

b-jet tagging performance with ALICE at the LHC

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on behalf of the ALICE Collaboration**

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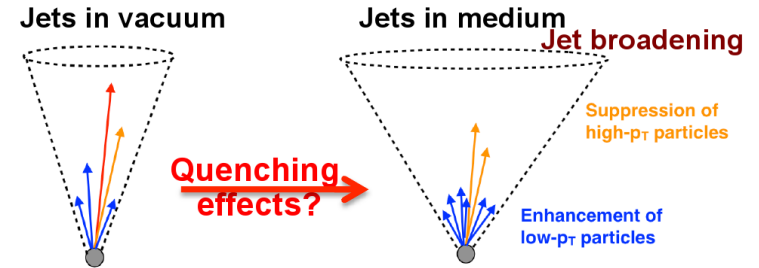
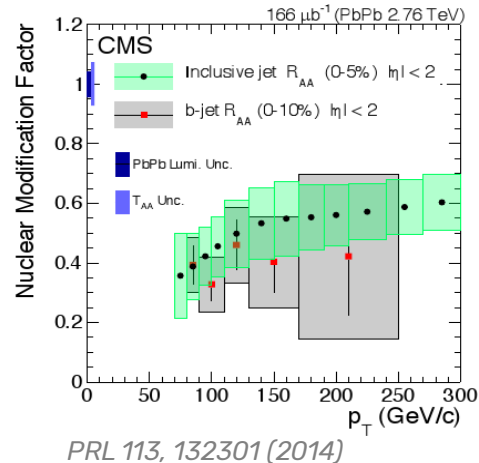
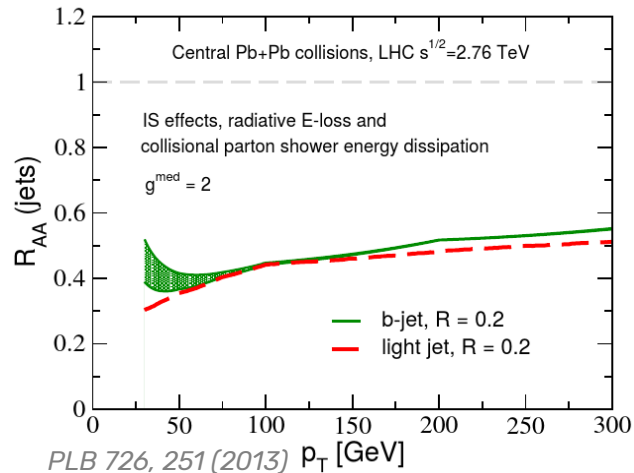


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b-jets: Motivation

- *pp collisions*: sensitive probes of pQCD
- *pA collisions*: initial-state effects
- *AA collisions*: energy loss of hard-scattered partons via collisional and radiative processes
 - Flavour dependence of the jet quenching
 - Spatial redistribution of the lost energy

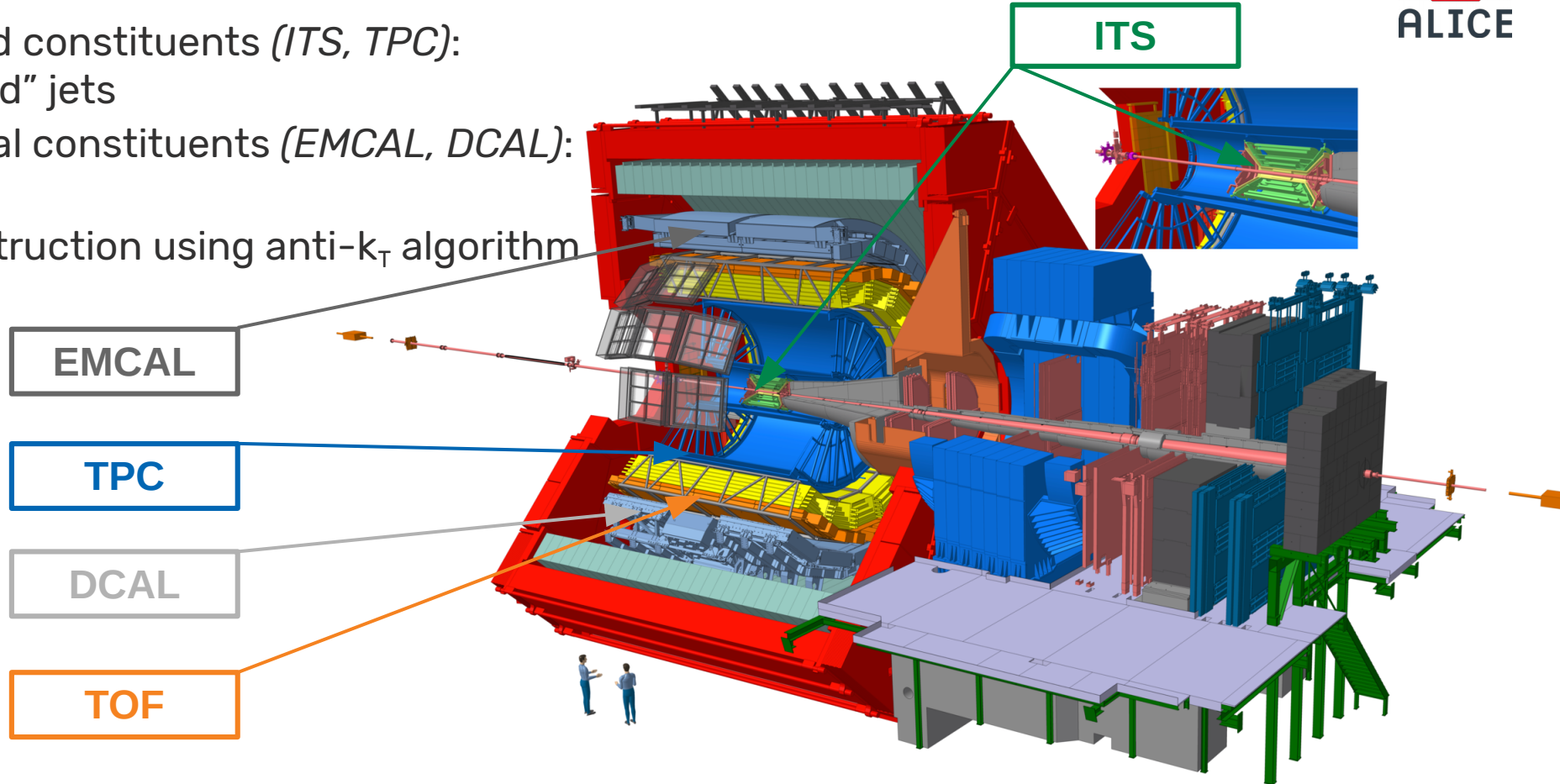
→ b-jets: probes of the QGP transport properties



- Mass effects relevant $p_T < 70 \text{ GeV}/c$
- Lower jet p_T reach with ALICE

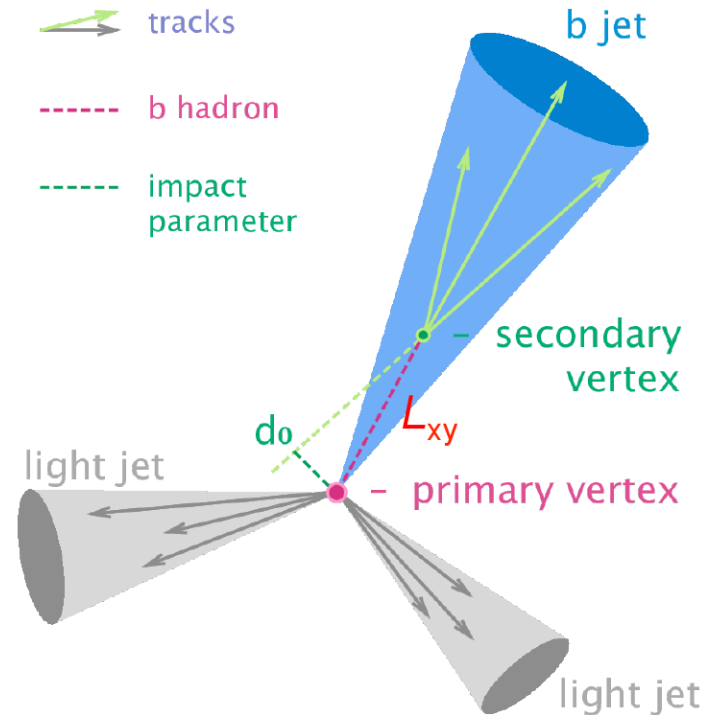
b-jets with ALICE

- Charged constituents (*ITS, TPC*):
“charged” jets
- + neutral constituents (*EMCAL, DCAL*):
full jets
- Reconstruction using anti- k_T algorithm



b-jets with ALICE

- Charged constituents (*ITS, TPC*):
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- Reconstruction using anti- k_T algorithm
 - b-tagging exploiting:
 - B-hadron long lifetime, $c\tau \sim 500 \mu\text{m}$
 - Displaced from primary vertex
 - Its large mass
 - Studied b-tagging algorithms:
 - Secondary vertex: using displaced vertices
 - Track counting: based on single tracks
 - Via heavy-flavour electron identification (charm+beauty)

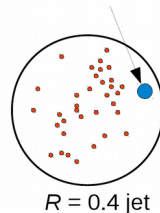


http://bartosik.pp.ua/hep_sketches/btagging

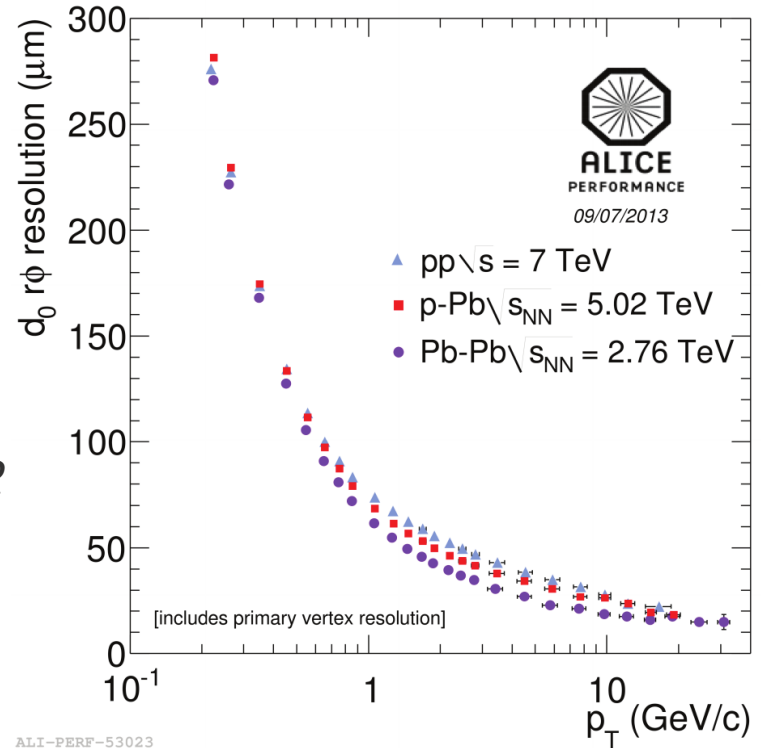
I. Secondary Vertex algorithm

- Exploiting long lifetime and large vertex mass
- TPC and ITS used for tracking and secondary vertex reconstruction
 - Track impact parameter d_0 resolution $< 75 \mu\text{m}$ for $p_T > 1 \text{ GeV}/c$
 - Secondary vertex resolution $\sim 120 \mu\text{m}$
- b-jet if B-hadron within given R
- Otherwise c-jet if charm hadron within given R
- Otherwise light-flavour jet

Heavy-flavour hadron found
in range \rightarrow Tag as HF-jet



$R = 0.4$ jet



SV algorithm: Simulations



- p-Pb at $\sqrt{s_{NN}} = 5.02$ TeV, Pythia 6 + Hijing
- FastJet anti- k_T , $R = 0.4$, $p_{T,track} > 150$ MeV/c
- 3-prong vertices: vertices reconstructed from 3-track combinations, $p_{T,prong} > 1$ GeV/c

→ Discriminators:

- Significance of signed secondary vertex flight distance $L_{xy}/\sigma_{L_{xy}}$:

$$L_{xy} = |\vec{L}'| \text{sign}(\vec{L}' \cdot \vec{p}_{jet})$$

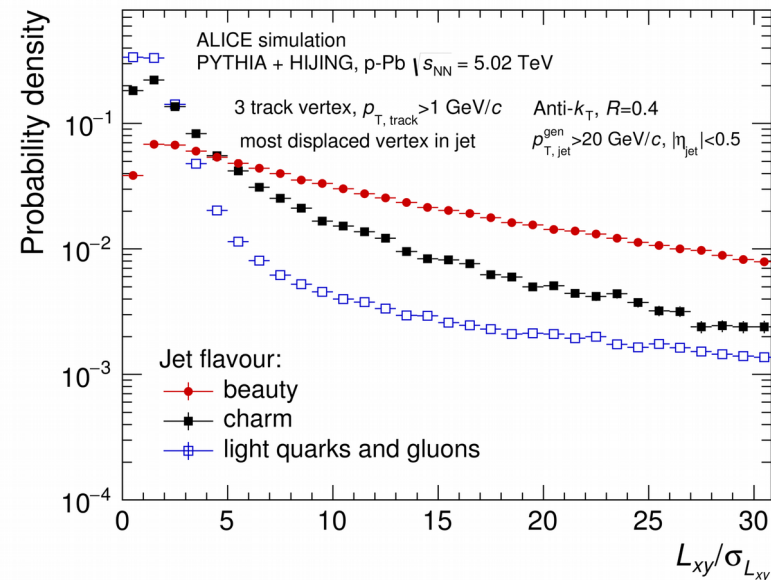
\vec{L}' - vector between primary and secondary vertices

$\sigma_{L_{xy}}$ - uncertainty corresponding to L_{xy}

- SV dispersion (vertex quality measure):

$$\sigma_{vtx} = \sqrt{d_1^2 + d_2^2 + d_3^2}$$

$d_{1,2,3}$ - distances of the tracks from secondary vertex

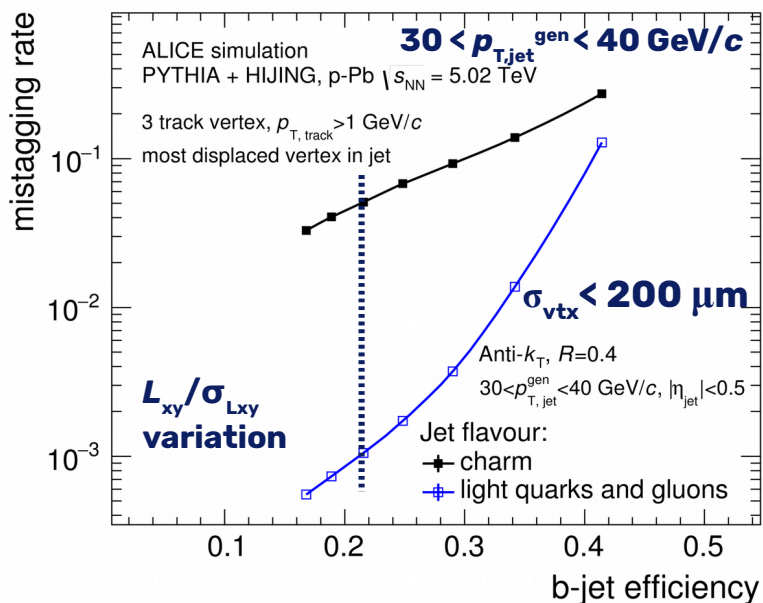


Signed flight distance significance of the most displaced secondary vertex

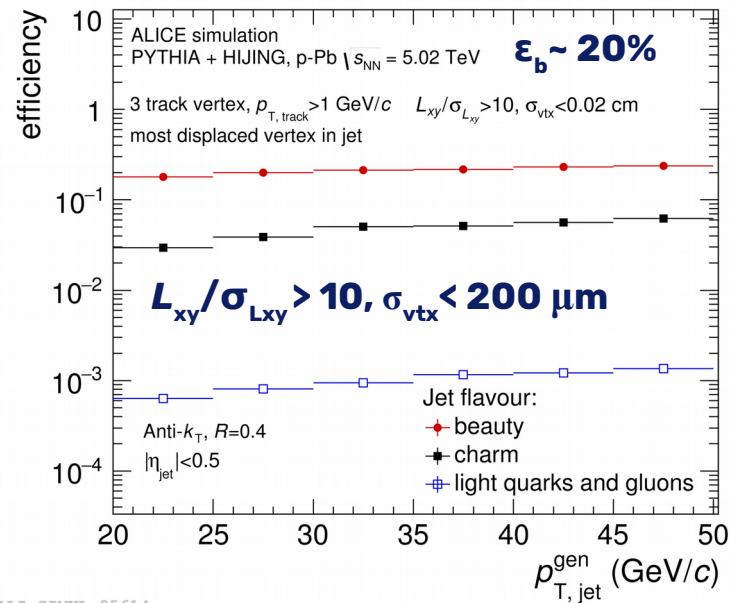
SV algorithm: Performance

- “Rectangular” cuts on the vertex properties
 - b-jet tagging efficiency and c-/udsg-jet misidentification
 - The higher b-jet efficiency, the higher the c-/udsg-jet mistagging efficiency
- ➔ Find condition of high purity and reasonably high efficiency

$$\epsilon_b(p_T) = \frac{dN_b^{\text{tagged}}/dp_T}{dN_b/dp_T}$$



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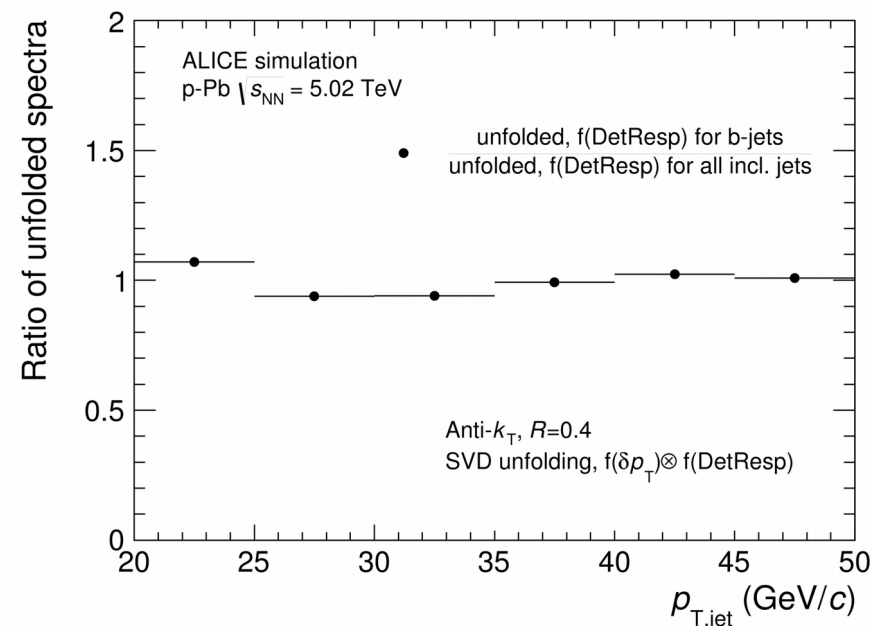
SV algorithm: Unfolding

- SVD (Singular Value Decomposition) [1] used for unfolding
- Background subtraction: background density calculated using CMS method [2], soft clusters found using FastJet k_T with $R = 0.4$

background density:
$$\rho_{\text{CMS}} = \text{median} \left\{ \frac{p_{T,i}}{A_i} \right\} \cdot C$$

C : correction factor for empty clusters

- Unfolding with combined detector and background fluctuations matrix
 - Background fluctuations using Random Cone method in MC:
$$\delta p_T = \sum_i p_{T,i} - \rho A_{\text{cone}}$$
- Correction stability tests performed
 - E.g.: applying efficiency and purity corrections before and after the unfolding



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[1] H. Hoecker, V.Kartverlishvili, Nucl. Instrum. Meth. A372 (1996) 469

[2] CMS collaboration, JHEP 1208 (2012) 130

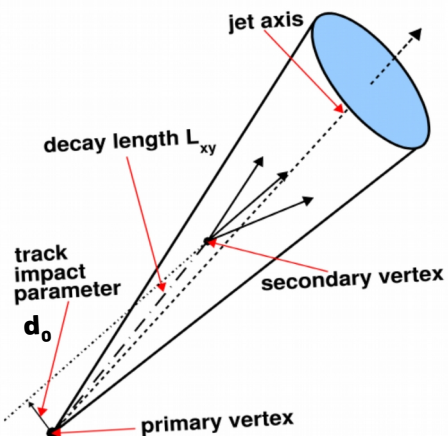
II. Track Counting algorithm

- Impact parameter d_0 in $r\phi$ for each track within a jet

$$\text{sign}(d_0^{\text{jet}}) = \text{sign}(\vec{d}_0 \cdot \vec{p}_{T,\text{jet}})$$

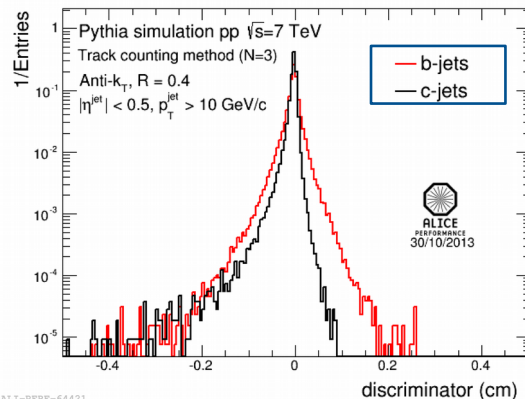
→ Discriminator:

third, second or first (N=3,2,1)
most displaced $\text{sign}(d_0^{\text{jet}})$

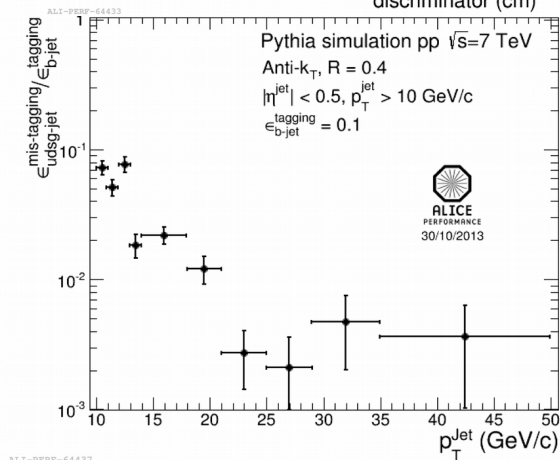
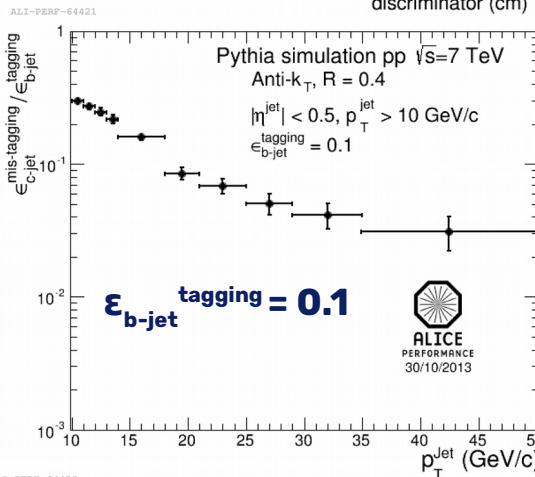
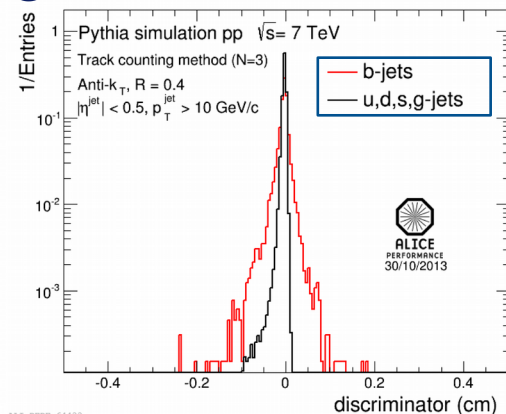


- Pythia simulation: pp at $\sqrt{s} = 7$ TeV

- FastJet anti- k_T , $R=0.4$
- $p_{T,\text{track}} > 150$ MeV/c
- $p_{T,\text{jet}} > 10$ GeV/c
- d_0 in $r\phi > 0.2$ cm
- $\text{DCA}(\text{jet}, \text{track}) < 700$ μm
- $\Delta R(\text{jet}, \text{track}) < 0.3$



N=3



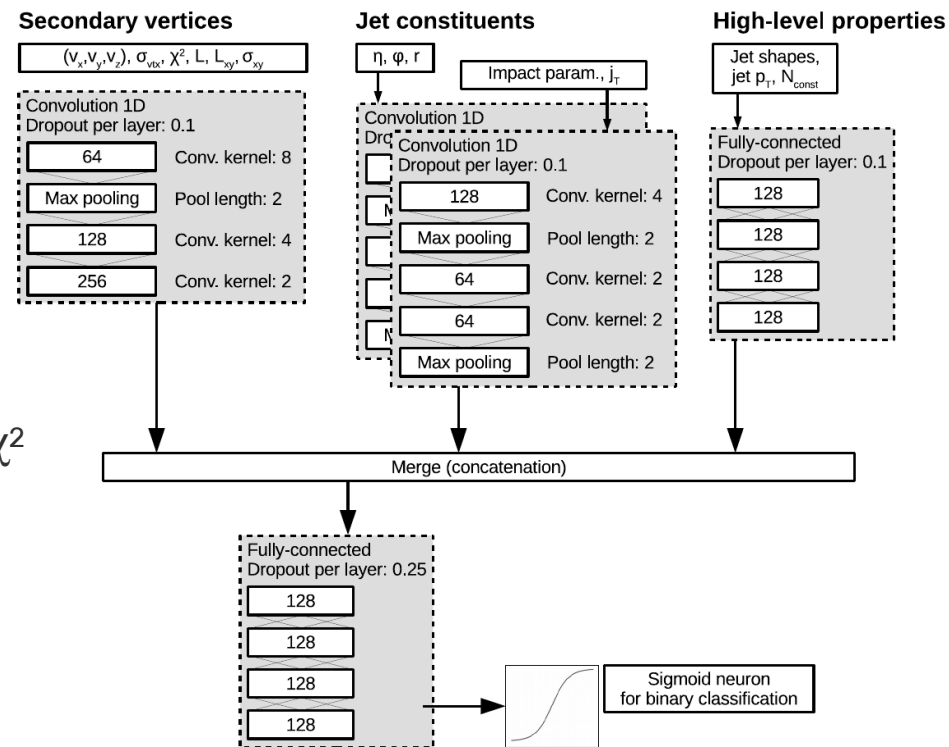
III. Machine-Learning based algorithm



- ML techniques applied to several low-level inputs: constituents, secondary vertices, track impact parameters
- General design: multibranch, multilayered neural network
- Different networks tested on different features

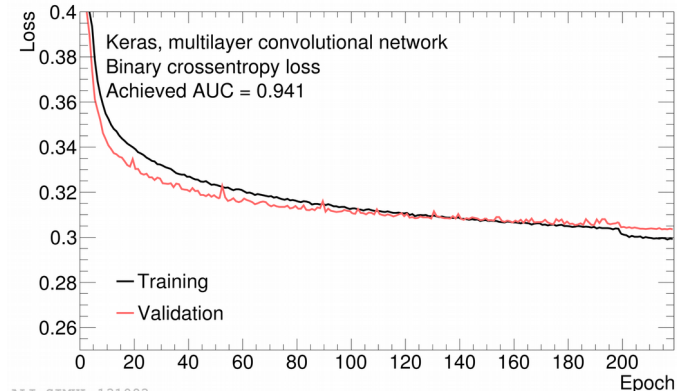
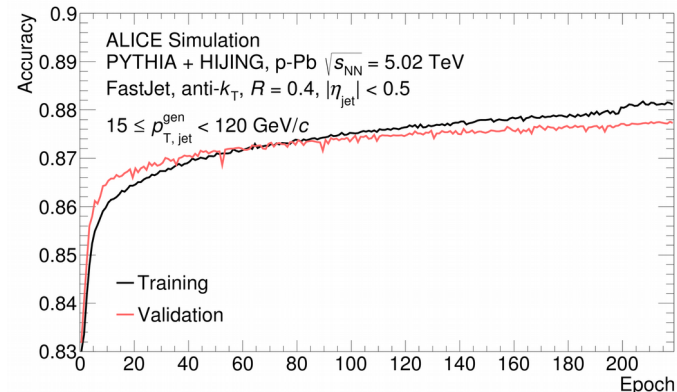
→ Features:

- Array of secondary vertices:
 - (x, y, z) rel. to primary vertex
 - Transverse plane distance and uncertainty: L_{xy}, σ_{xy}
 - Vertex track dispersion σ_{vtx} , fit quality χ^2
- Array of constituents:
 - η, φ, r (relative to jet axis)
 - Track impact parameters D, Z and j_T



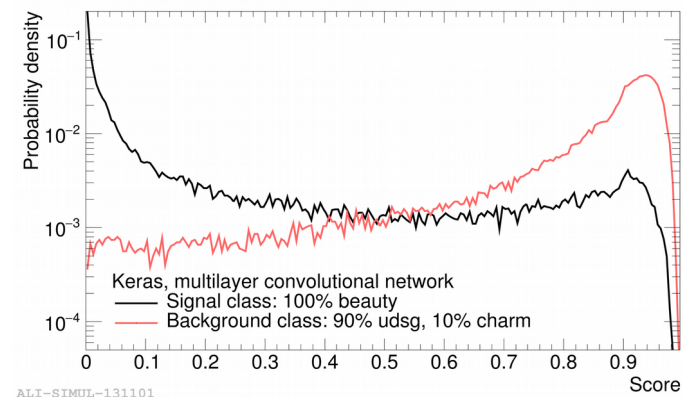
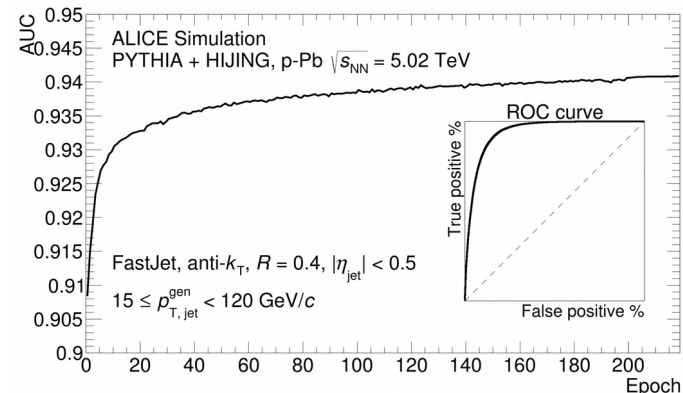
ML method: Simulations

- p-Pb at $\sqrt{s_{NN}} = 5.02$ TeV, Pythia 6 + Hijing
- FastJet anti- k_T , $R = 0.4$
- Underlying event corrected
- 200k training, 50k validation samples
 - Control parameters: accuracy and loss
 - Slow learning up to high epoch counts
 - Learning rate parameters lowered after 200 epoch
 - Not much to gain with longer training



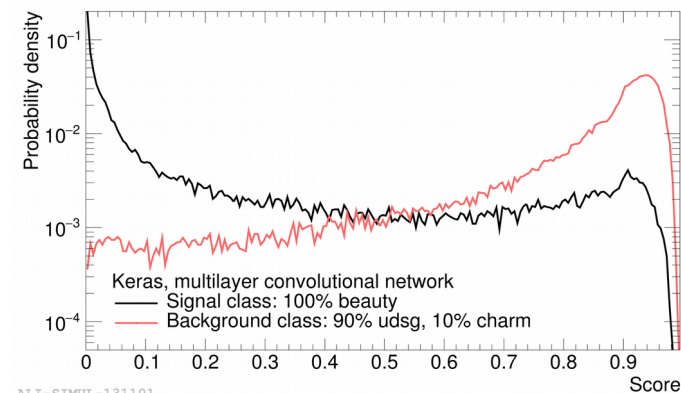
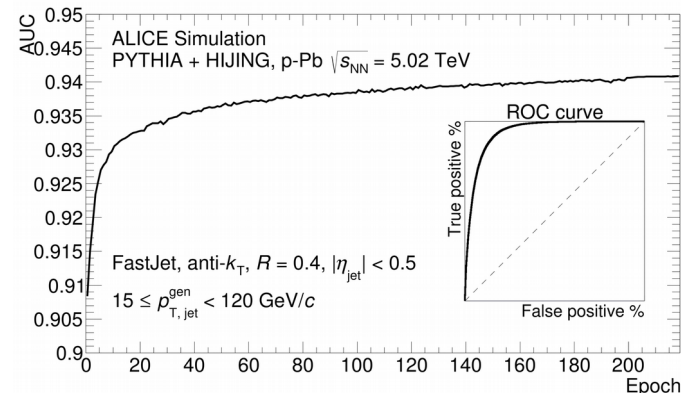
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 - AUC: Area Under ROC Curve, slow but constant learning up to 220 epoch
 - Clearly separated score distribution

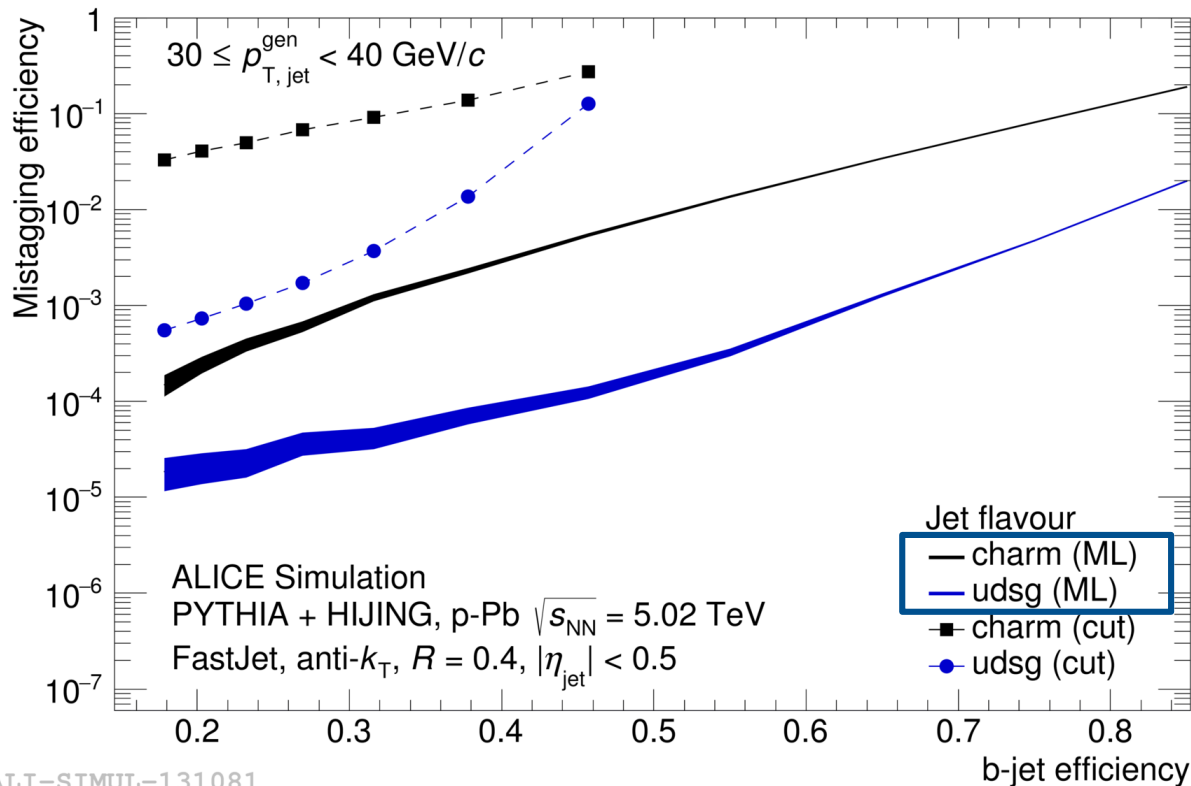


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 - Clearly separated score distribution
- Comparison to “rectangular” cuts method
- Optimizing: b-jet tagging and c-/udsg-mistagging efficiency



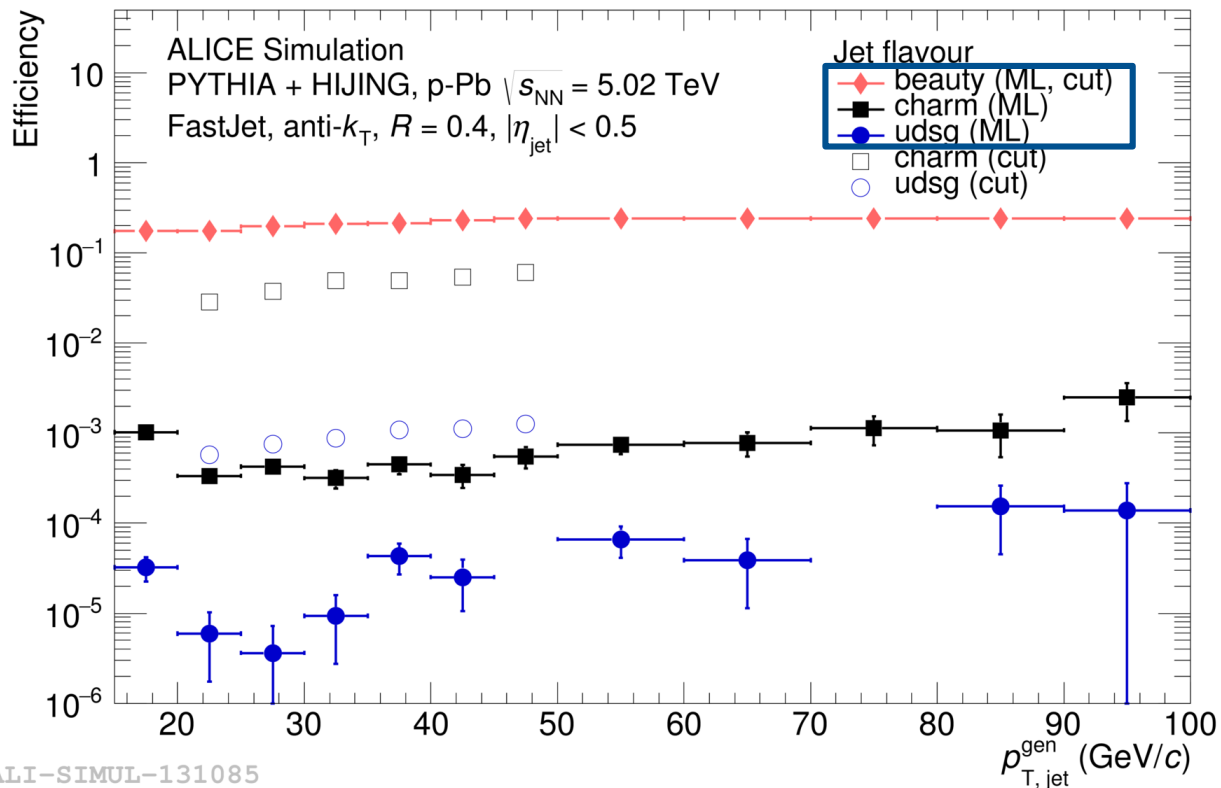
ML: b-jet vs c,udsg mistagging efficiency



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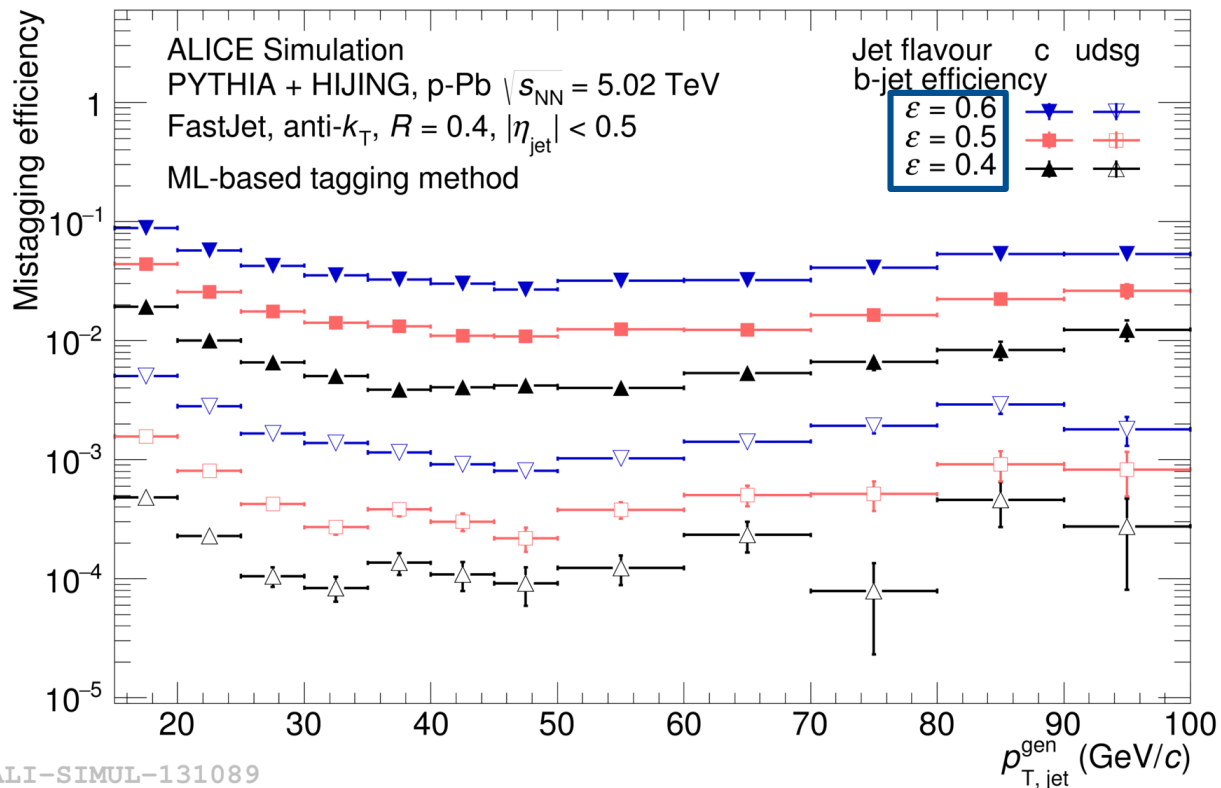
- Mistagging efficiency much lower for c-/udsg-jets
- Very promising method
- Solid: ML-based method
- Dashed: cut-based method (previous SV slides)

ML: Mistagging efficiency vs jet p_T



- Mistagging efficiency vs jet p_T
- Fixed ($\sim 20\%$) b-jet efficiency to compare to cut-based method
- Solid: ML-based method
- Open: cut-based method (previous SV slides)

ML: Mistagging efficiency



- Mistagging efficiency for higher b-jet efficiencies
 - c-jet efficiency: below 5-10%
 - udsg-jet efficiency: below 0.5-1%
- **Higher b-jet efficiency possible**
 - Solid: c-jets
 - Open: udsg-jets

Summary



- Performances of different b-tagging jet algorithms have been studied in pp and p-Pb MC simulations
 - Based on track counting, displaced secondary vertices and machine learning
 - **Very promising ML-based method in pp, p-Pb**
 - Allows for much higher b-jet efficiency
- Data analysis (in pp and p-Pb) being finalized
- Studies will be extended to Pb-Pb collisions: upcoming Pb-Pb run and run 3 and 4
 - Major detector upgrade with new ITS (x3 (x5) better spatial resolution on r_ϕ (z) coordinates), improved readout (able to sustain 50 kHz Pb-Pb collisions: collect $L_{\text{int}}=10 \text{ nb}^{-1}$, x100 gain for min. bias)
 - ➔ **Major boost for heavy-flavour jet physics with ALICE**

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Thank you !