm: multivariate method

▣ **Multivariate classifier:** MLP neural network combines several weak discriminants into one strong classifier discriminate between ghost and good tracks

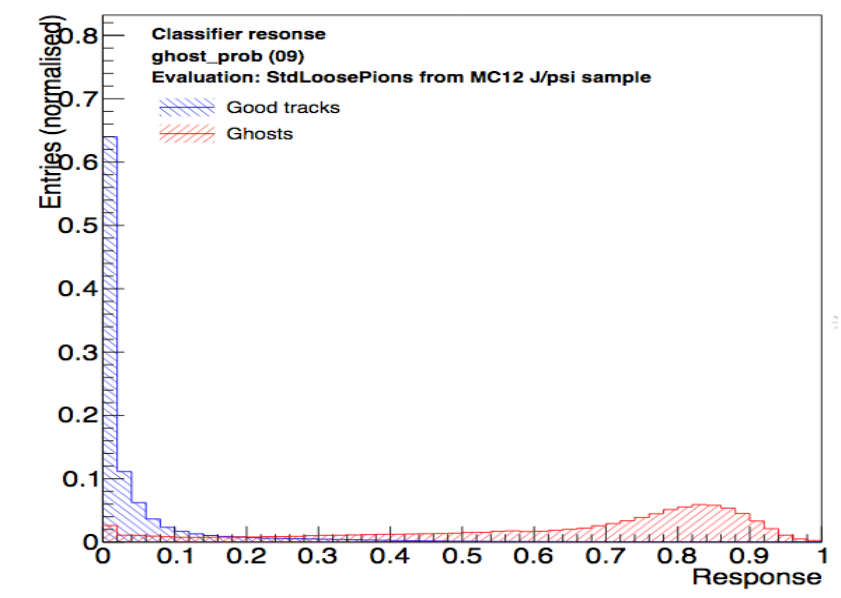
▣ Input files(B inclusive MC):

- Signal: good tracks (>70% hits matched to a MC particle)
- Background: ghost tracks (<70% hits matched to a MC particle)

Ghost probability



Neural network



▣ Various tracking variables, no PID

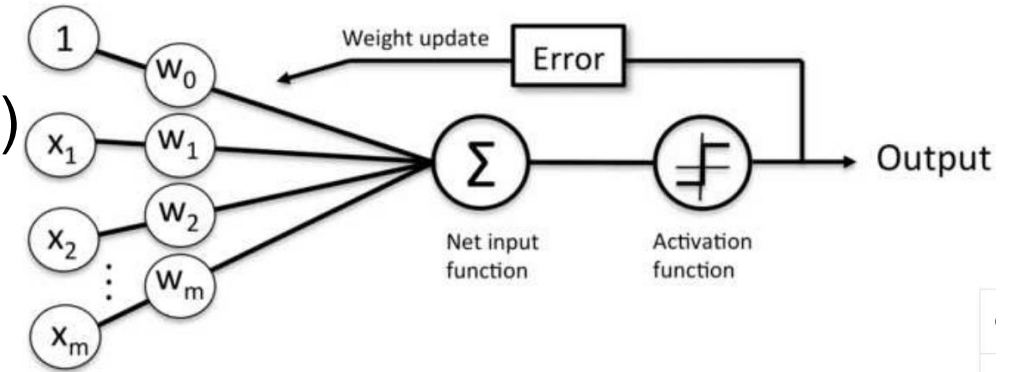
- Track fit quality, number of hits, momentum, detector occupancies

	Velo Track	Upstream Track	Downstream Track	Long Track	T Track
Total hit multiplicite in VELO	✓	✓	✓	✓	✓
Total hit multiplicite in UT	✓	✓	✓	✓	✓
Total Track fit χ^2	✓	✓	✓	✓	✓
Total Track fit NDOF	✓	✓			✓
η	✓	✓	✓	✓	✓
p_T		✓	✓	✓	✓
FitVeloChi2	✓	✓		✓	
FitVeloNDOF	✓	✓		✓	
FitTChi2			✓	✓	✓
FitTNDOF			✓	✓	✓
Observed hits in VELO	✓	✓		✓	
Observed hits in UT		✓	✓	✓	
Observed hits in FT		✓	✓	✓	✓
UT Hits not used for the fit		✓	✓	✓	
Match fit χ^2				✓	
Exp VPHits	✓	✓		✓	
Exp UTHits			✓	✓	
Exp FTHits					✓

Activation function

□ Introduces non-linear factors to compute nontrivial problems using only a small number of nodes

- **Hidden layer:** $\text{Tanh}(\text{RunI}) \rightarrow \text{Sigmoid}(\text{RunII})$
- **Last layer:** exponential function($\text{RunI}+\text{RunII}$)

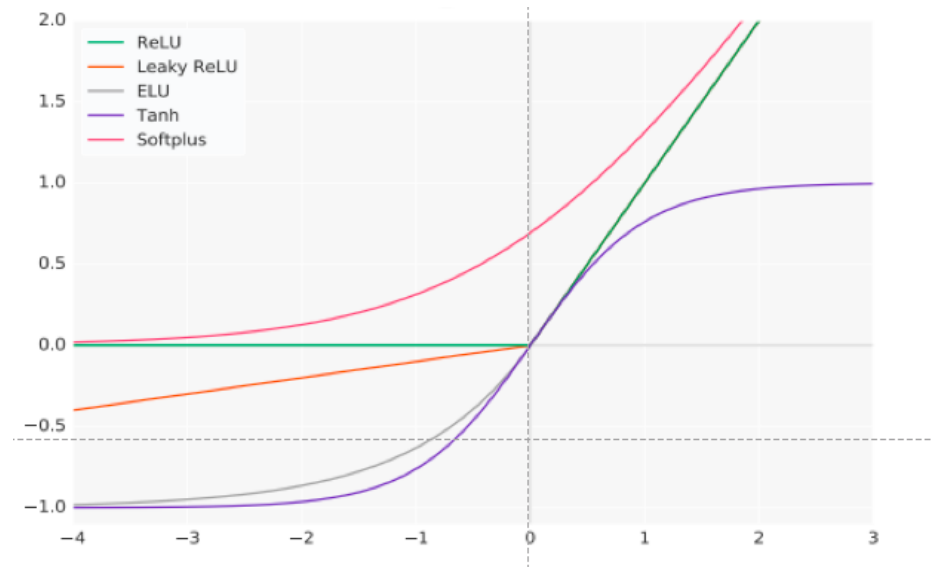


$$\text{Tanh: } f(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$

$$\text{Sigmoid: } f(x) = \frac{1}{1 + e^{-x}}$$

$$\text{Simple-Sigmoid: } f(x) = \frac{1}{1 + x^2}$$

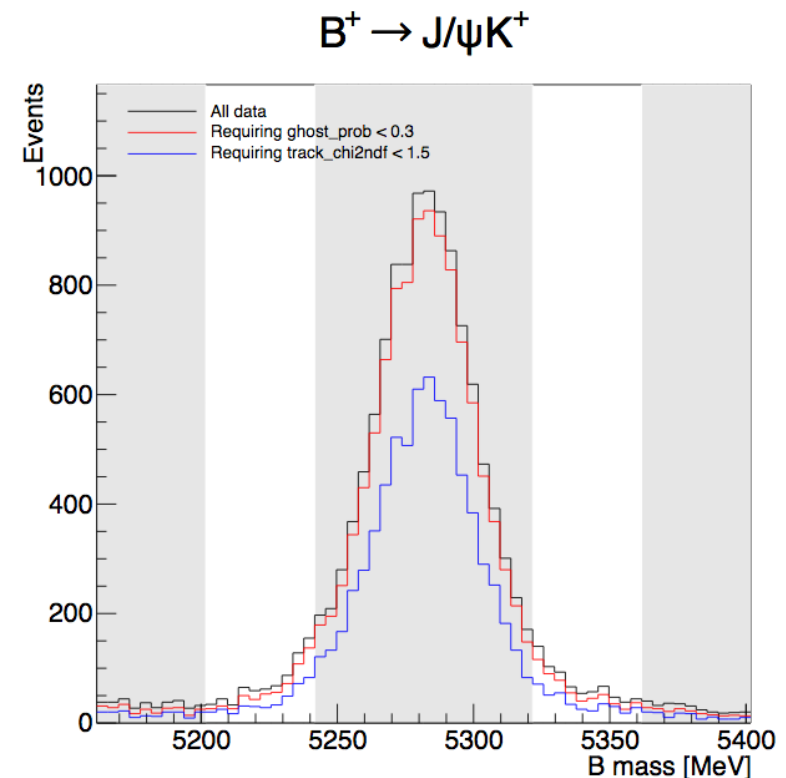
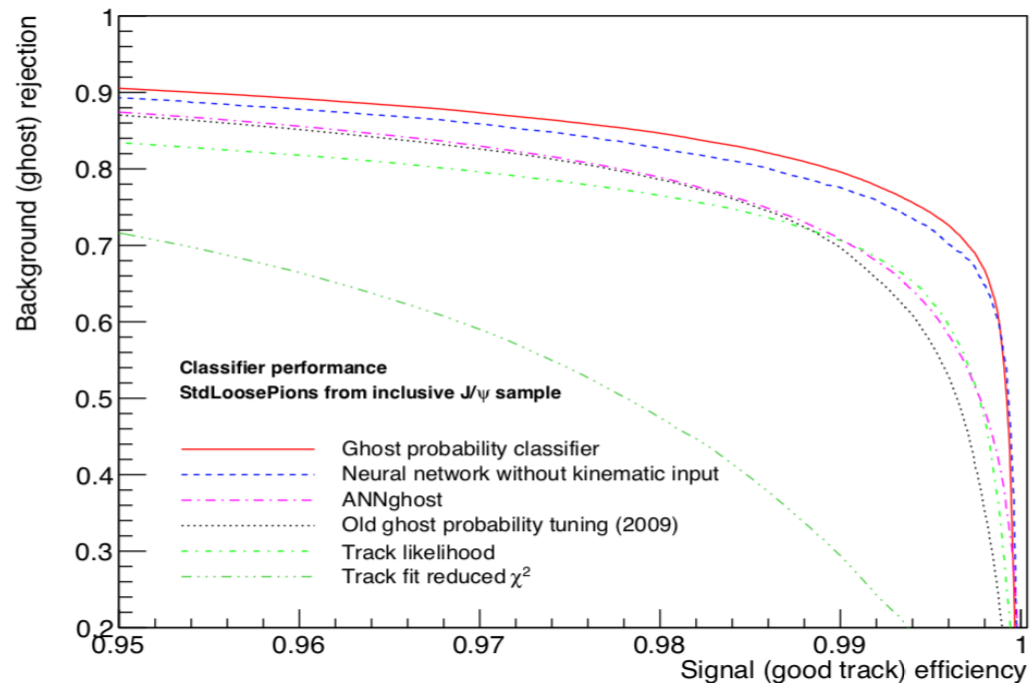
$$\text{ReLU: } f(x) = \max(0, x)$$



Discrimination power

□ Good performance

- For the same signal efficiency, **cutting on ghost probability is most efficient**
- With similar ghost rejections, the signal efficiency of the ghost prob cut is much better
 - ✓ $\chi^2/ndof < 1.5$: 63.3%
 - ✓ Ghost prob < 3 : 96.1%



Impact on trigger CPU consumption

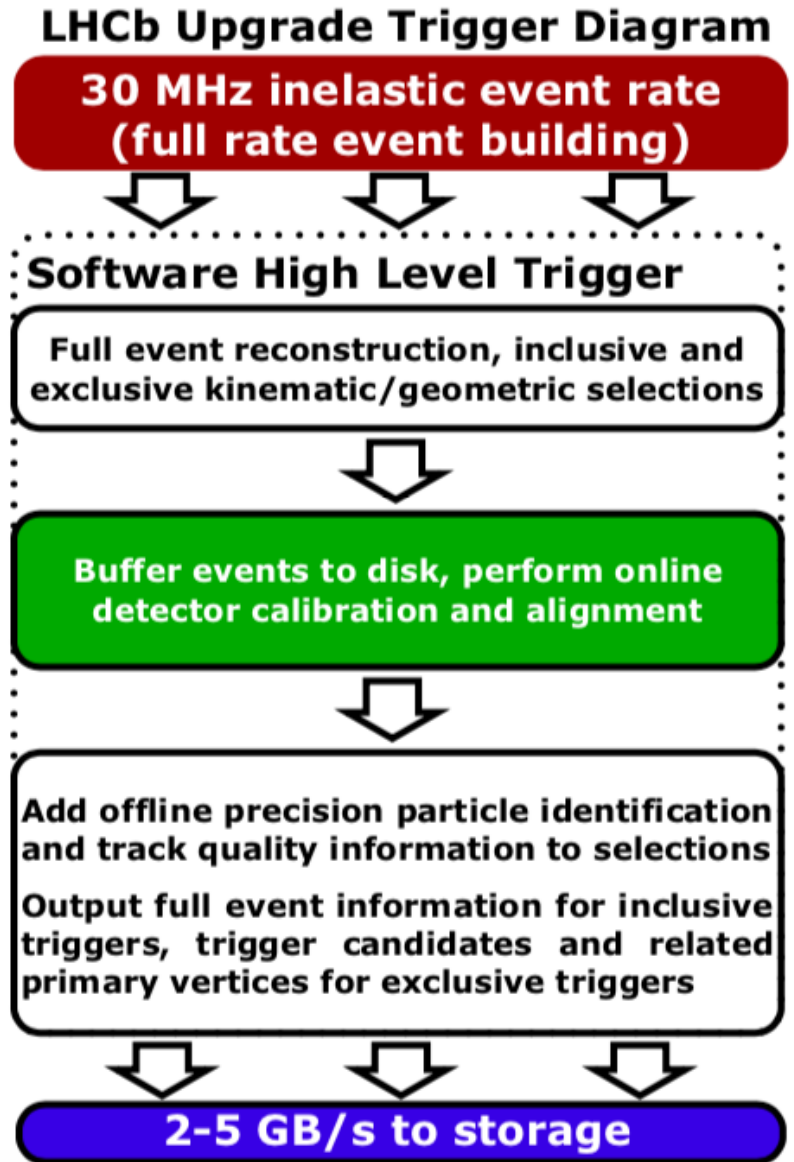
- offline→online quantity:
 - RunI+2015(50ns) : offline quantity
 - 2015(25ns): all tracks in HLT2 had to pass ghost probability selection, **speed-up by a factor ~90**
 - Since 2016: processing in HLT1+HLT2, HLT2 input rate reduced by **4%**
- CPU consumption decreased
 - ghost probability computation itself is **0.2%** of HLT CPU budget
 - **22%** fewer tracks, **23%** less CPU time in RICH PID
 - Overall saving of **16%** entire HLT CPU
- **Ghost probability can not only suppress background but also decrease CPU consumption**

LHCb Upgrade

- CPU timing decreased is expected
 - Detector upgrades to cope with increased luminosity
 - ✓ Run III: $2 \times 10^{33} \text{ cm}^{-2} \text{ s}^{-1}$ (**~5 factor larger than RunII**)
 - Remove hardware trigger, increase output rate to storage

Integrated luminosity

LHC era		HL_LHC era	
Run2 (2015-18)	Run3 (2021-24)	Run4 (2027-30)	Run5+ (2031+)
9 fb^{-1}	23 fb^{-1}	46 fb^{-1}	$>300 \text{ fb}^{-1} ??$



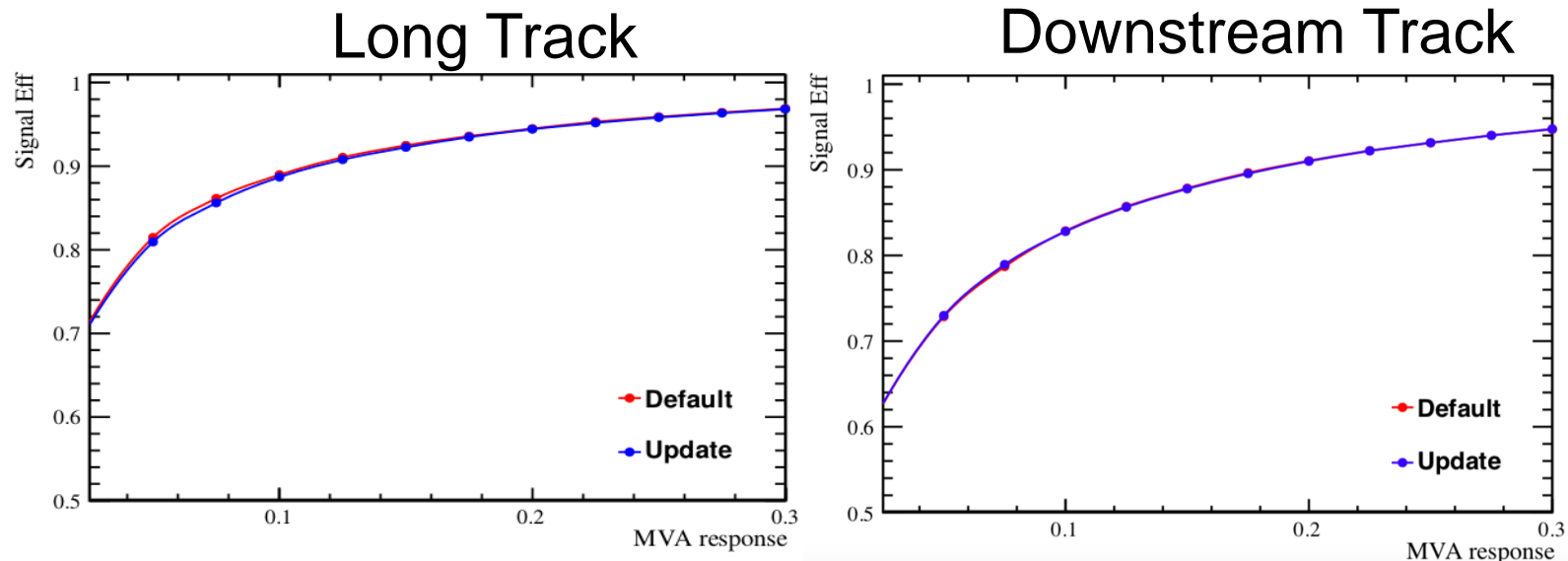
Improvements in the upgrade

□ Upgrade:

- Removed time-consuming variables(hit estimation)
- Activation functions at hidden layer: Sigmoid→ ReLU
 - ✓ Efficient gradient propagation
 - ✓ Fast computation: Only comparison, addition and multiplication
- Activation functions at Last layer: exponential → finite exponential
 - ✓ Replace the exponential operation with the finite number of multiplications

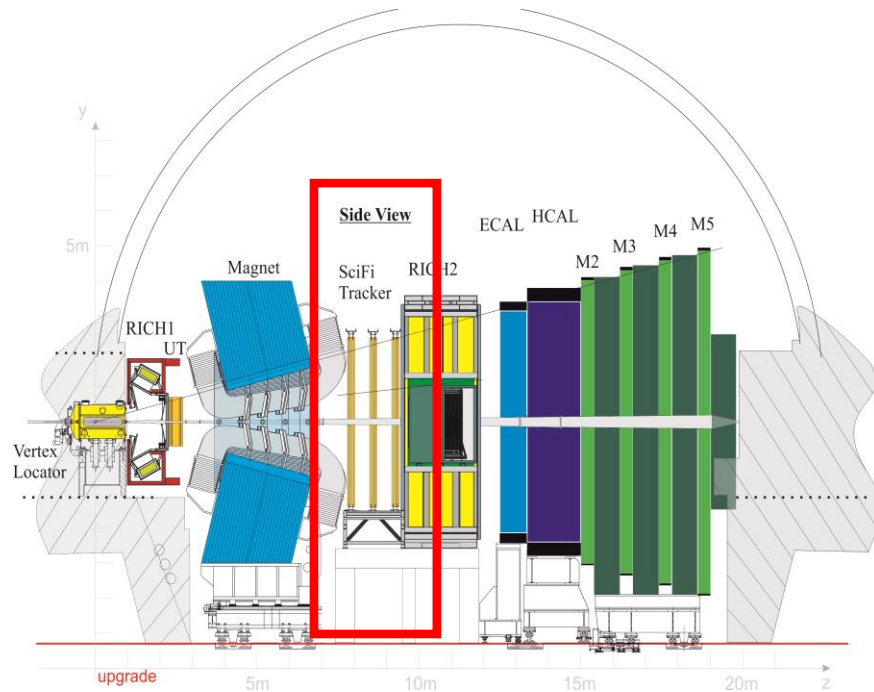
□ The efficiency performance is almost the same

□ Reduced timing of ghost probability algorithm by **97%**



Conclusion

- ❑ Ghost tracks are a significant source of background
- ❑ Ghost probability: important part online reconstruction
 - excellent ghost reduction with minor signal loss
 - Highly decrease the CPU timing
- ❑ Room for improvement
 - fast VDT-like implementations for SIMD types
 - add Scifi track position information in neural work



Backup

Ghost types

