



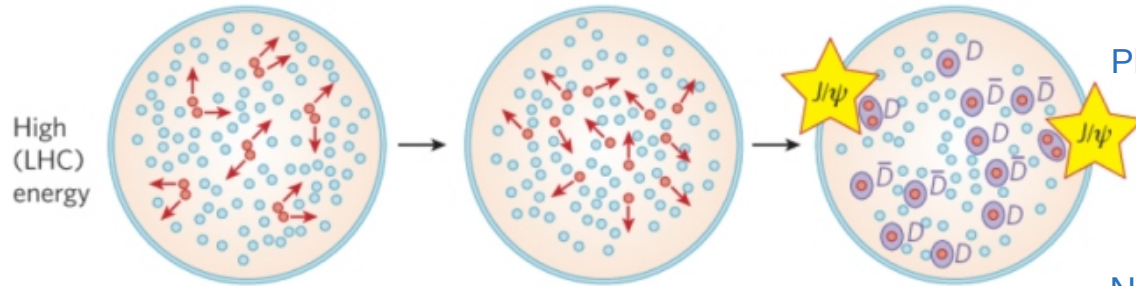
# **$B_c$ measurement in PbPb in ALICE : preliminary study with fast simulation and BDT**

Maurice Coquet (CEA/Irfu), Fabrizio Grosa (CERN),  
Michael Winn (CEA/Irfu)

Aussois Quarkonium & QCD meeting (21/06/2021) 1 / 14

# Motivation

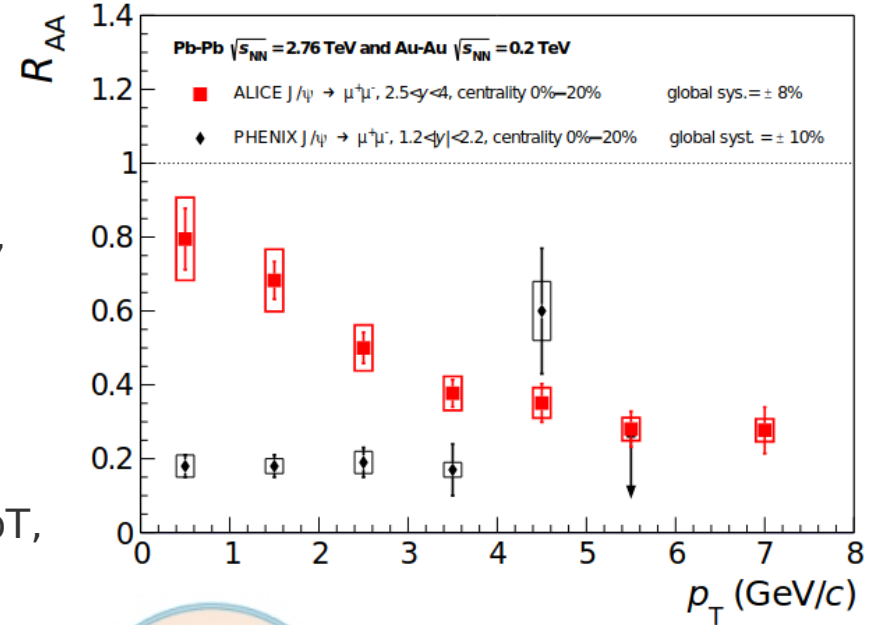
- One signature of the production of hot deconfined medium in heavy ion collisions (Quark Gluon Plasma, QGP) : relative suppression of heavy quarkonia compared to pp
- Observed reduced J/psi suppression at LHC energies compared to RHIC energies  
 → points towards a regeneration mechanism at low  $p_T$ , recombination of cc pairs



Phys.Lett. B734 (2014) 314-327

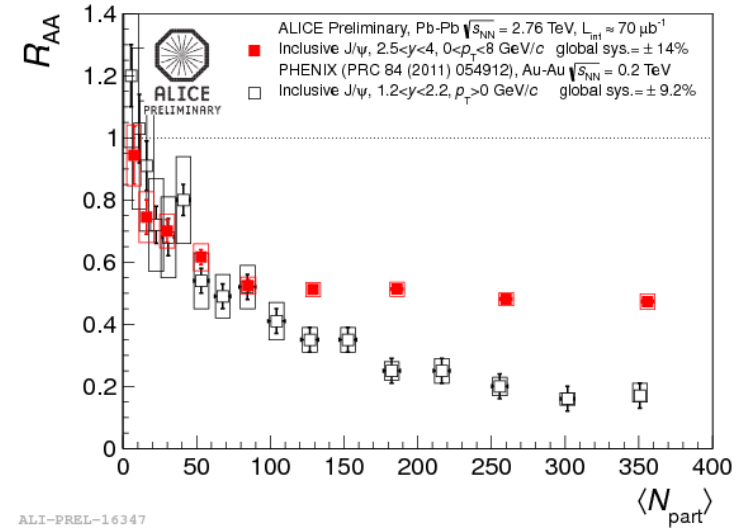
Nature 448, 302-309 (2007)

- Regeneration accounted for in transport model (continuous dissociation/recombination of bound states) as well as statistical hadronization model (statistical binding of pairs at hadronization)



# Motivation

- $B_c$  ( $\bar{b}c$ ) as a potential probe : in pp requires production of a cc pair and a bb pair in the same collision.  
→ in HIC, regeneration would contribute to a huge enhancement of  $B_c$  yields at LHC (as predicted in early 2000's in [PhysRev C.62.024905](#))  
→ Transport and thermal models predict respectively an  $R_{AA}$  of 2.3-17.5 and 3.5-13 in 2.76 TeV Pb+Pb ... ( [PhysRev C.87.014910](#))
- Also possible insight into mass hierarchy of energy loss : smaller suppression for heavy quarks than for light quarks (dead cone effect ?)

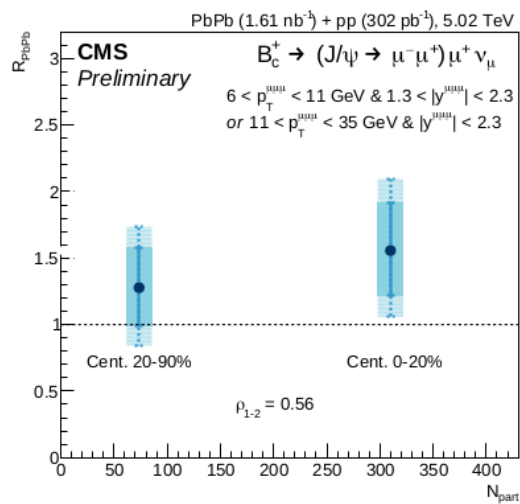
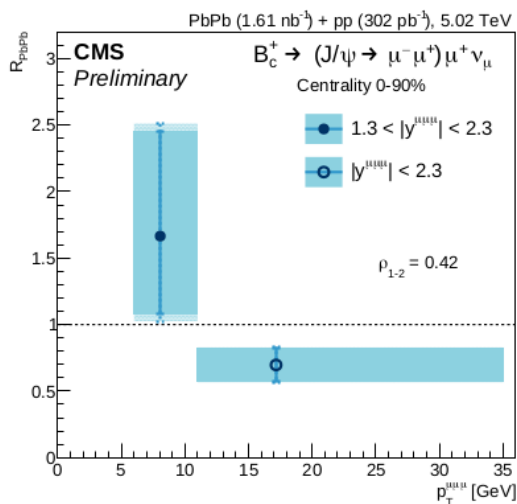


# LHCb & CMS results

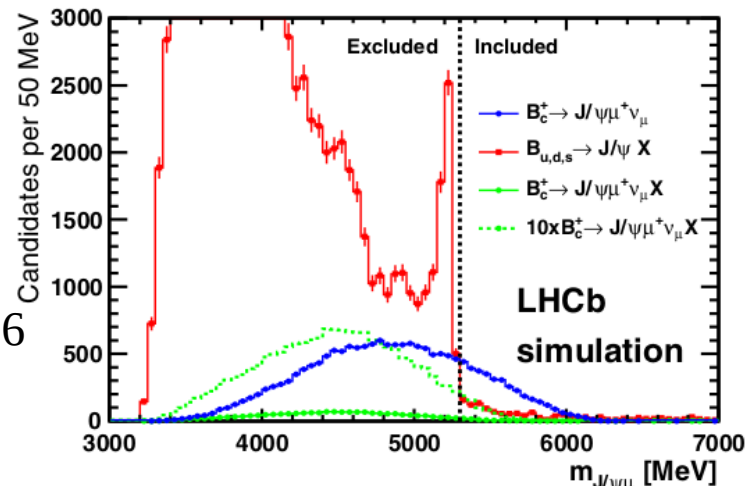
- Measurement of relative branching ratio for  $B_c$  in pp by LHCb (forward rapidities) : correlated background of  $B \rightarrow J/\psi X$  decays eliminated with mass selection

$$R = \frac{B(B_c \rightarrow J/\psi \pi)}{B(B_c \rightarrow J/\psi \mu \nu_\mu)} R(m_{J/\psi \mu} > 5.3 \text{ GeV}) = 0.271 \pm 0.016 \pm 0.016$$

$$R = 0.0469 \pm 0.0028 \pm 0.0046$$



CMS PAS HIN-20-004



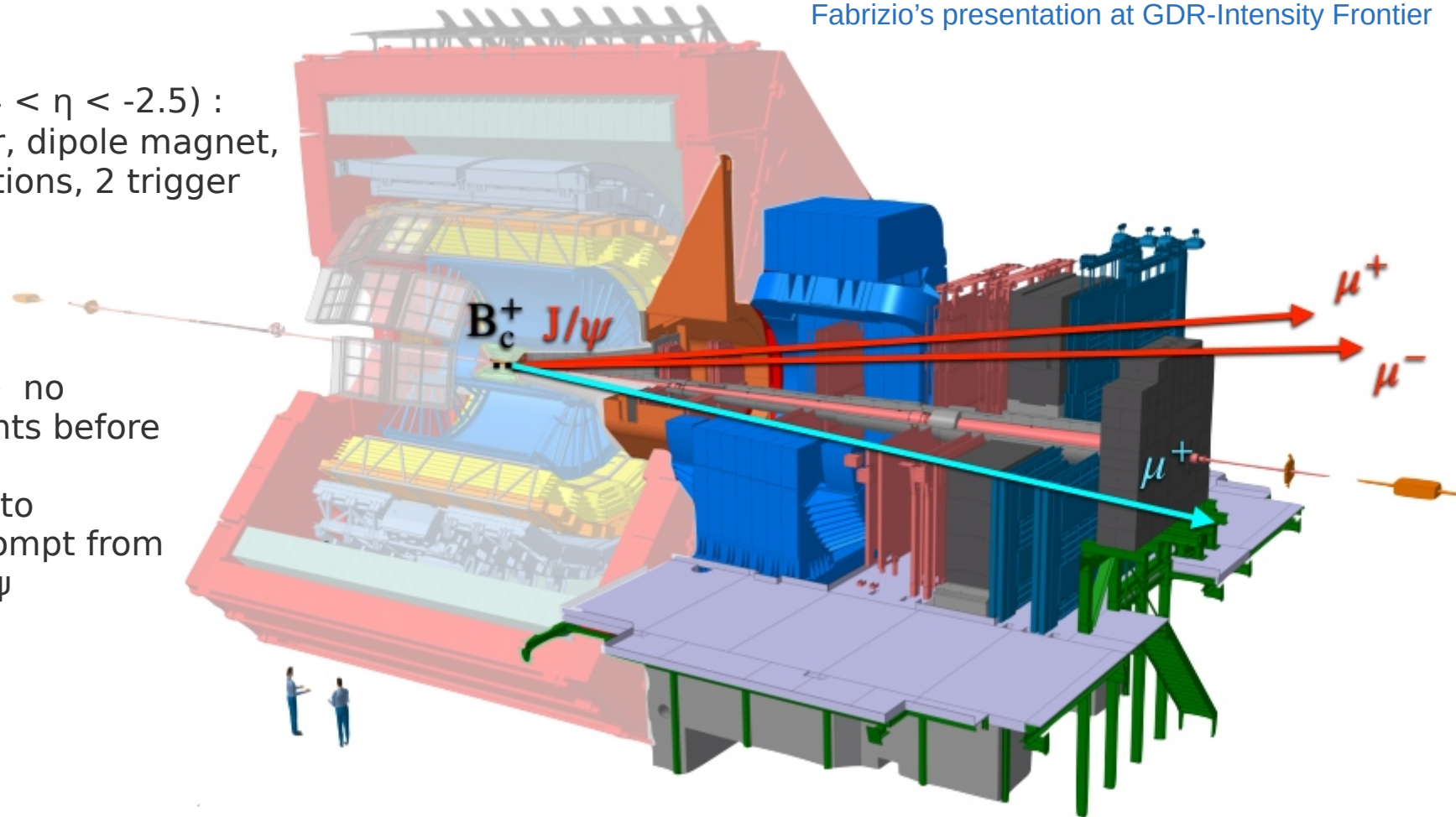
LHCb-PAPER-2014-025

- Recent analysis in pp and PbPb at 5.02 TeV by CMS in two pT bins
  - Signal extraction done with BDT and mass fit.
  - Large uncertainties mainly due to fit uncertainties and poorly known kinematics of  $B_c$  ( $p_T$  spectrum affects acceptance and efficiency corrections)

# Bc in ALICE at forward rapidity

Fabrizio's presentation at GDR-Intensity Frontier

- MUON Arm ( $-4 < \eta < -2.5$ ) :  
Front absorber, dipole magnet, 5 tracking stations, 2 trigger stations
- Run 1 and 2  $\rightarrow$  no measured points before absorber  
 $\rightarrow$  impossible to distinguish prompt from non-prompt  $J/\psi$

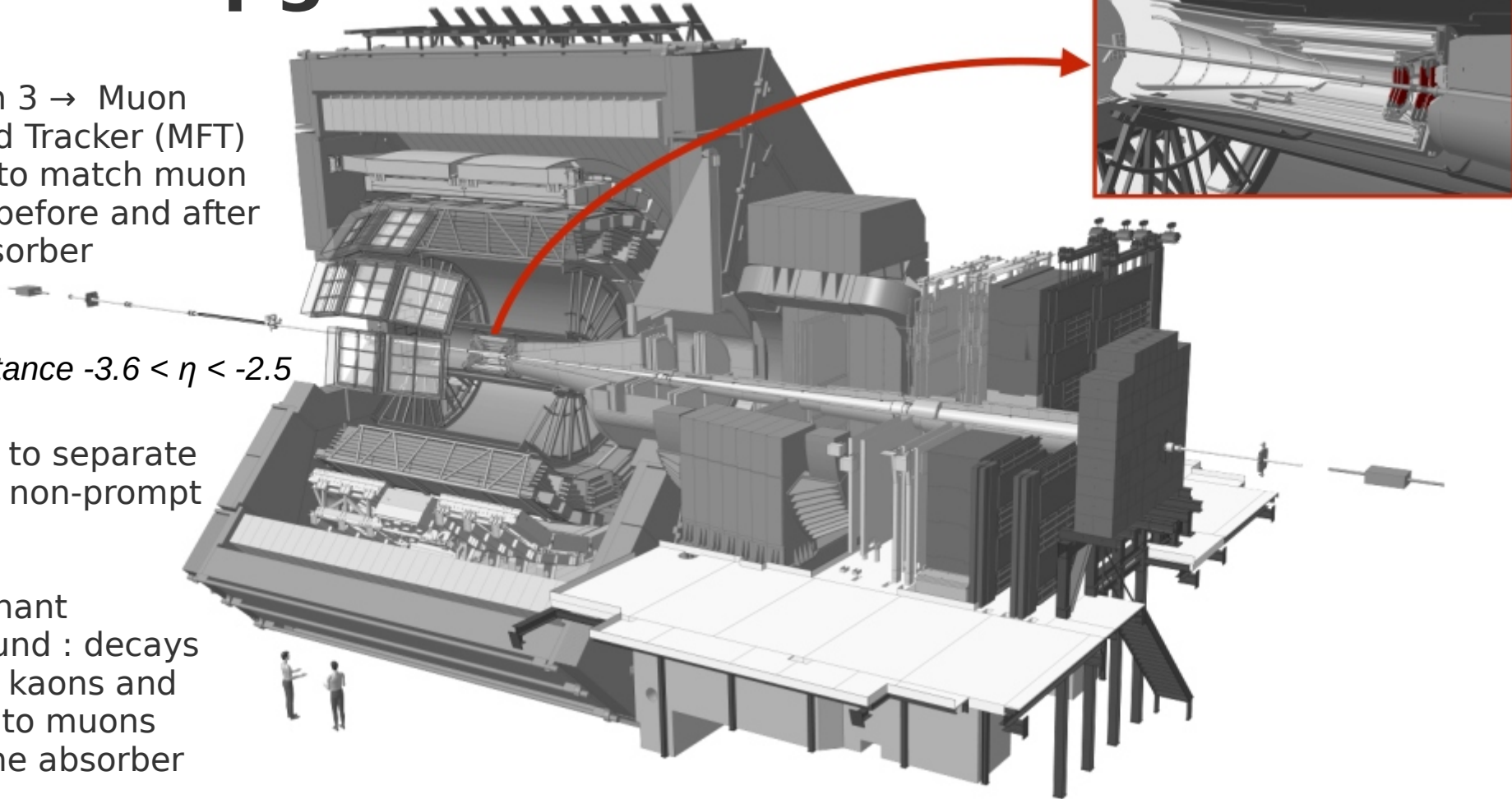


# ALICE Upgrade

- For Run 3 → Muon Forward Tracker (MFT) added to match muon tracks before and after the absorber

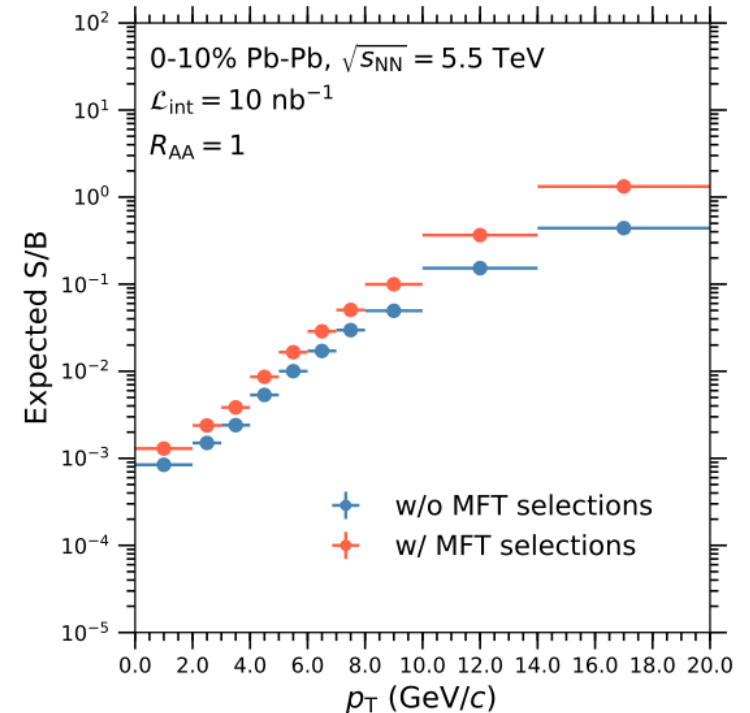
*MFT acceptance  $-3.6 < \eta < -2.5$*

- → allows to separate prompt / non-prompt  $J/\psi$
- → Dominant background : decays of pions, kaons and charm into muons before the absorber



# Study of $B_c \rightarrow \mu \nu_\mu$ ( $J/\psi \rightarrow \mu \mu$ ) channel

- Based on estimation of signal/background for  $B_c$  made by Fabrizio Grosa for channel  $B_c \rightarrow \mu \nu_\mu$  ( $J/\psi \rightarrow \mu \mu$ ) at forward rapidity in ALICE, see his presentation at [GDR-Intensity Frontier](#).
- Fabrizio also looked at the  $B_c \rightarrow D^{*+} D^0$  in central barrel, seems less promising
- One of the difficulties of the 3 muon channel : only partially reconstructed  $\rightarrow$  not peaked signal but broad mass spectrum
- Cross section estimated from LHCb measurements ([arXiv:1411.2943](#), [arXiv:1407.2126](#)) and FONLL
- $\rightarrow$  estimated  $S/B \sim 10^{-3}$  and significance  $\sim 1$  in the lowest  $p_T$  bin, *But does not exploit secondary vertexing and assuming an  $R_{AA}=1$*

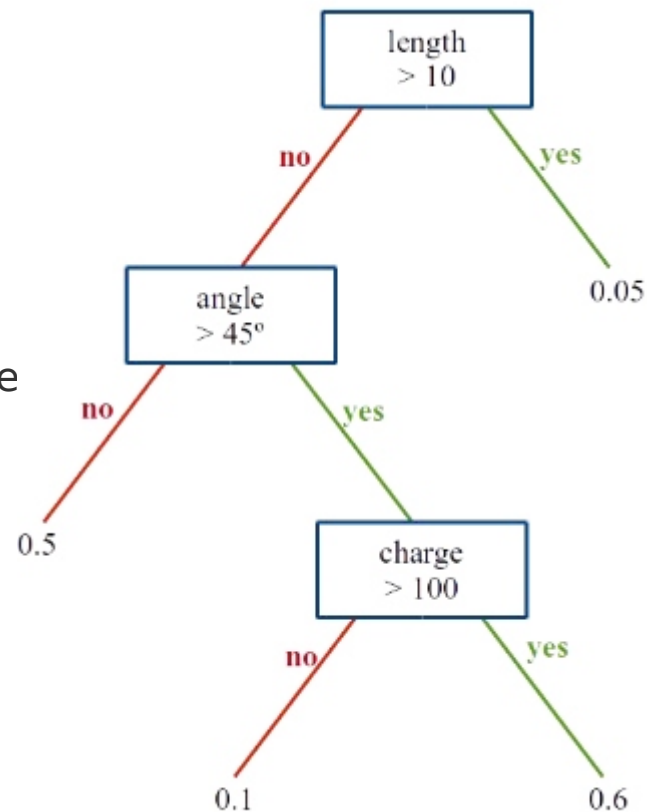


$\rightarrow$  Can this be improved using secondary vertexing and ML ?

# Boosted decision tree with XGBoost

- Used in e.g.  $\Lambda_c^+$  in Pb+Pb
- Uses ensemble learning : many weak learners (classification/regression trees) are combined to produce a strong learner
- → Each tree's output  $f_t(x)$  is given a weight  $w_t$  relative to its accuracy
- → The ensemble output is the weighted sum:
$$\hat{y}(x) = \sum_t w_t f_t(x)$$
- Boosting : each learner in the ensemble is created iteratively

[xgboost.readthedocs.io](http://xgboost.readthedocs.io)  
[arXiv:1603.02754](https://arxiv.org/abs/1603.02754)

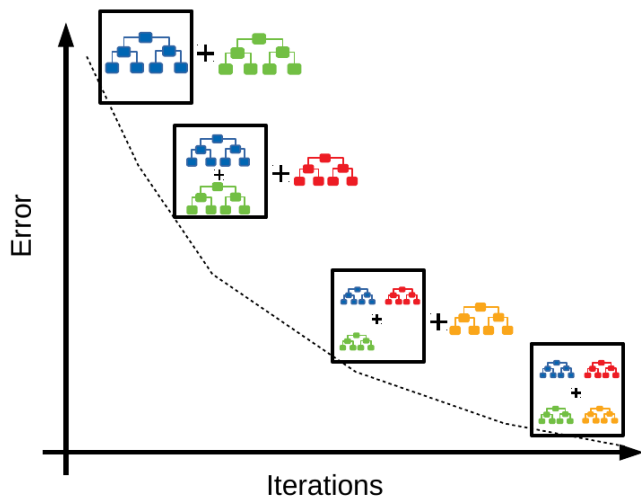


Source : K.Woodruff



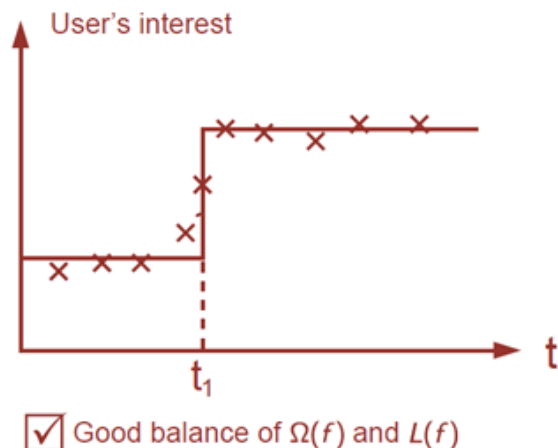
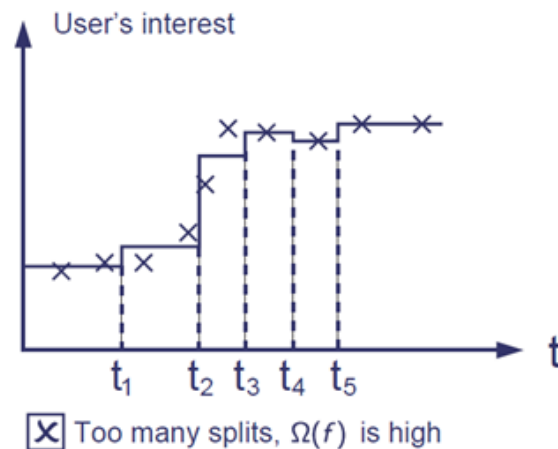
- Gradient boosting : At each iteration, the new tree is constructed to minimize the function :

$$O(x) = \sum_i l(\hat{y}_i, y_i) + \sum_t \Omega(f_t)$$



<https://morioh.com/p/e108a4521555>

(  $l(\hat{y}, y)$  is a loss function, for example the logistic map for binary classification, and  $\Omega(f)$  is a regularisation function to avoid over-fitting)



# Features (input variables)

## Three-prong variables

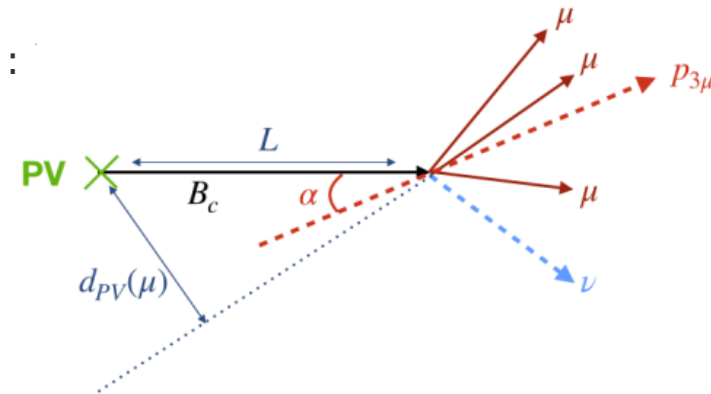
- Longitudinal  $B_c$  pseudo-proper time :  $\tau_{z, B_c} = (L_z M_{B_c})/p_{z, B_c}$ , where  $L_z$  is the longitudinal projection of the  $B_c$  decay length
- Dispersion of tracks at the secondary vertex (vertex quality) :  $\sigma_{\text{vtx}}^2 = \sum_{i=1}^3 d_i^2$ , where  $d_i$  is the DCA between the  $i$ th track and the reconstructed decay vertex
- Cosine of pointing angle  $\cos \alpha$ , where  $\cos \alpha = \vec{L} \cdot \vec{p}_{3\mu} / (|\vec{L}| |p_{3\mu}|)$  with  $\vec{L} = \vec{S\bar{V}} - \vec{P\bar{V}}$

## Two-prong variables

- Longitudinal  $J/\psi$  pseudo-proper time :  $\tau_{z, J/\psi} = (L_z M_{J/\psi})/p_{z, J/\psi}$ , where  $L_z$  is the longitudinal projection of the  $J/\psi$  decay length

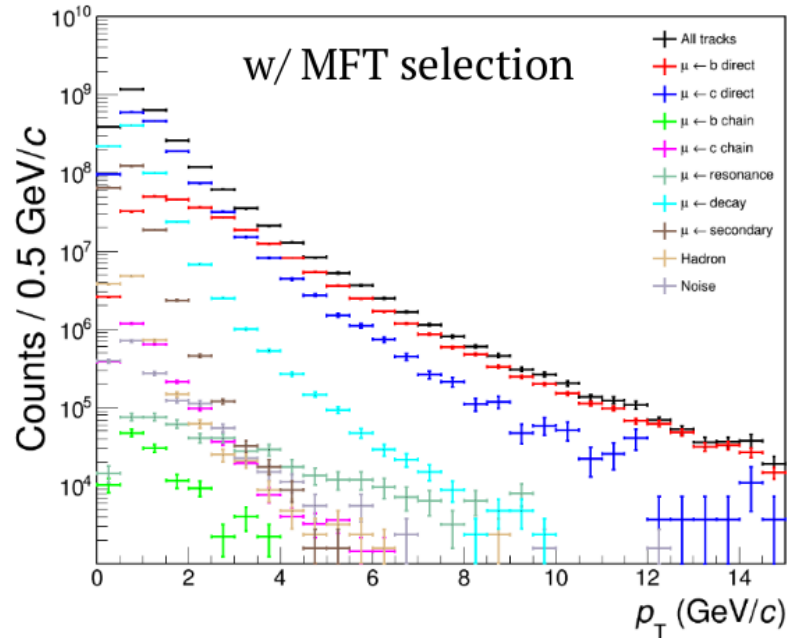
## Single track variables

- Impact parameter (distance of closest approach) in the transverse plane of the three muons to the primary vertex



G.Falmagne, GDR-Intensity Frontier

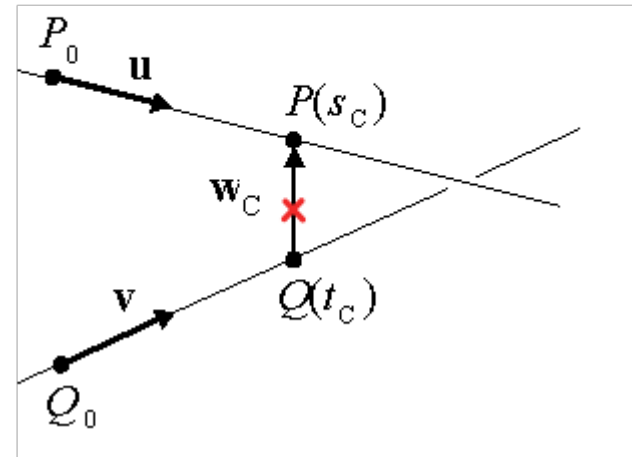
# Fast simulation



→ Considered only the uncorrelated background from decay of K,  $\pi$  and charm before the absorber (at 46 cm)

→ Checked that we could reproduce J/psi background

- Done with fast simulation software RapidSim, using FONLL cross sections and designed for LHCb  
→ Modified geometry (distance to primary vertex, magnet)  
→ Modified resolutions (momentum, DCA, vertex)



→ Simplified secondary vertexing : finding the vector of minimal length between two tracks, taking its midpoint & smearing this point with fixed resolution (no  $p_T$  dependence)

# Background composition

Three contributions to the uncorrelated background considered :

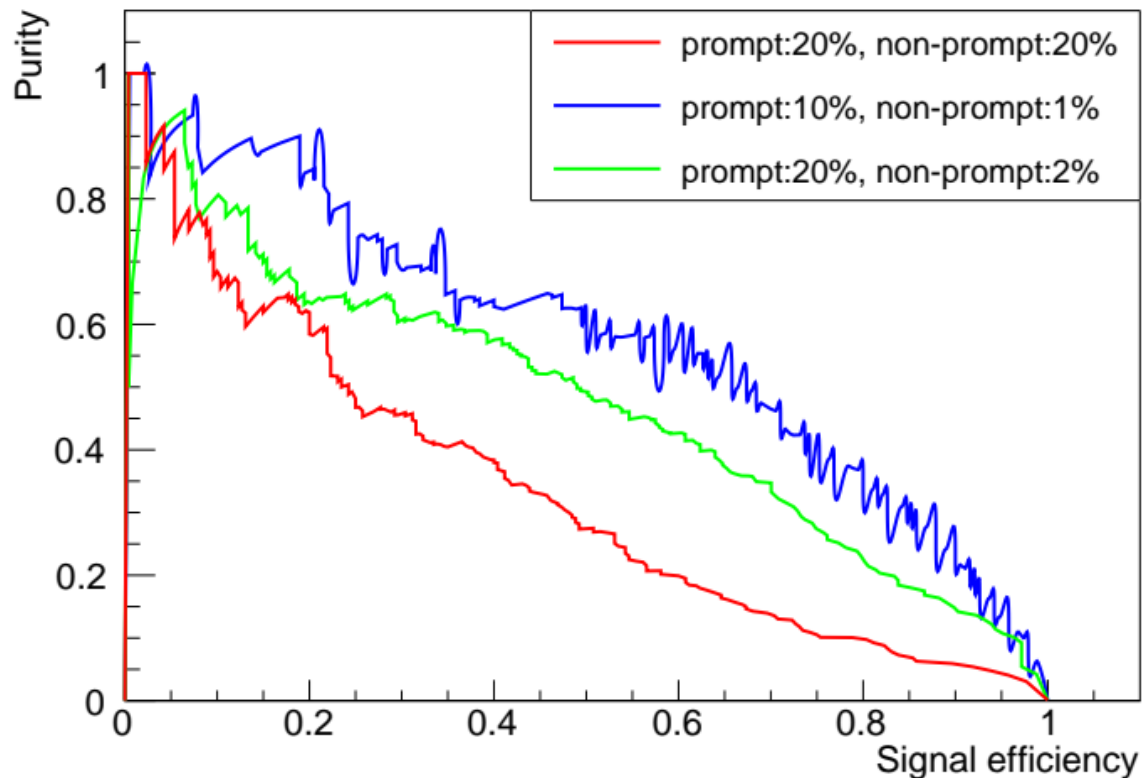
- **'Fake'  $J/\psi$  + muon** : Three background muons from the decay of  $\pi$ , K or charm in the same event (two of them represent a 'Fake'  $J/\psi$ , i.e. background under the  $J/\psi$  peak)
- **Prompt  $J/\psi$  + muon** : Mixing a prompt  $J/\psi$  with a background muon in the same event
- **Non-prompt  $J/\psi$  + muon** : Mixing a non-prompt  $J/\psi$  with a background muon in the same event
  - Varying the composition of the background by varying the percentage of these three types in the total background, while keeping a Signal/Background  $\sim 10^{-3}$  (Reproducing Fabrizio's estimate) with a total sample of  $\sim 1\text{M}$  candidates
  - Monitor the impact on signal efficiency and purity at the output of the BDT
  - All  $p_T$  and mass integrated

# BDT training and testing

- Varying background composition in sample
- 80% of sample for training, 20% for testing
- Learning rate = 0.1
- Maximum depth = 3
- ratio of features used = 0.3
- ratio of the training instances used = 0.3
- Iterate over 400 trees

→ Before selection : total signal

→ After selection : true signal (true positive) & false signal (false positive)



$$\text{Purity} = \text{true signal} / (\text{true signal} + \text{false signal})$$

$$\text{Efficiency} = \text{true signal} / \text{total signal}$$

# Conclusion

- Simplified modelling of background and secondary vertexing with fast simulation, not taking into account correlated background, feed-down pT dependence of resolution
- Significant improvement of S/B with gradient boosting :  
e.g. for signal efficiency of 0.4 (starting with  $S/B \sim 1e-3$ ,  $\text{sig} \sim 1$ )  
 **$S/B \sim 0.5$ ,  $\text{sig} \sim 11$**  (20% prompt, 20% non-prompt),  
 **$S/B \sim 1.7$ ,  $\text{sig} \sim 16$**  (10% prompt, 1% non-prompt),  
 **$S/B \sim 1.4$ ,  $\text{sig} \sim 15$**  (20% prompt, 2% non-prompt)  
without any mass fitting (compared to  $\text{sig} \sim 4$  in CMS analysis)
- Seems too good to be true ... Need full simulation !

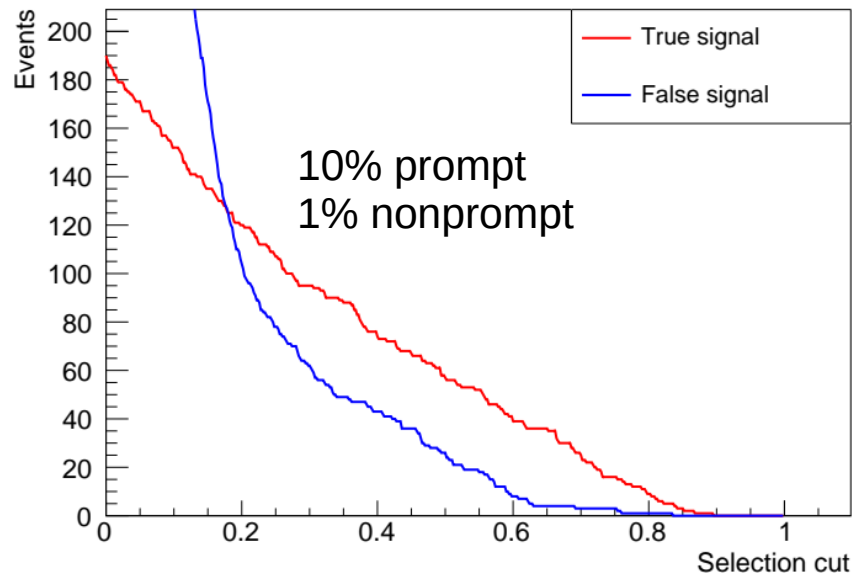
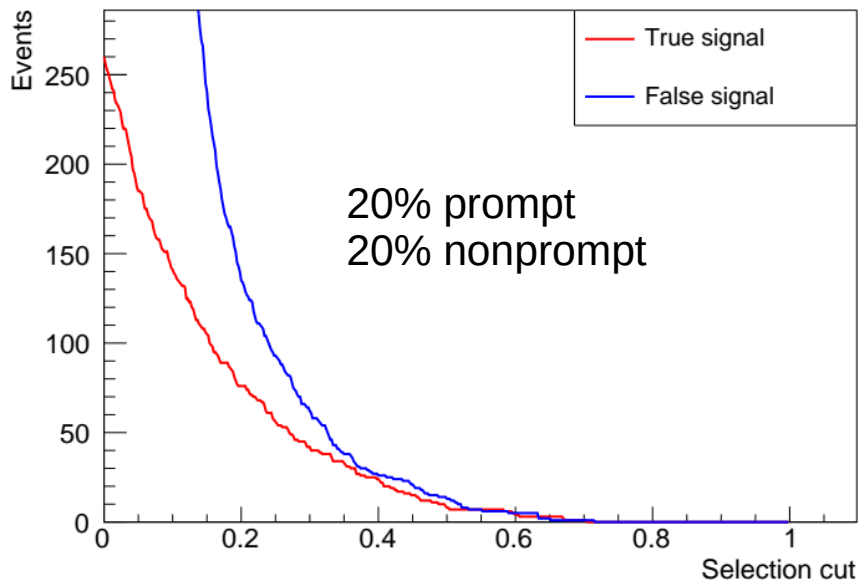


**Thank you !**

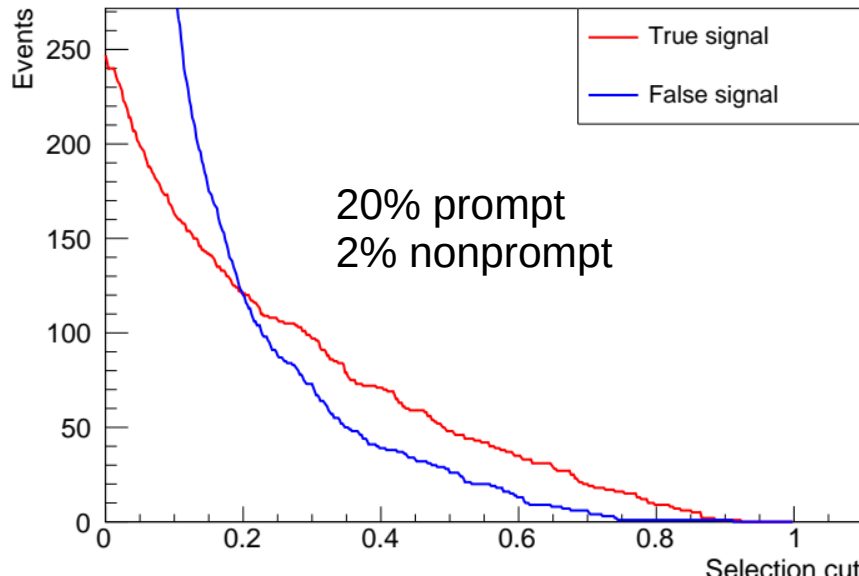


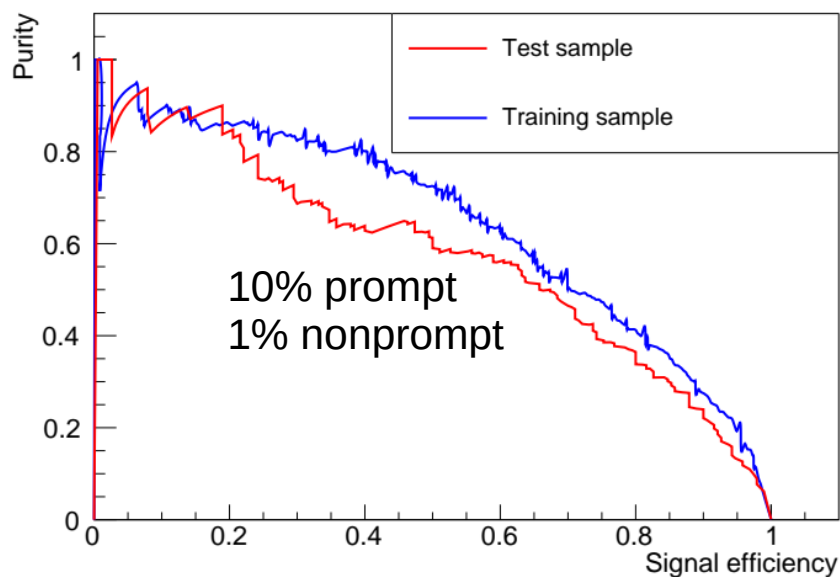
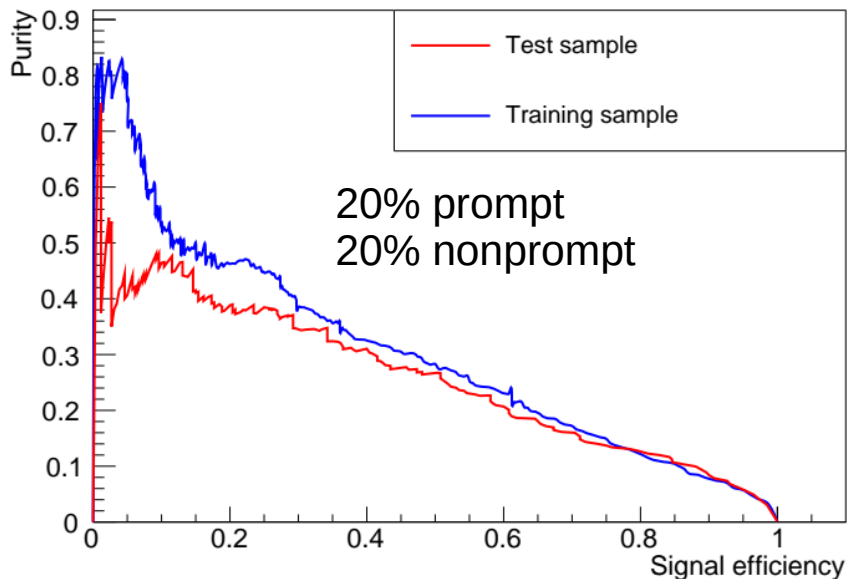
# Backup



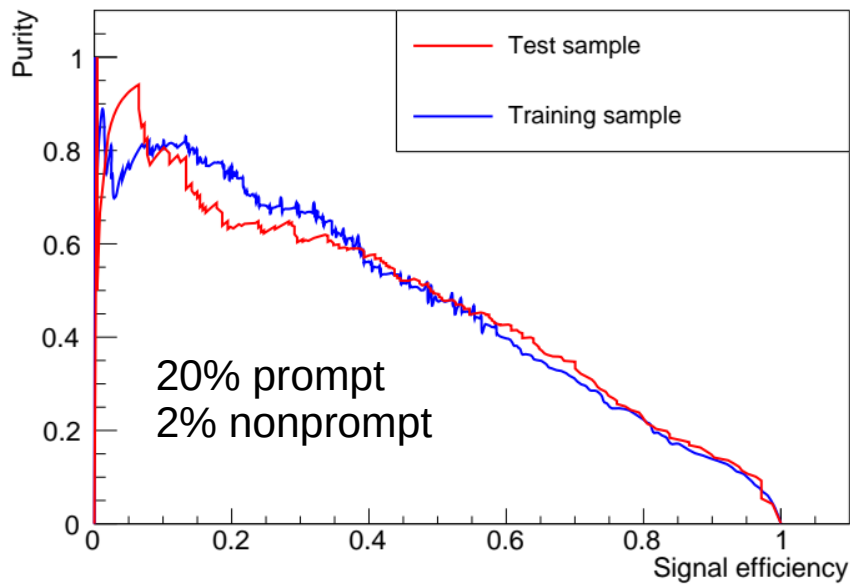


- Compare True signal and False signal in the selected sample as a function of the selection cut parameter (i.e. the output classifying parameter of the BDT)





- Compare signal efficiency vs purity results from test sample and training sample  
→ Estimates the overfitting



# Motivation

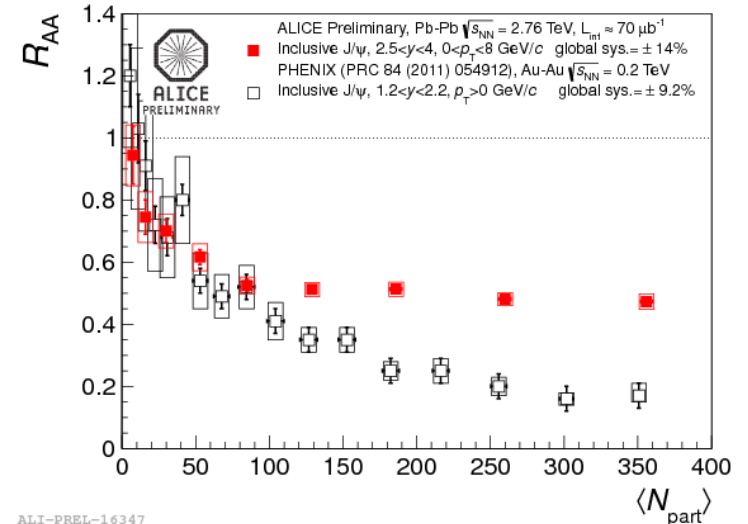
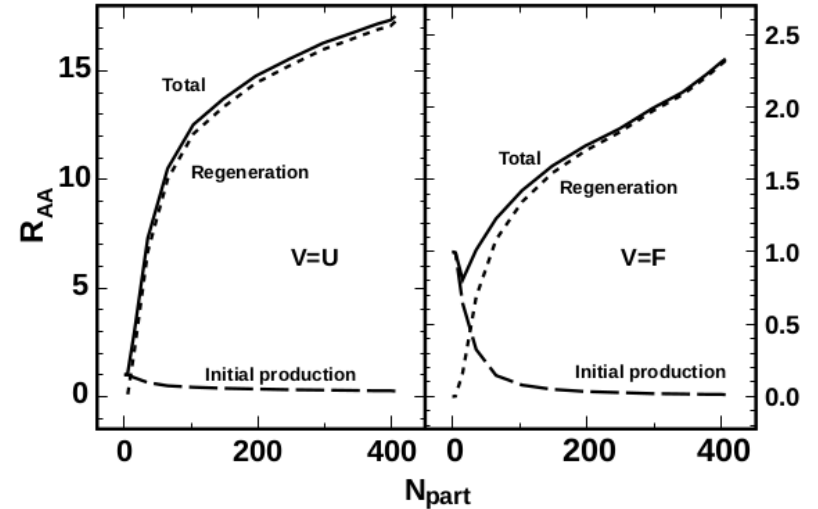
- $B_c (\bar{b} c)$  as a potential probe : in pp requires production of a cc pair and a bb pair in the same collision.

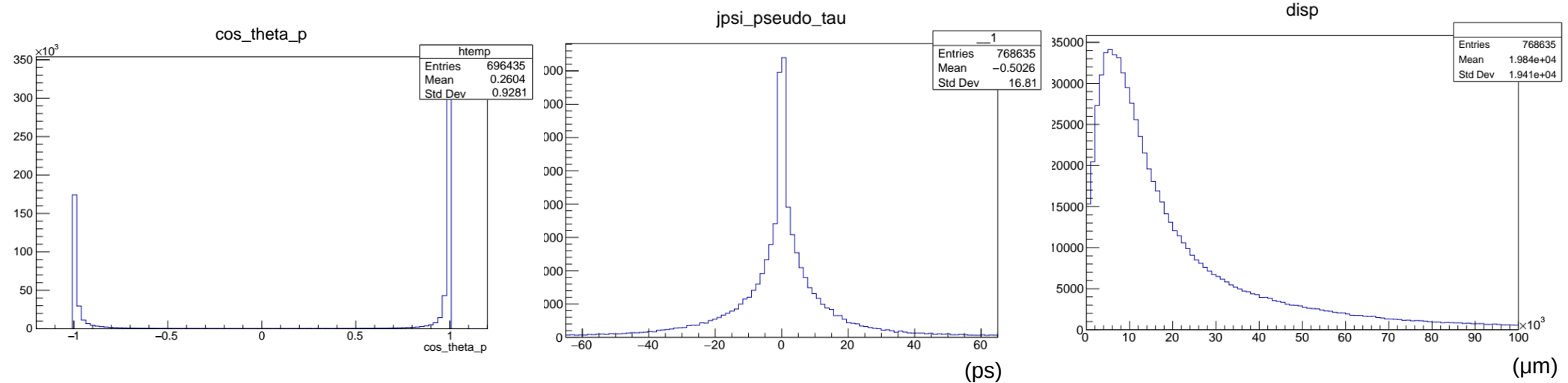
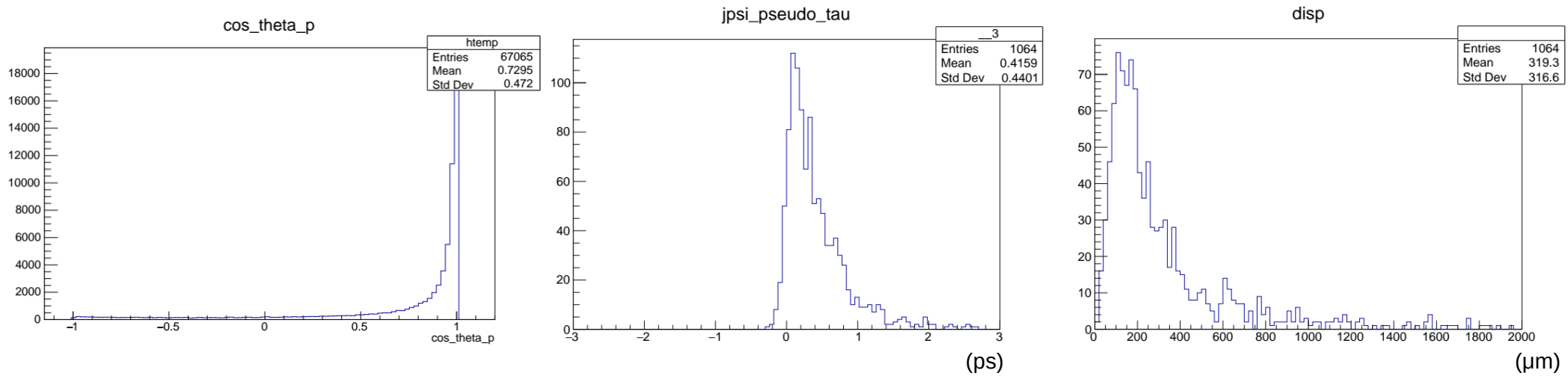
→ in HIC, regeneration would contribute to a huge enhancement of  $B_c$  yields at LHC (as predicted in early 2000's in [PhysRev C.62.024905](#))

→ Transport and thermal models predict respectively an  $R_{AA}$  of 2.3-17.5 and 3.5-13 in 2.76 TeV Pb+Pb ... (arXiv:1207.2366)

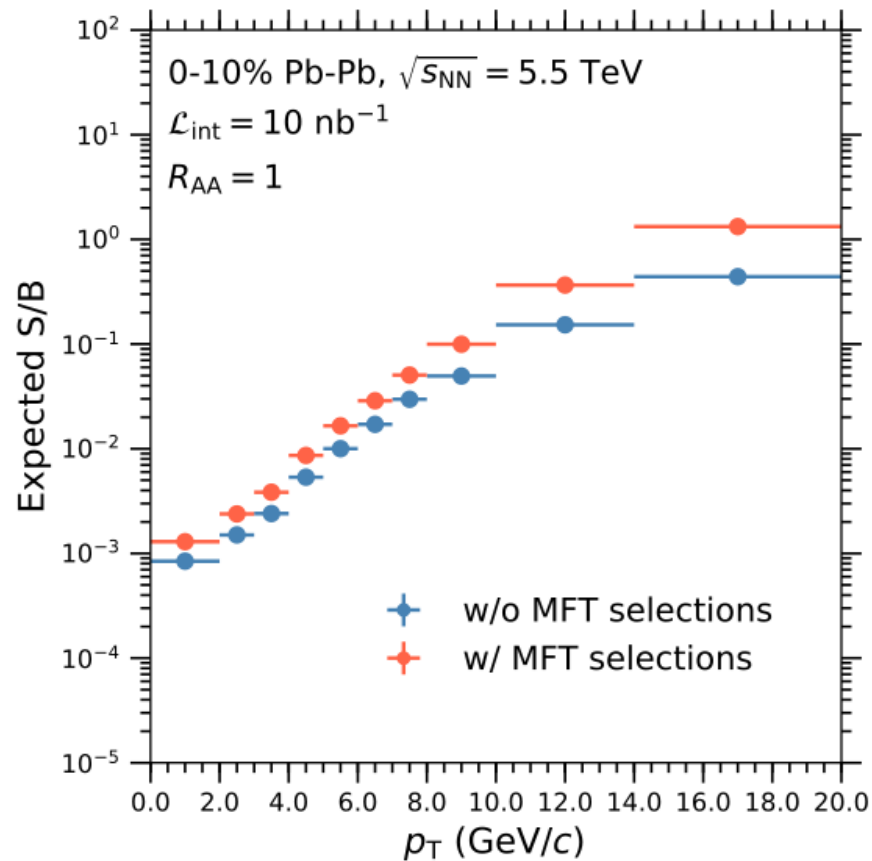
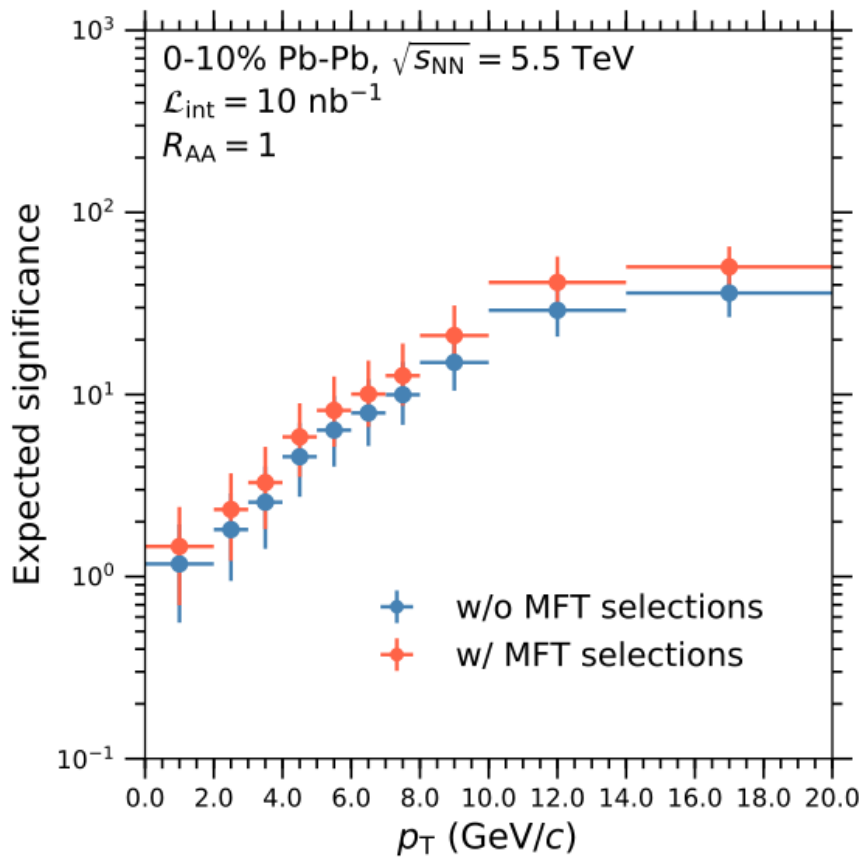
- Also possible insight into mass hierarchy of energy loss : smaller suppression for heavy quarks than for light quarks (dead cone effect ?)

$$(R_{AA}(b) > R_{AA}(c) > R_{AA}(u,d,s))$$

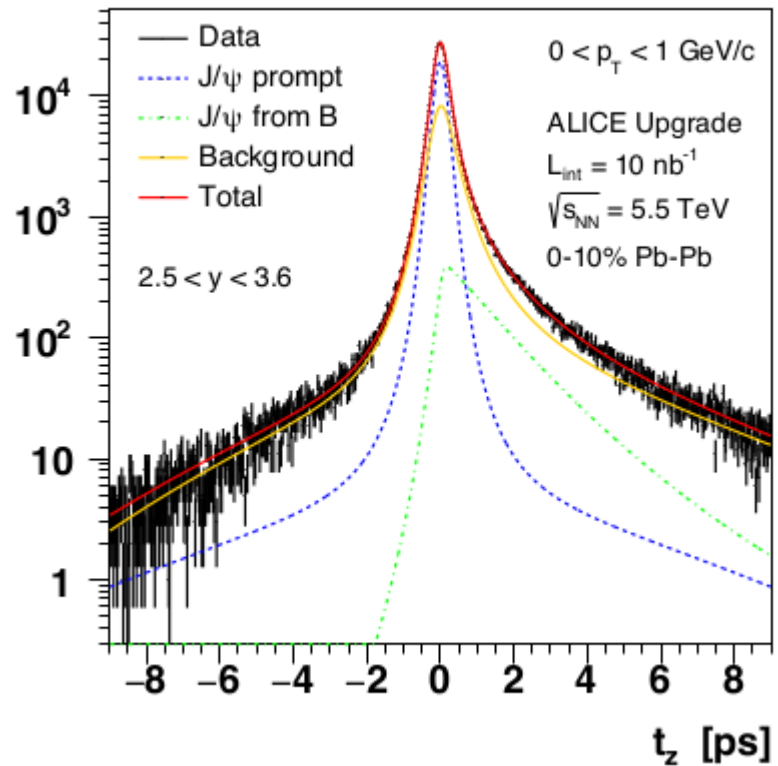
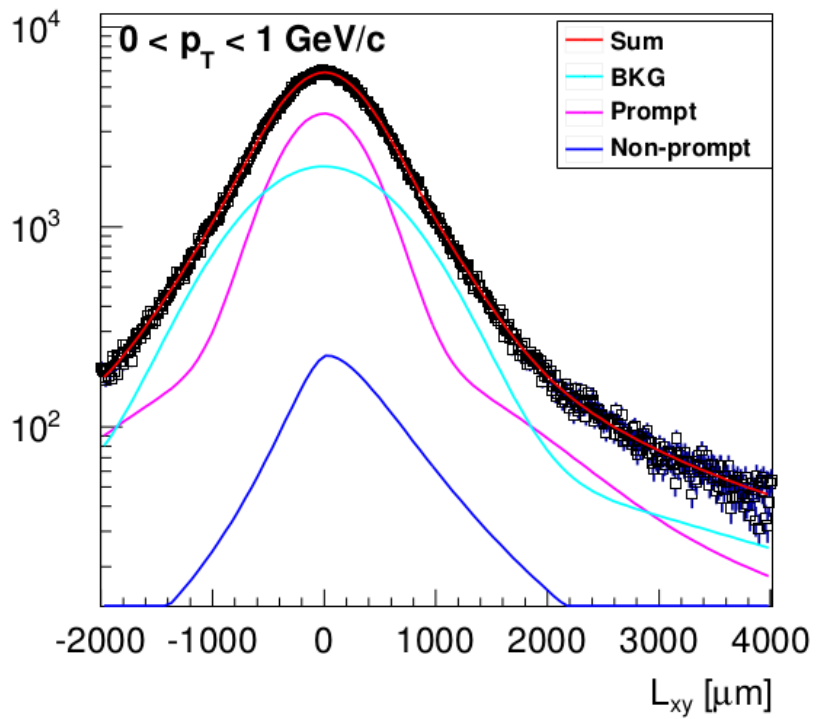


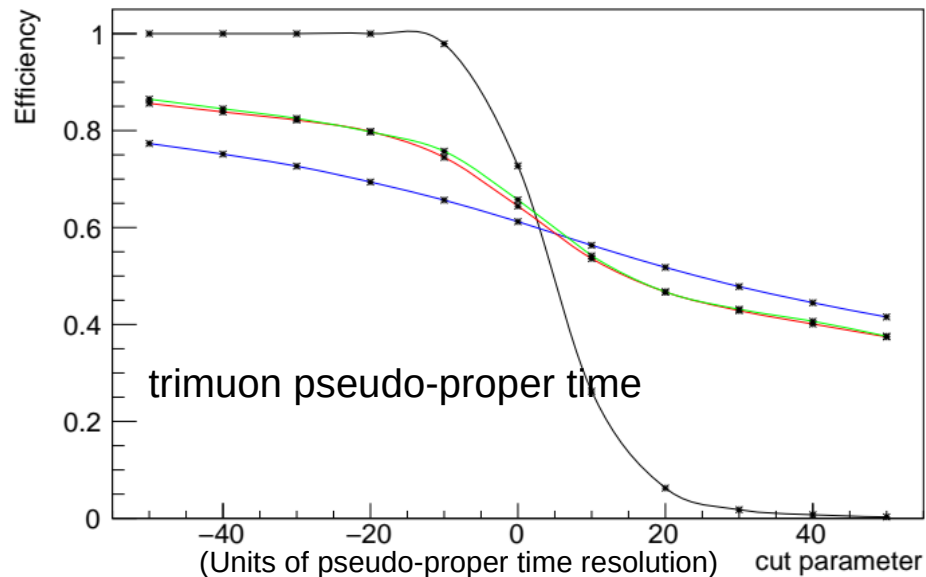
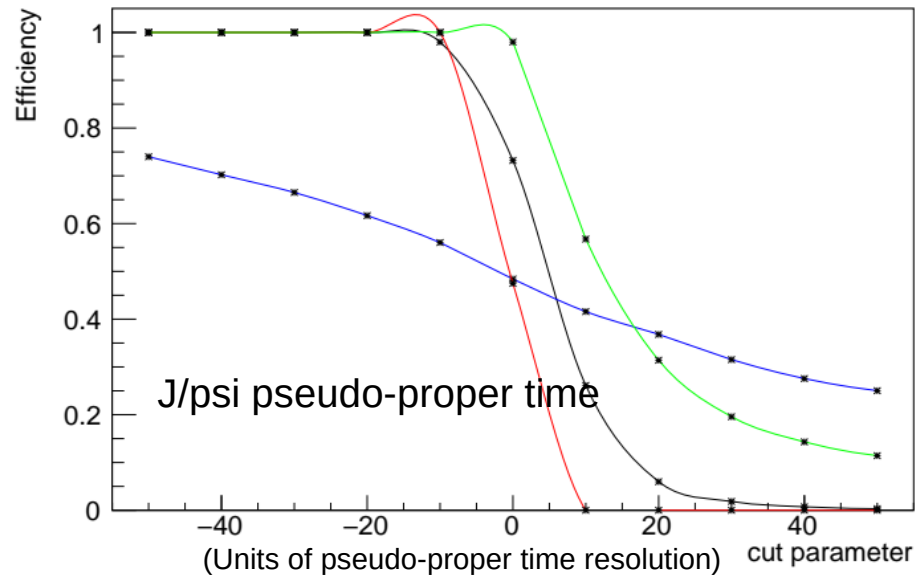
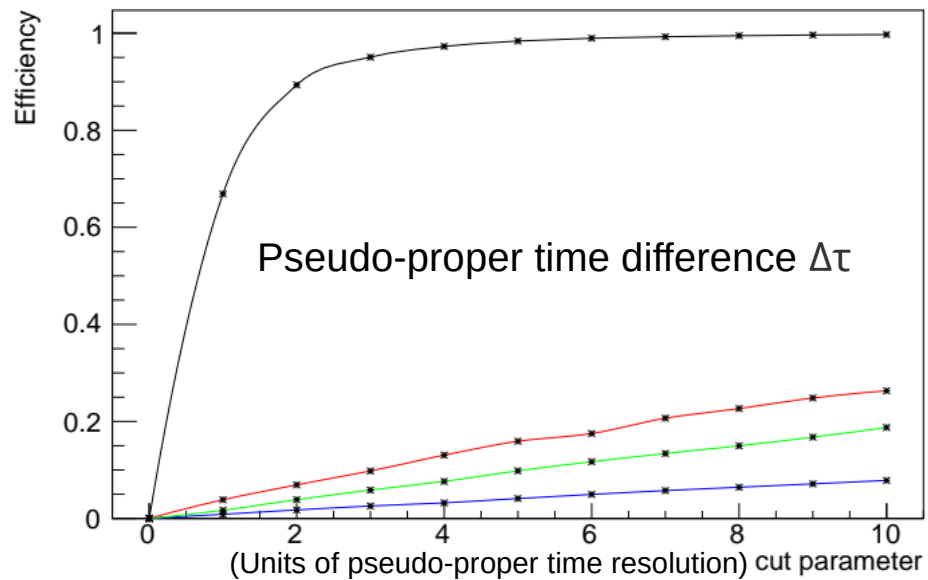
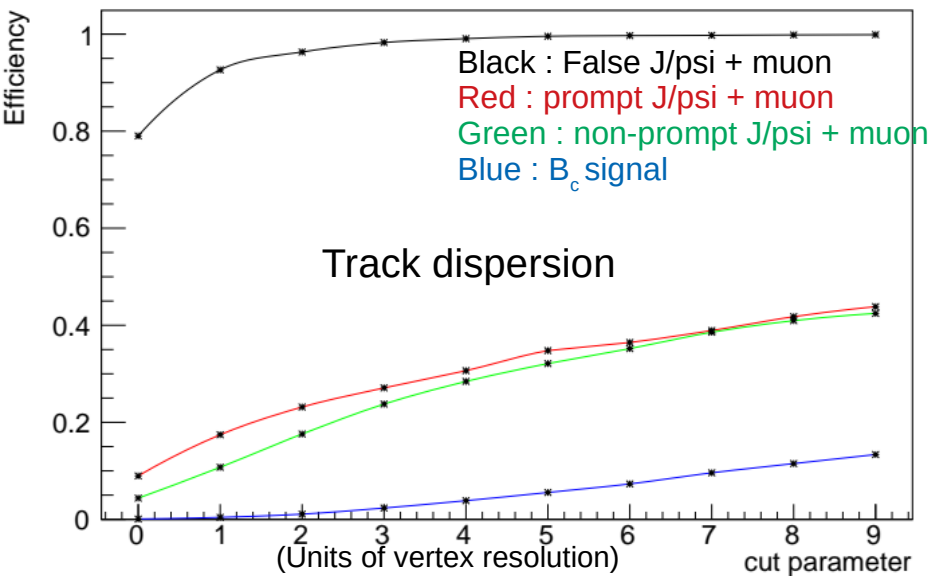


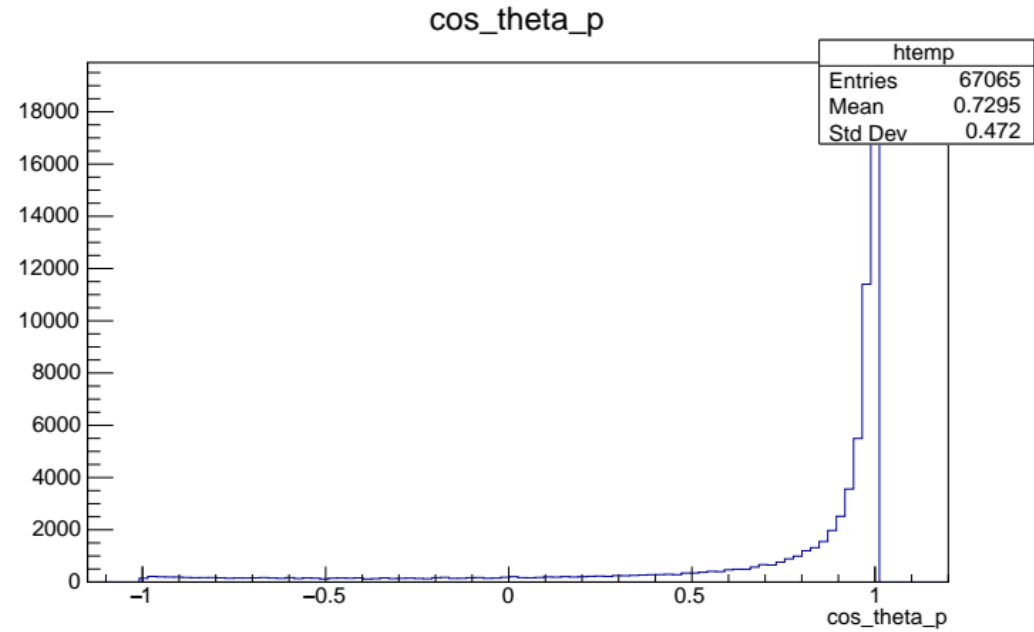
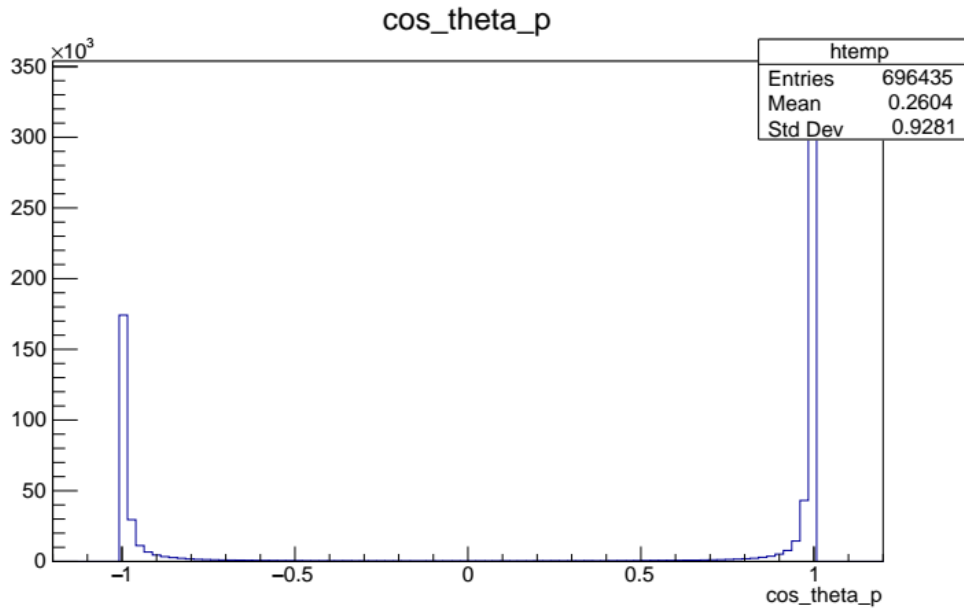
(Top : signal distributions for  $\cos\theta_p$ ,  $J/\psi$  candidates pseudo-proper times and track dispersion at secondary vertex)  
 (Bottom : same distributions for background candidates (10% prompt, 1% non-prompt))



From [Fabrizio's presentation](#)







Background (left) and signal (right) distributions for the  $\cos\theta_p$  input variable