

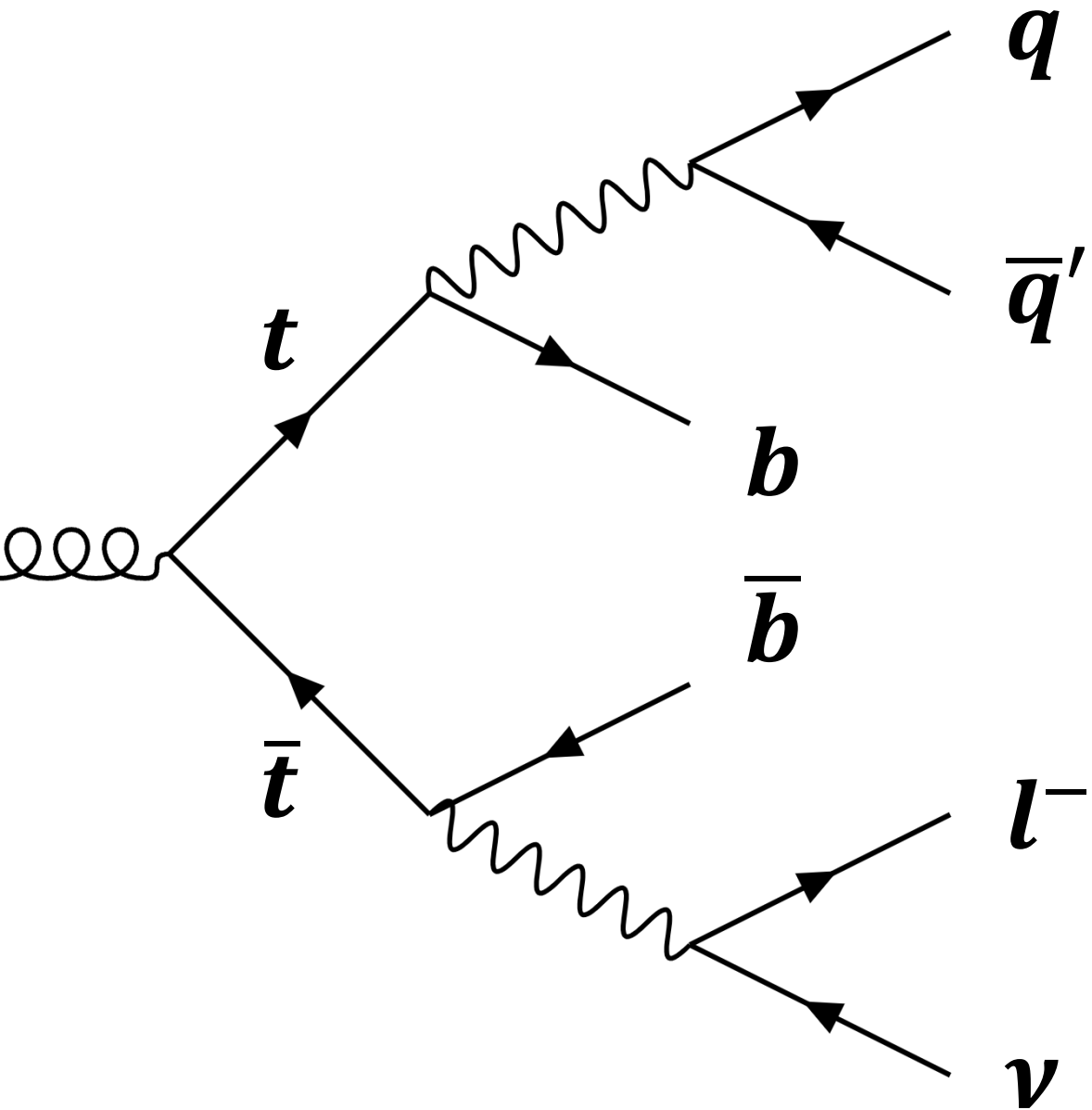
# Run 3 detector performance and object reconstruction at CMS

September 23, 2024

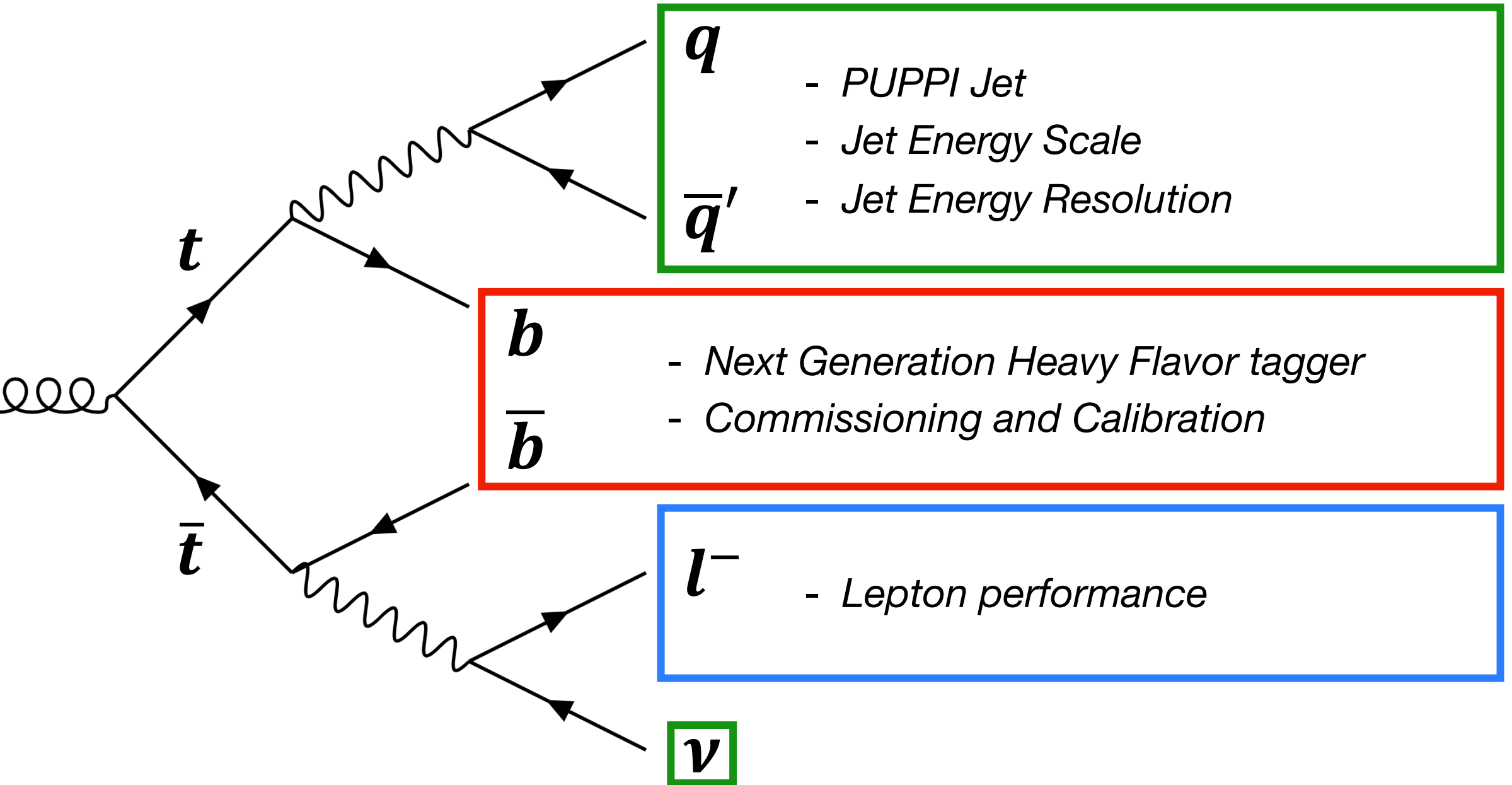
Jieun Choi

On behalf of the CMS collaboration

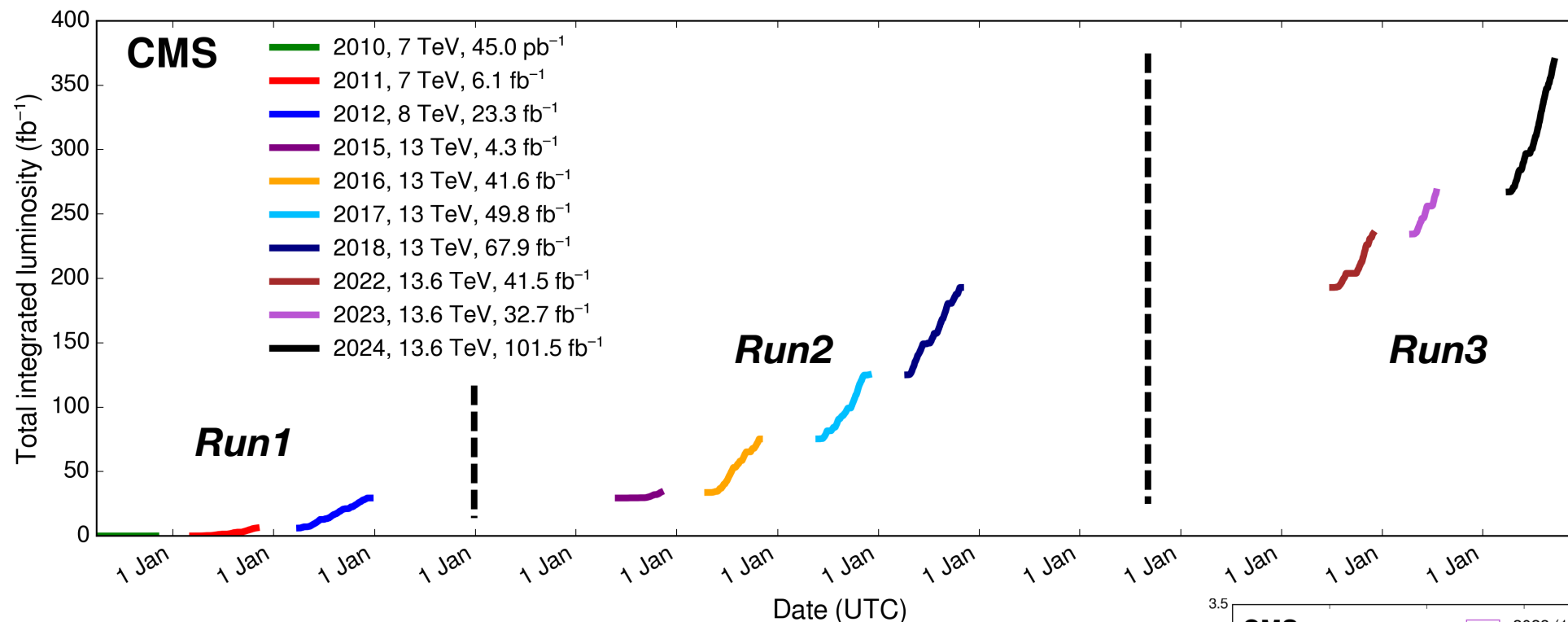
# Outline



# Outline

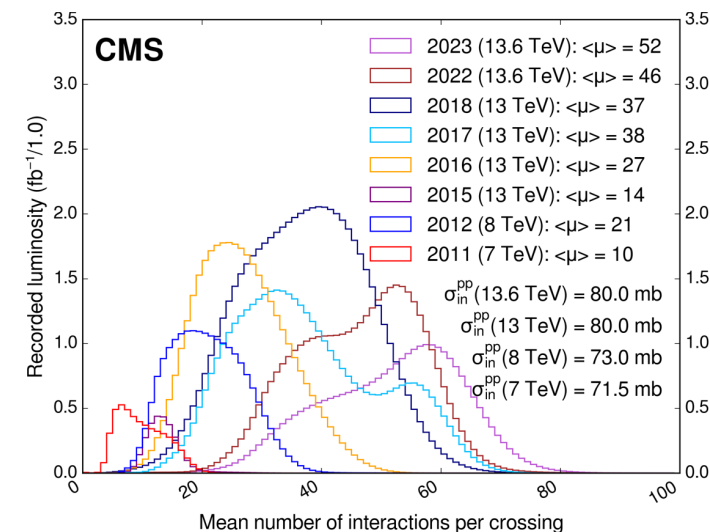


# Where we are at data taking



## Cumulative luminosity in CMS

- Luminosity delivered to CMS by the end of Run2 is > 190 /fb
- Luminosity delivered in Run3 is > 160 /fb
  - 1.3% of uncertainty in 2023
- **Average number** of pp interactions per crossing in Run3 is > 50

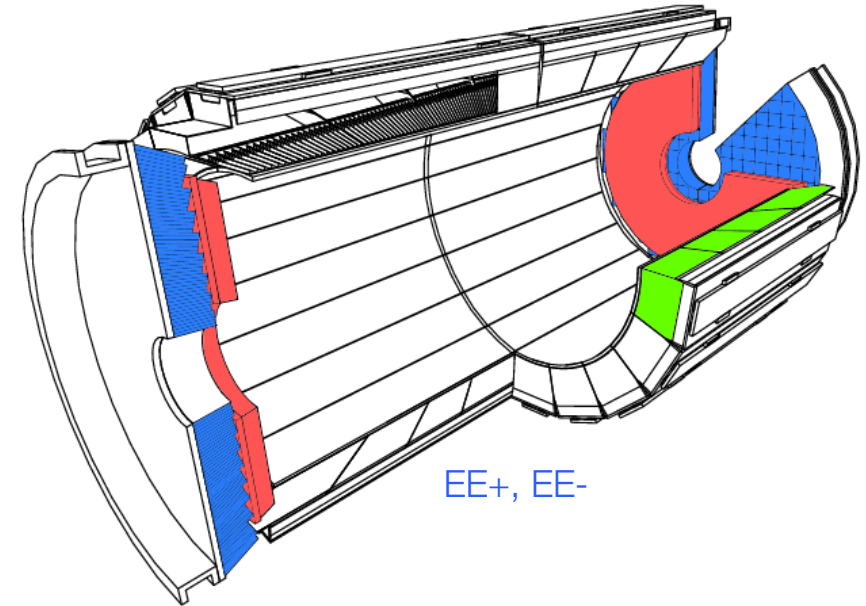




# The Known Issues in Run 3

## “ECAL power cooling issue”

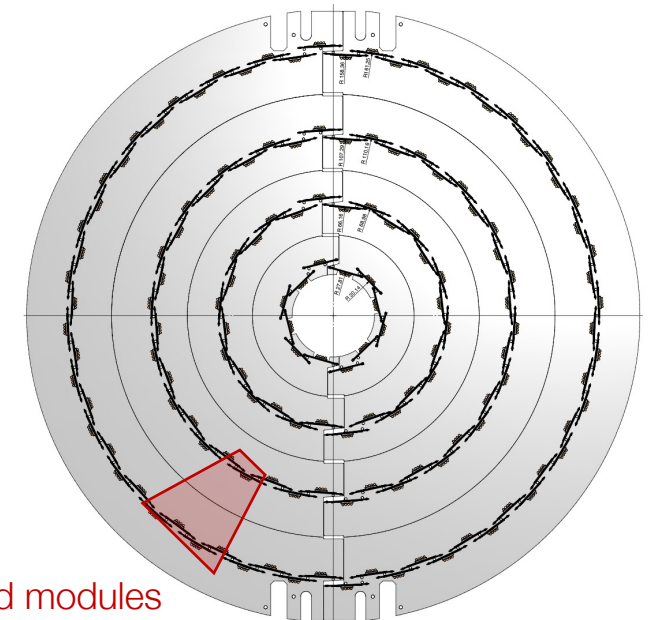
- Early September 2022
- ~7% of the positive side of the ECAL (EE+ water leak region) had turned off
- Due to cooling limitations caused by a water leak
  - [problems-and-solutions-ecal-leak-story](#)



EE+, EE-

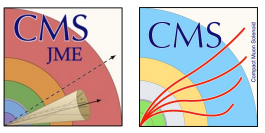
## “BPix issue”

- 19-24 June 2023 (After Technical Stop 1)
- 27 modules (~1.5%) in the Barrel Pixel Layers 3 and 4 became inoperable
- They cover a sector spanning approximately 0.4 radius (~23 degrees) in phi at negative pseudorapidity
- Due to issue in distributing the LHC clock signals



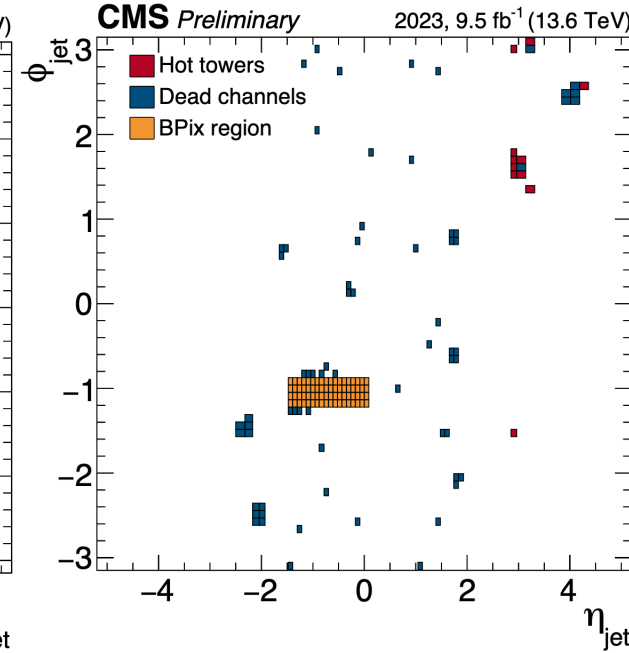
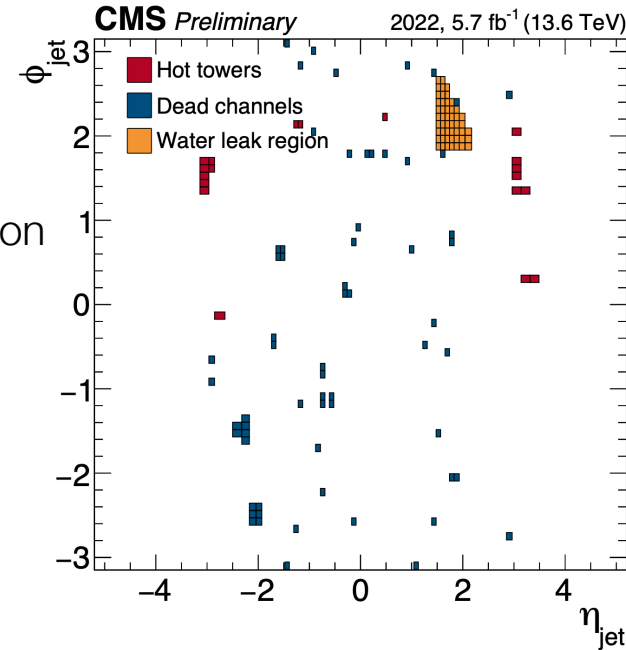
Affected modules

# From known issues to Jet



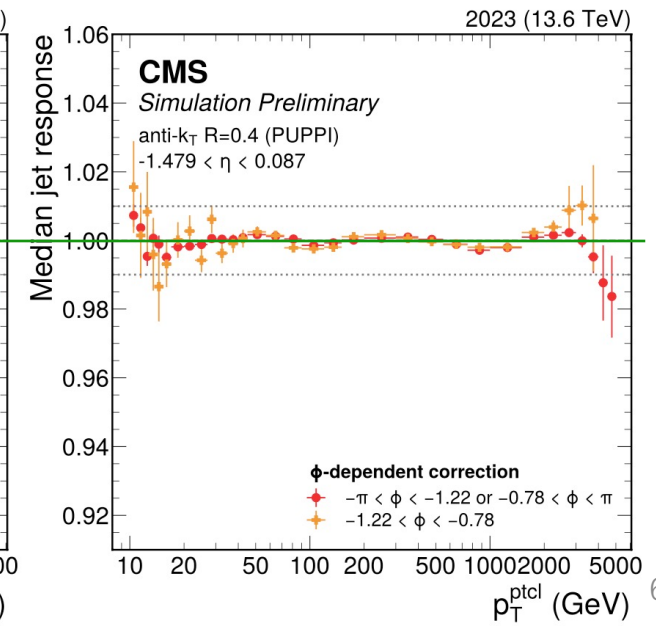
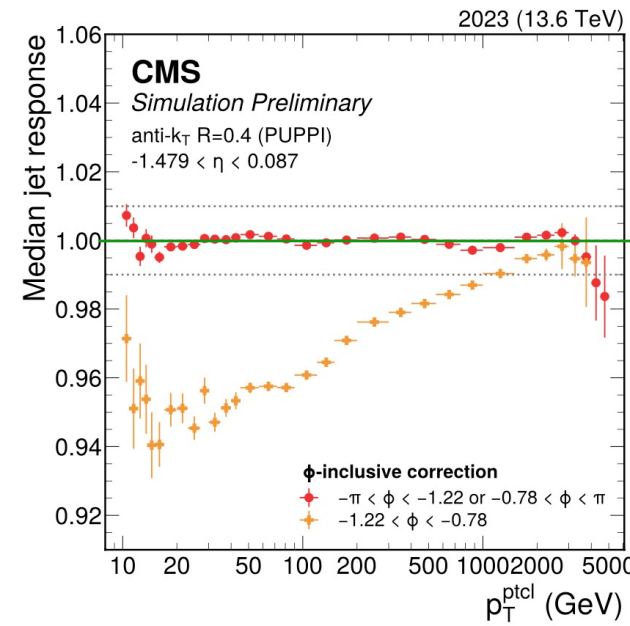
## Affected regions in Jet

- ECAL power cooling issue
  - mis-reconstruction of electrons and jets in the affected region
  - Events to be vetoed
- BPix issue
  - Inefficiency in the track reconstruction in a small region with implications for the jet energy scale (JES)
  - Events to be included

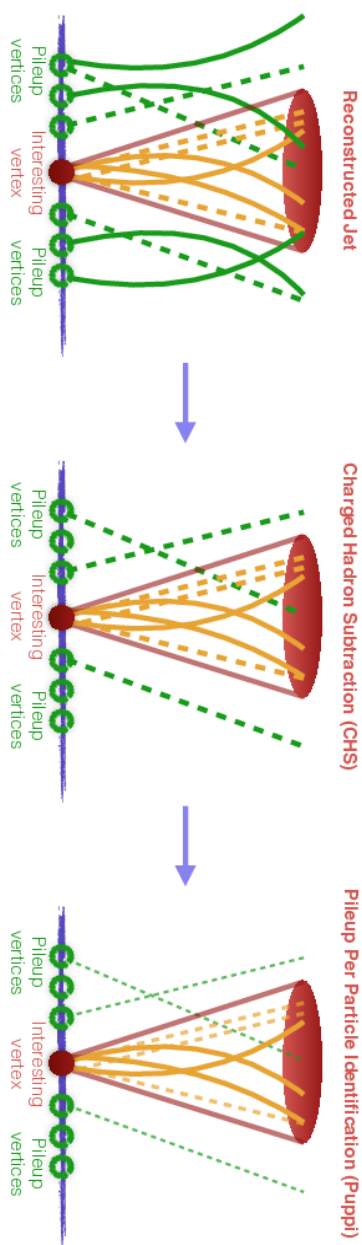
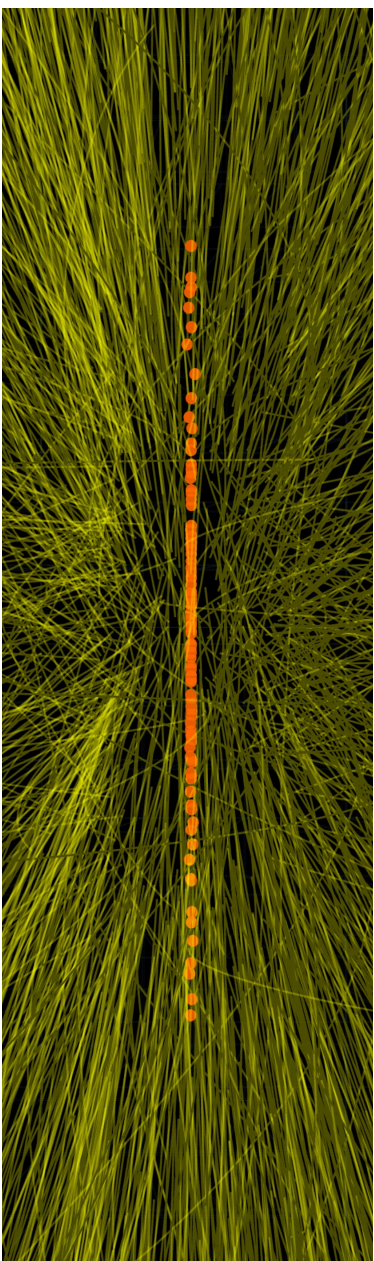
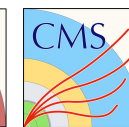
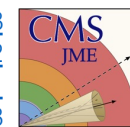


## Addressing BPix issue to JES

- Comparison from  $\phi$ -dependent vs inclusive correction
  - $\phi$ -inclusive corrections
    - a drop in the response of up to 6% is observed in the affected  $\phi$  region
  - $\phi$ -dependent corrections
    - $-1.22 < \phi < -0.78$
    - a closure within 1% is obtained in most of the analyzed phase space



# Jet Energy Performance



## PileUp mitigation

- Pileup (PU) particles can overlap with jet candidates

## Charged Hadron Subtraction (CHS) ; Run2 default

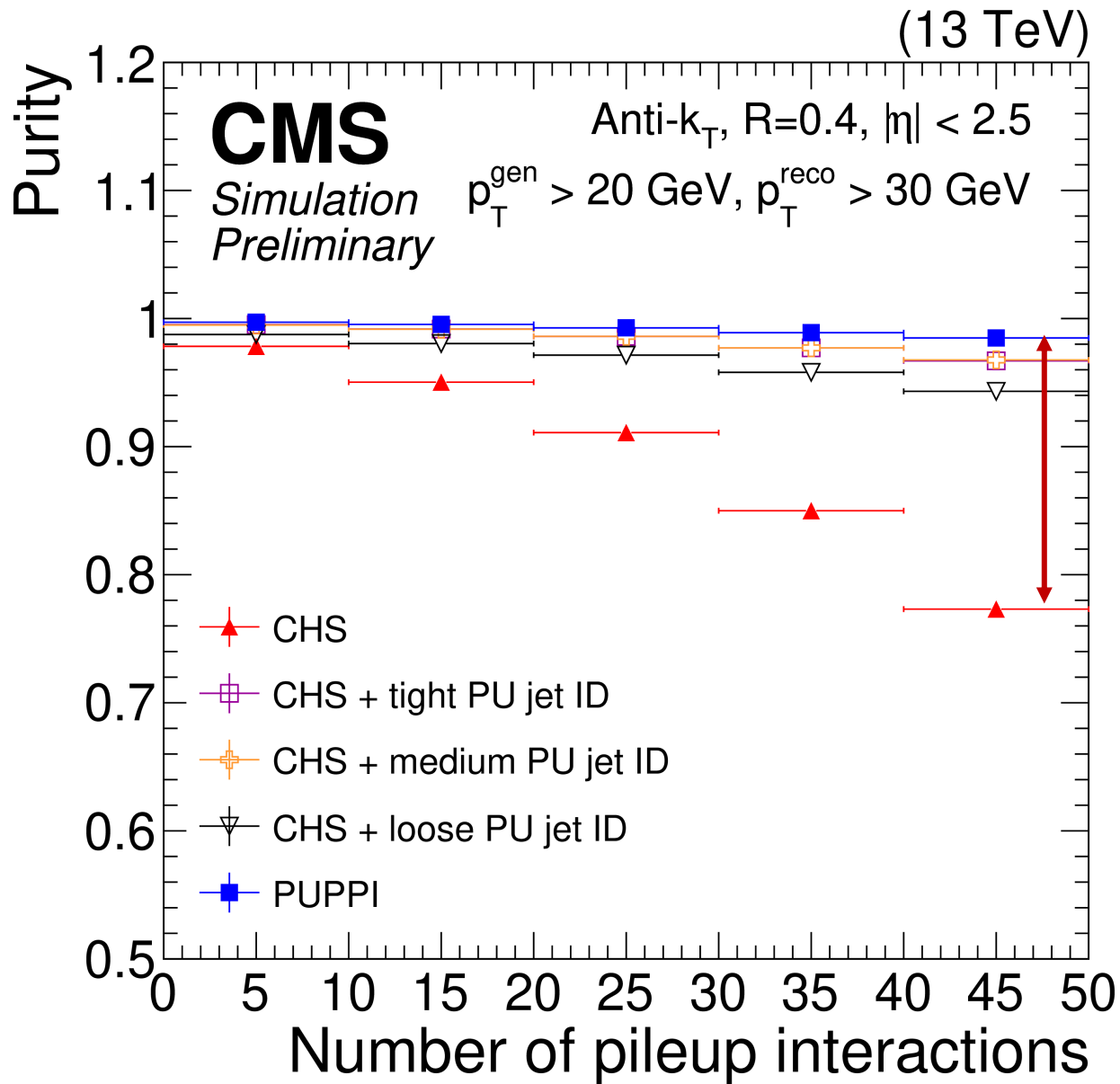
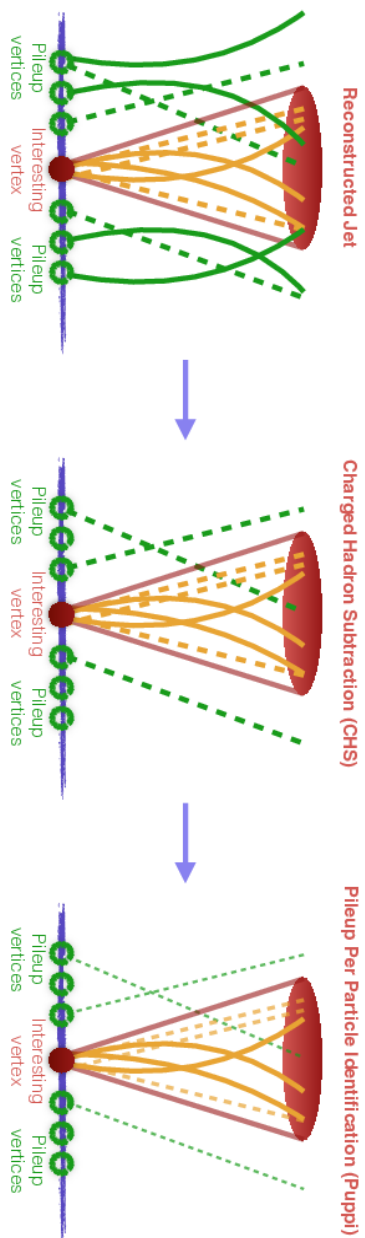
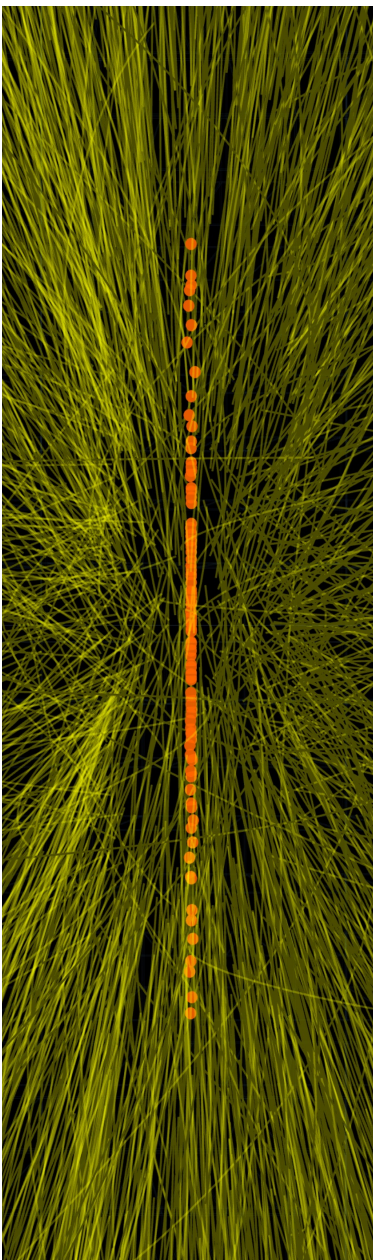
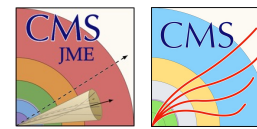
- A majority of PU is from charged particles!
- Removed charged particles originating from PU vertices
- Only works within the tracker covered region
- Does not remove neutral PU contribution

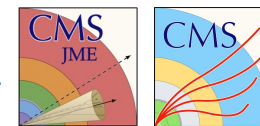
## PileUp Per Particle Identification (PUPPI) ; Run3 default

- Intended to remove PU + identify PU at particle level
- Event-by-event basis **weights** for each particle if they are PU-like
- Used to re-scale the four momenta of the particles
- **Jet+MET, jet substructure variables are expected to be less susceptible to PU**



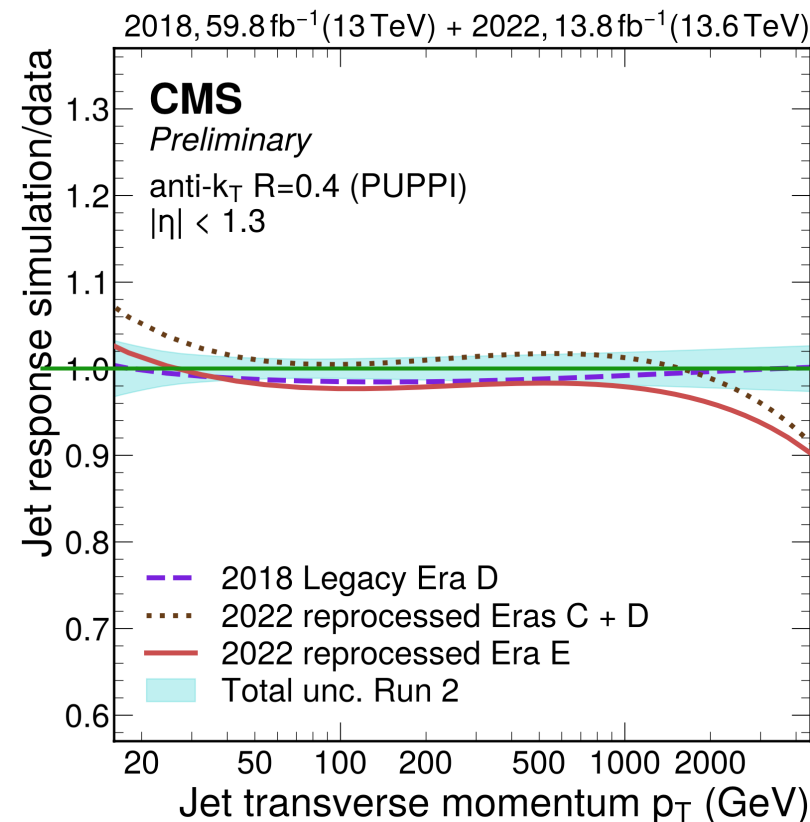
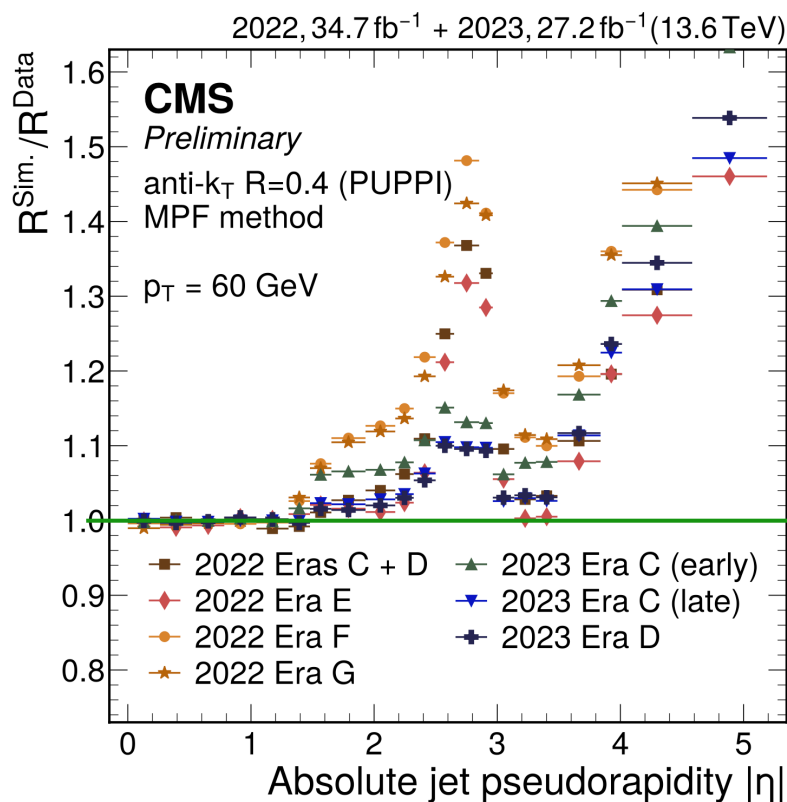
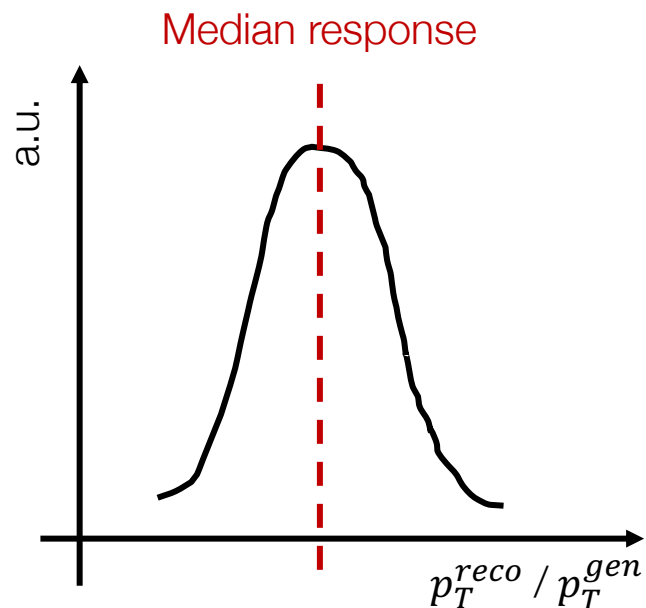
# Jet Energy Performance





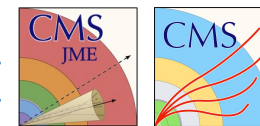
## Residual correction

- Jet energy corrections are applied to the reconstructed-level jets
  - $\eta$ -dependent correction (different response from each subdetector) +  $p_T$ -dependent correction (scale difference in central)
- Great performance in the barrel region (correction < 2%)
- Stable difference between data and simulation in 2022
  - Similar performance to Run 2 legacy reconstruction in  $50 < p_T < 500$  GeV



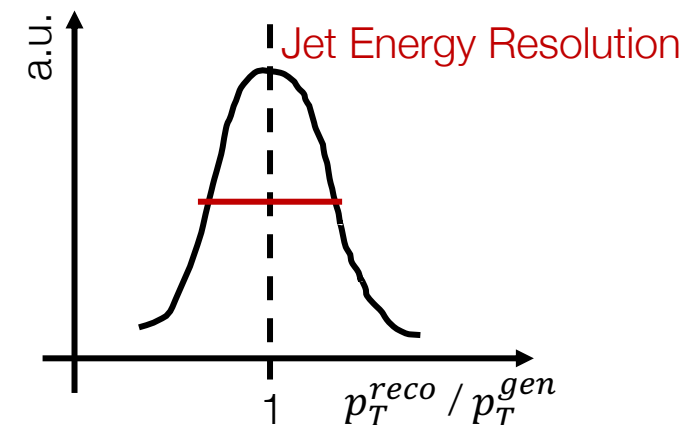


# Jet Energy Resolution



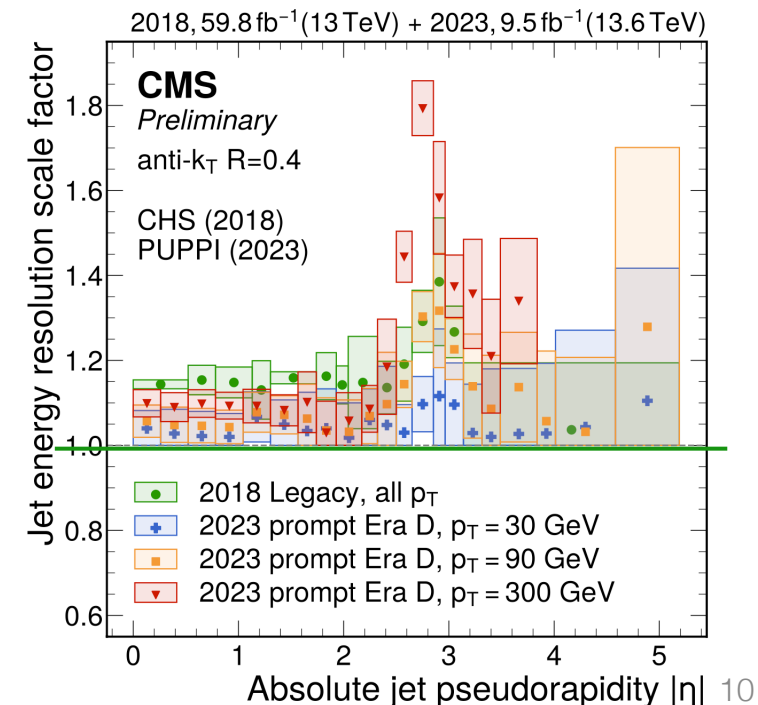
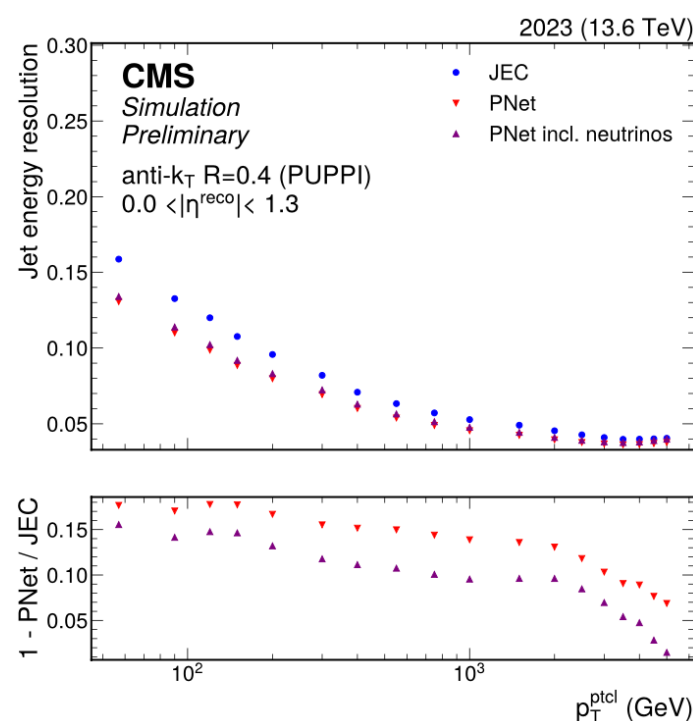
## Smearing

- Jet energy resolution (JER) in simulation needs to be smeared to match that of data
  - Defined as the spread of the response distribution; gaussian fit
  - SFs are extracted using data-based methods
- Prompt construction in  $|\eta| < 2.5$  outperforms legacy Run2
- The imperfect calibration of overlapping sub-detectors ( $2.5 < |\eta| < 3$ ) leads to larger SFs



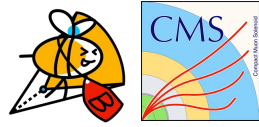
## $p_T$ regression

- ML based algorithm **ParticleNet**
  - Jet classification
  - Jet  $p_T$  regression
  - Jet energy resolution estimation
- First full calibration of flavor-aware regressed  $p_T$  for small-cone jets
  - Correcting raw jet  $p_T$  to the truth-level
- Significant resolution improvement  $\sim 15\%$
- Calibration with data gives a non-closure of 2-5 % in  $|\eta| < 2.5$



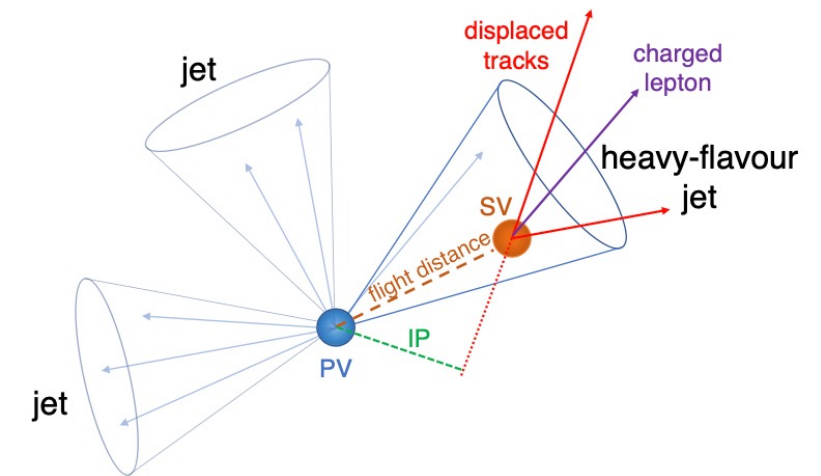
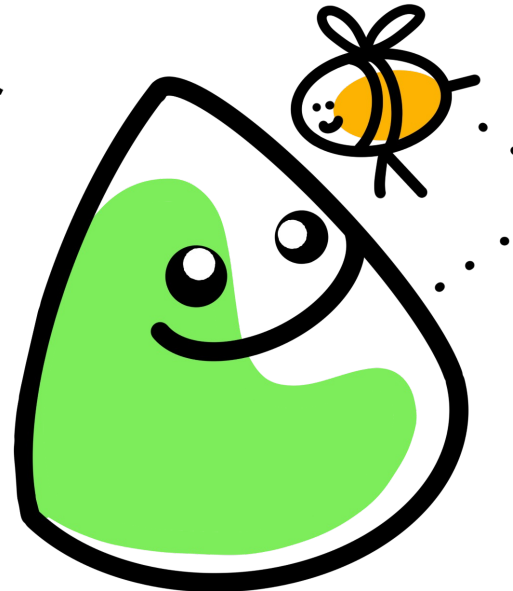
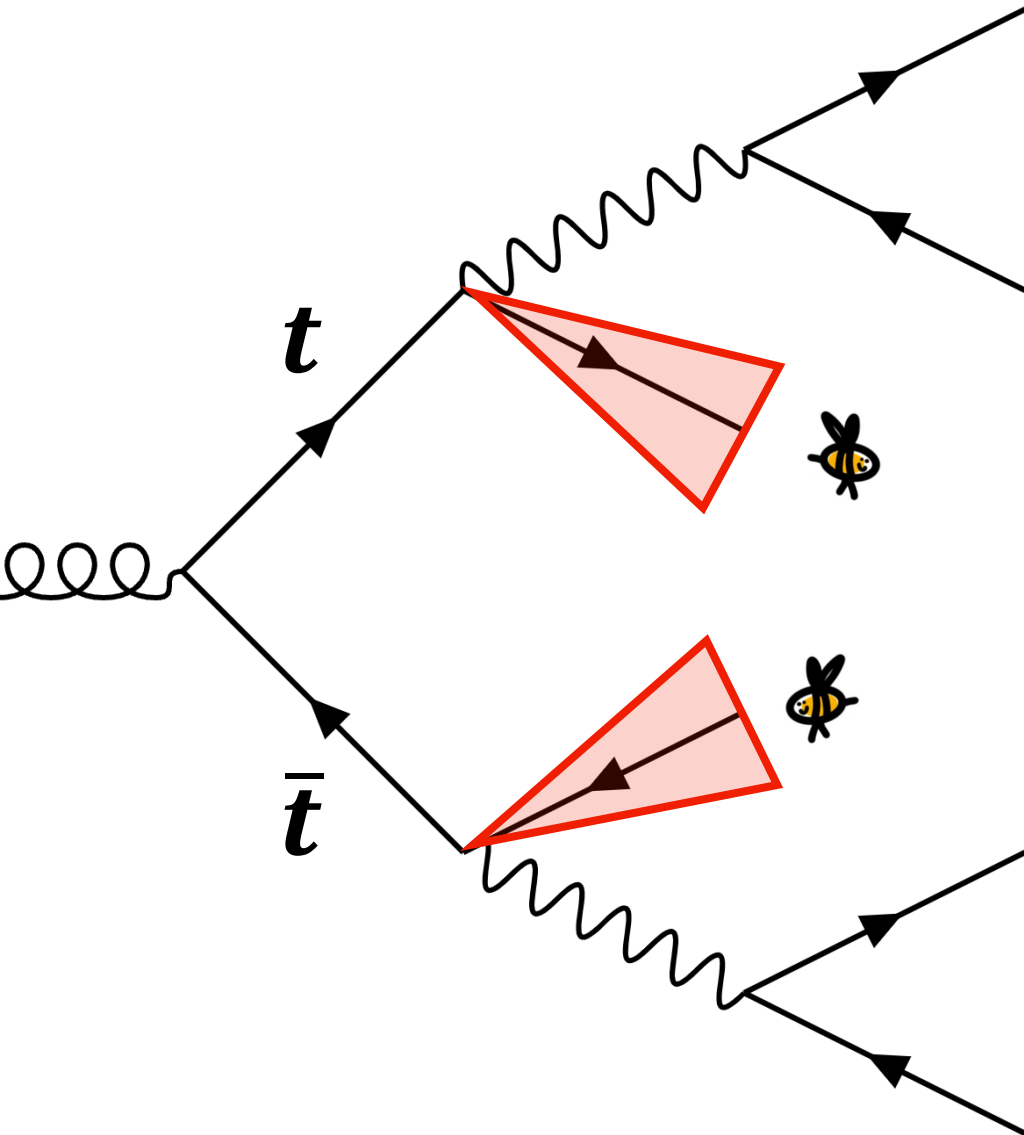
# Heavy flavor tagging in CMS

Picture taken from  
2018 JINST 13 P05011



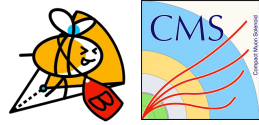
## Heavy flavor jet in Top topology

- Quarks and gluons hadronized and form particle jets
- Top quarks decay before hadronization
  - Almost exclusively  $t \rightarrow bW$  decays
  - Final state determined by W boson decays



- B hadrons live long enough to resolve decay
  - Exceptional signature
    - displaced vertices, soft lepton inside..
  - Significant improvement in machine learning based tagging algorithms
  - Performs beyond the heavy flavor tagging (s, tau-jets)

# The evolution of Jet taggers

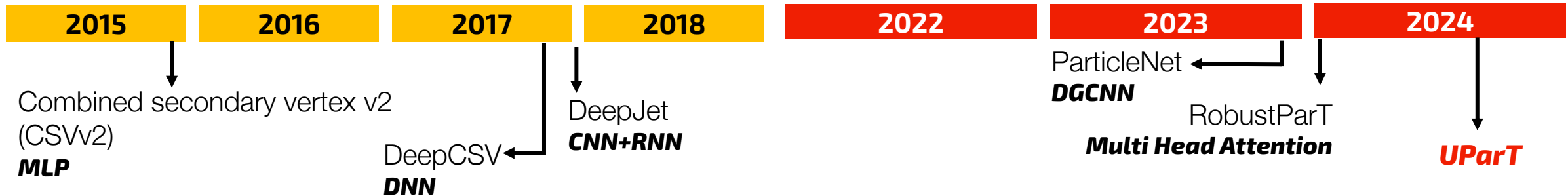


## Run 2

- Shallow ML: BDTs or feedforward NNs
- Deep ML: sequence-based deep NNs

## Run 3

- Particle Clouds: Graph Neural Networks
- Transformers: Attention model



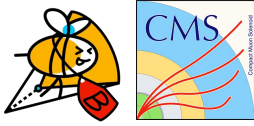
## Extension to a unified approach

- From Transformer Model (Particle Transformer)
  - Based on the “Attention” model designed for particles
  - Not only single particle information, but also pair-wise features
- To **Unified Particle Transformer (UParT)**
  - b / c jet identification
  - Hadronic tau + s-tag (!)
  - Simultaneous flavor aware jet energy/resolution regression

**Discriminators are determined:**

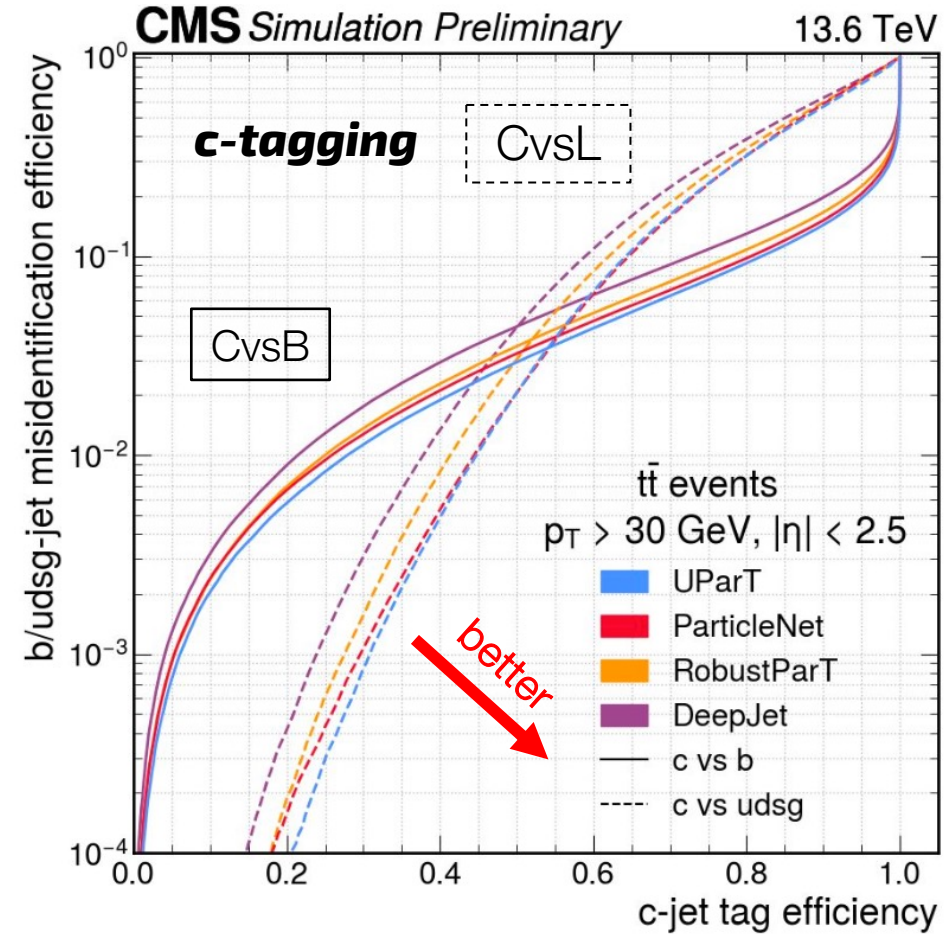
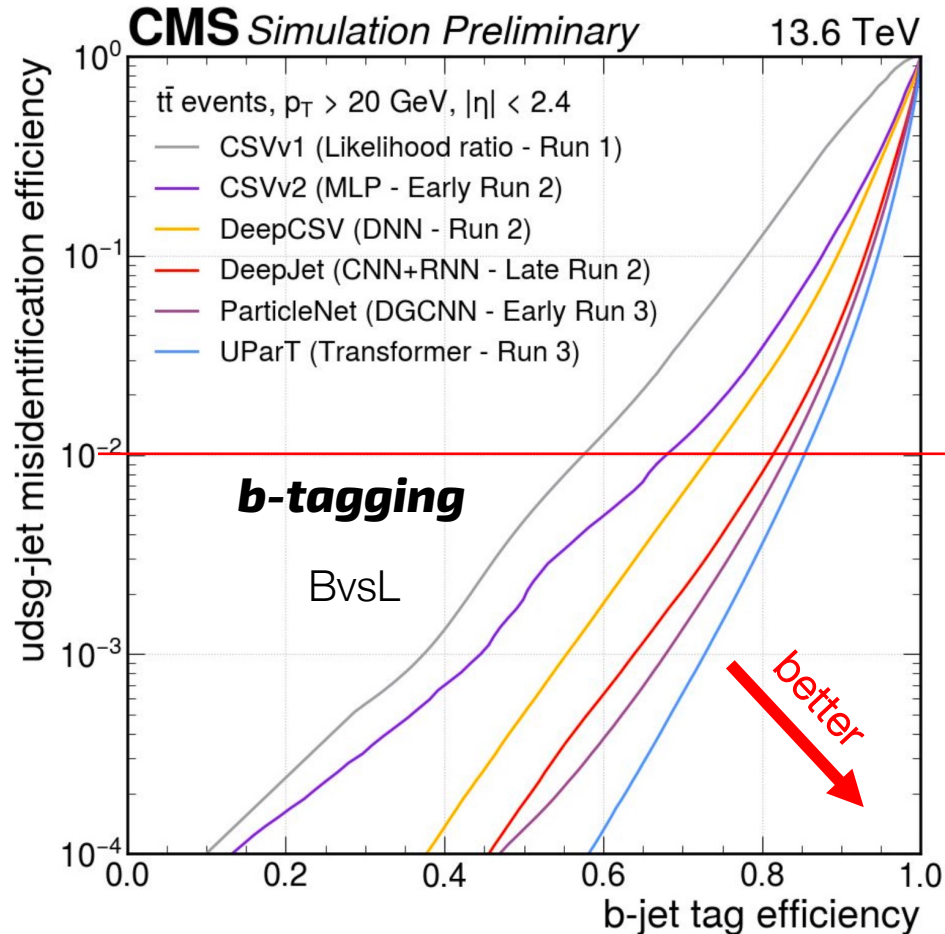
$$X \text{ vs } Y = \frac{P(X)}{P(X) + P(Y)}$$

# Flavor Tagging Performance



## b/c-tagging performance

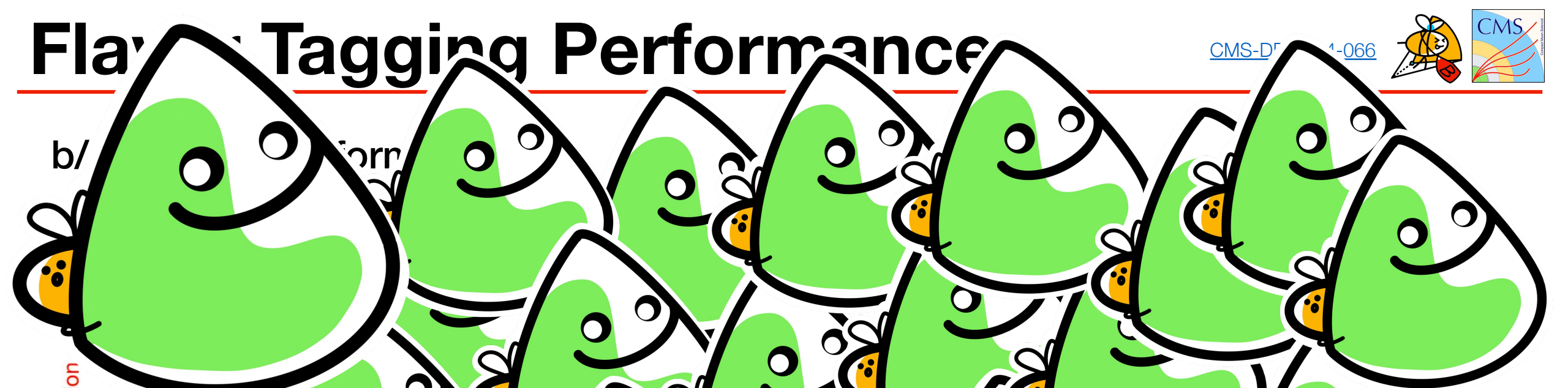
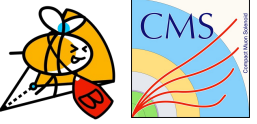
- Promising performance compared to previous taggers
  - ~5% better b-tag efficiency than DeepJet having the same misidentification rate at 1%
  - The highest/comparable c-tag efficiencies compared to the previous taggers



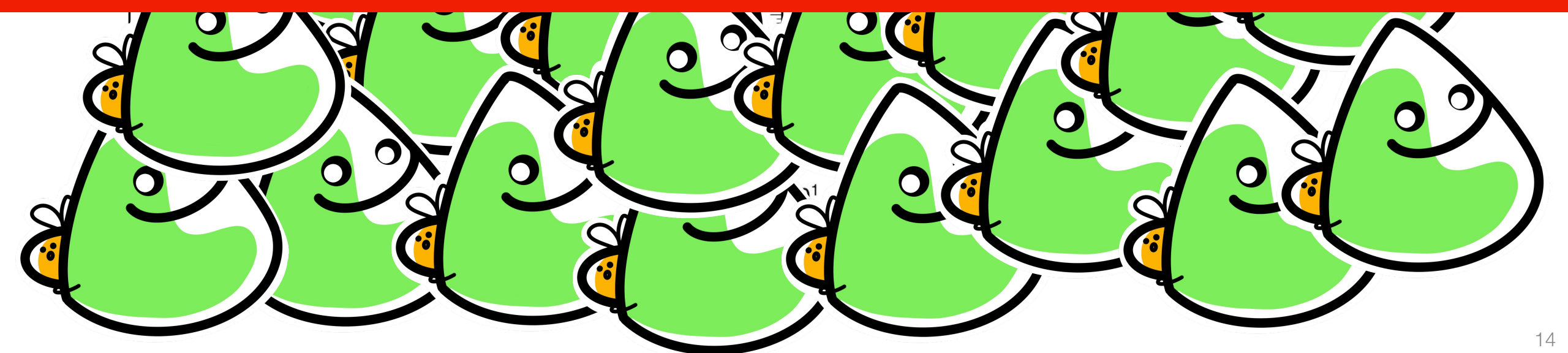


# Flavor Tagging Performance

CMS-DP 1-066



***~10% more  $t\bar{t}$  than Run2!***



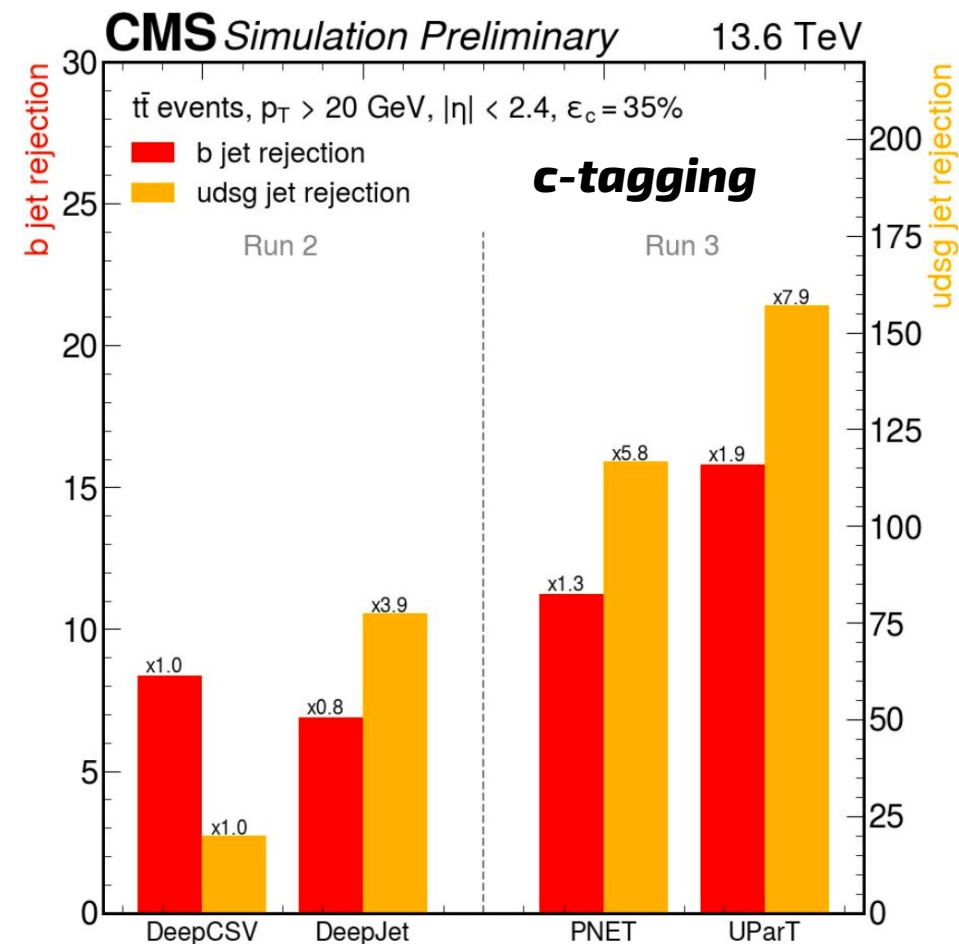
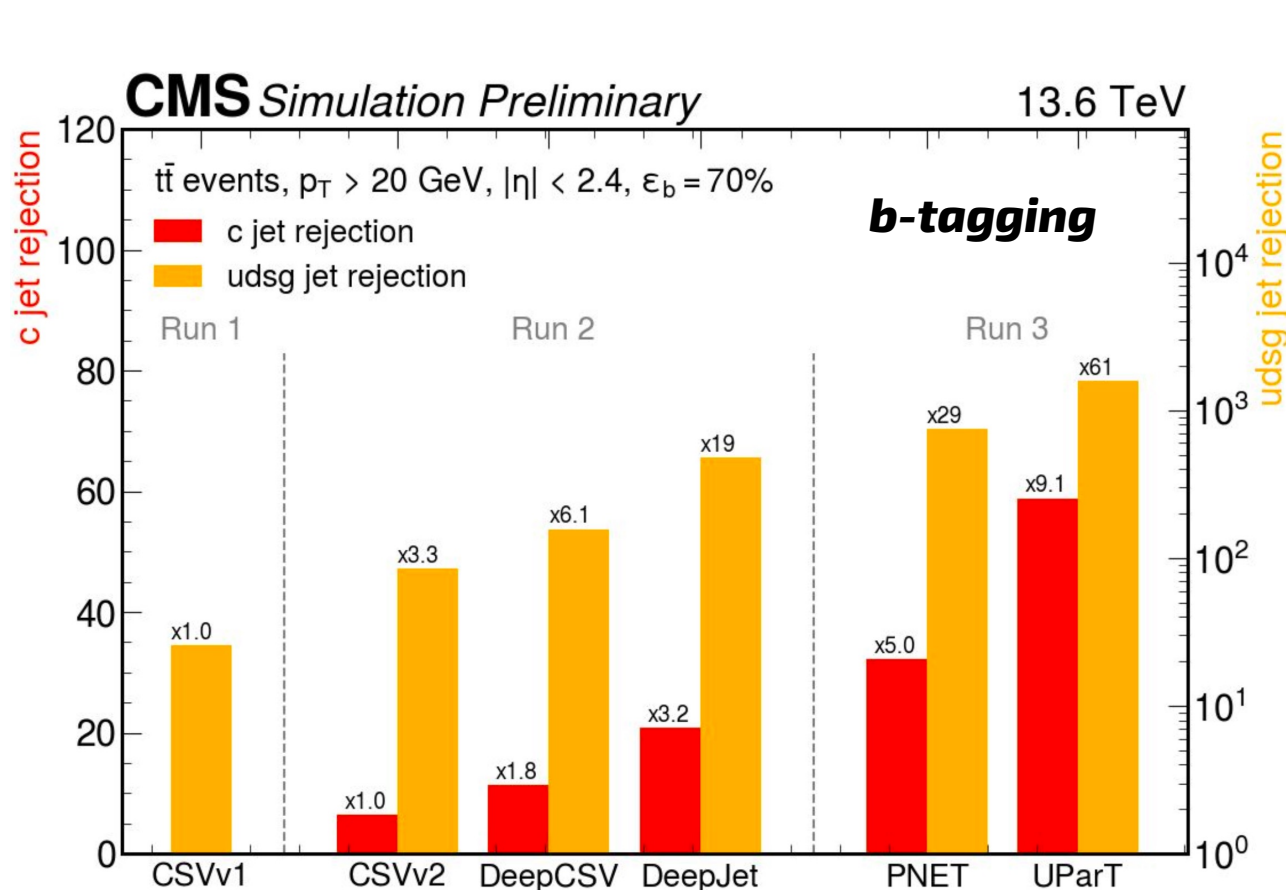


# Flavor Tagging Performance

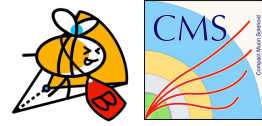


## b/c-tagging performance

- Promising performance compared to previous taggers
  - x3 better light jet rejection (at b-jet eff 70%) than DeepJet
  - x2 better light rejection + x2 better b-jet rejection (at c-jet eff 35 %) than DeepJet



# Data vs Simulations

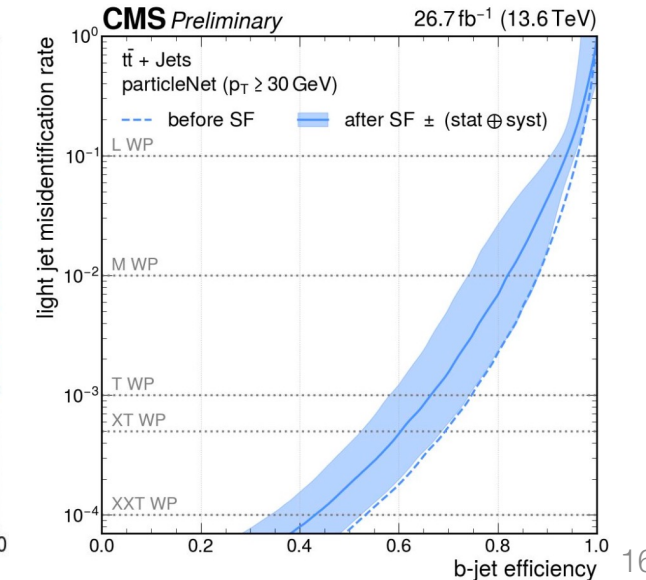
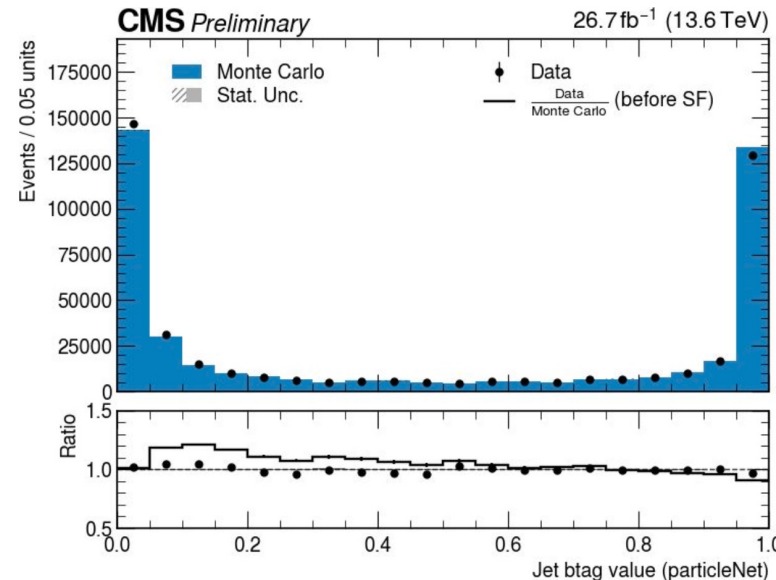
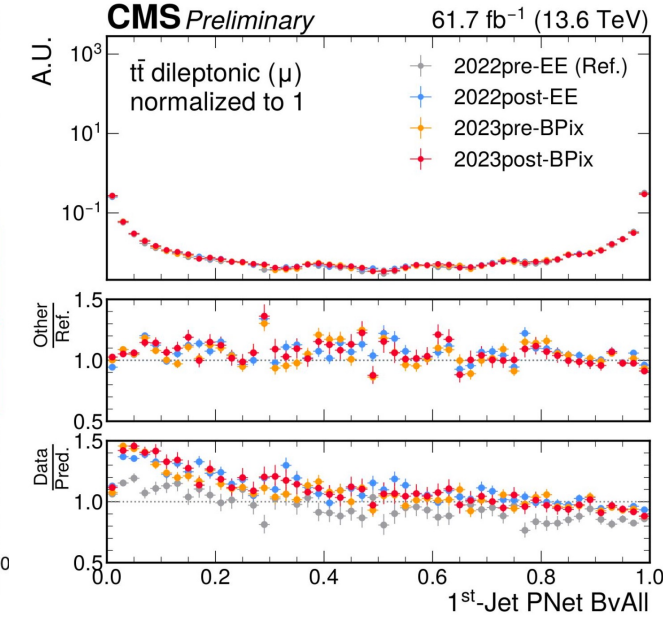
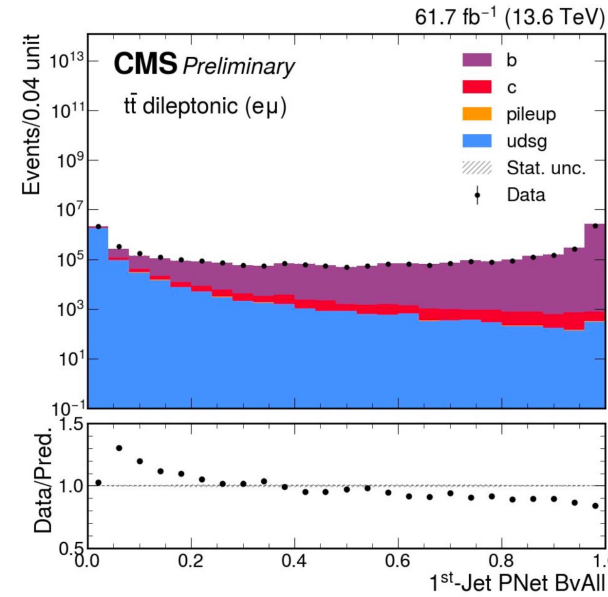


## Commissioning

- Disagreement due to mismodelling in simulations
  - input variables  $\rightarrow$  output discriminator
- Prone to changes in the calibration of the detector alignment
- Only small differences in b-tag output discriminator for different data taking period in Run3
  - Performed in CvsL and CvsB as well

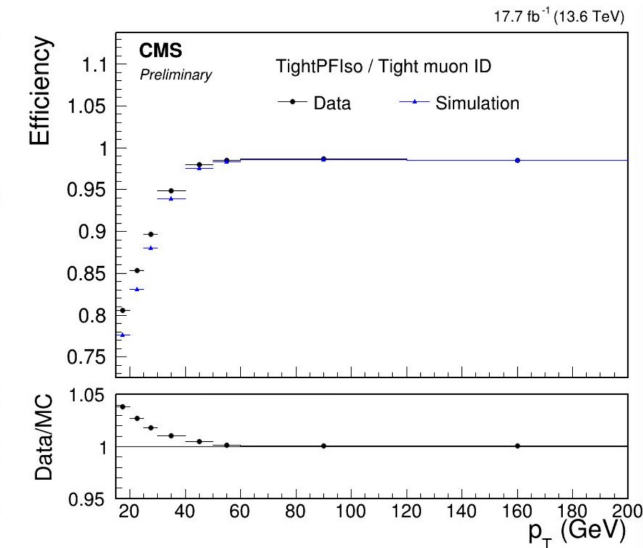
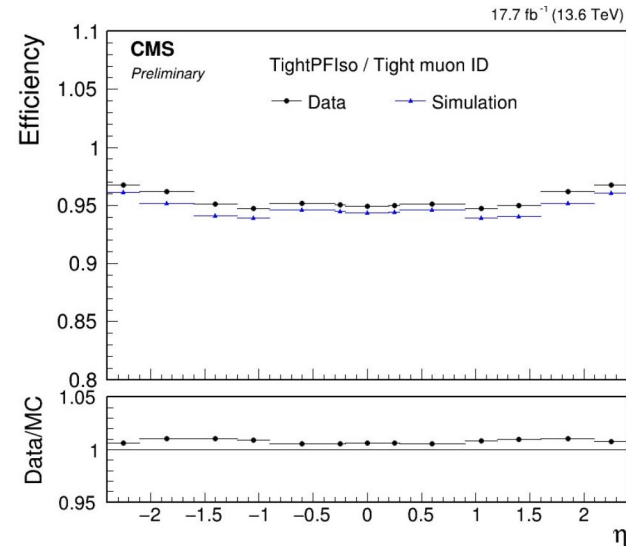
## Calibrations

- Scale-Factors defined from efficiencies for a jet with flavor in data / simulation
- Define heavy flavor enriched data for SF derivation
  - Phase space dedicated to class
- Good Data vs MC closure after applying SFs
- Performance changed after SFs application
- No public results in c-tagging yet



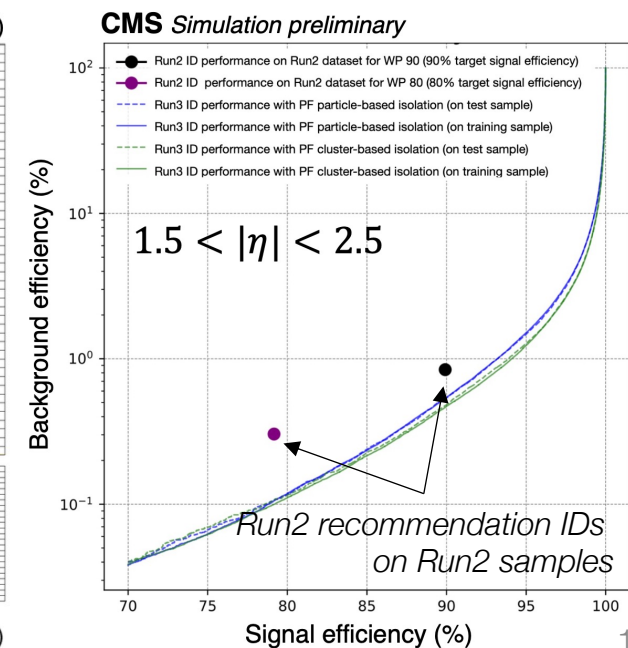
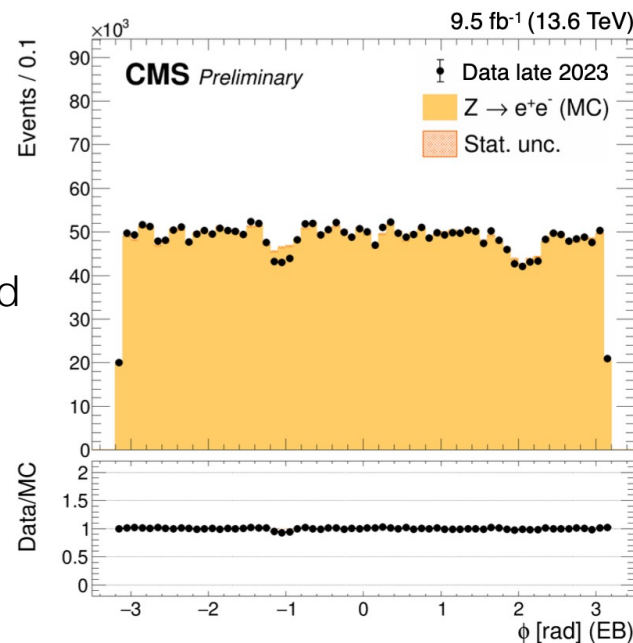
## Muon performance

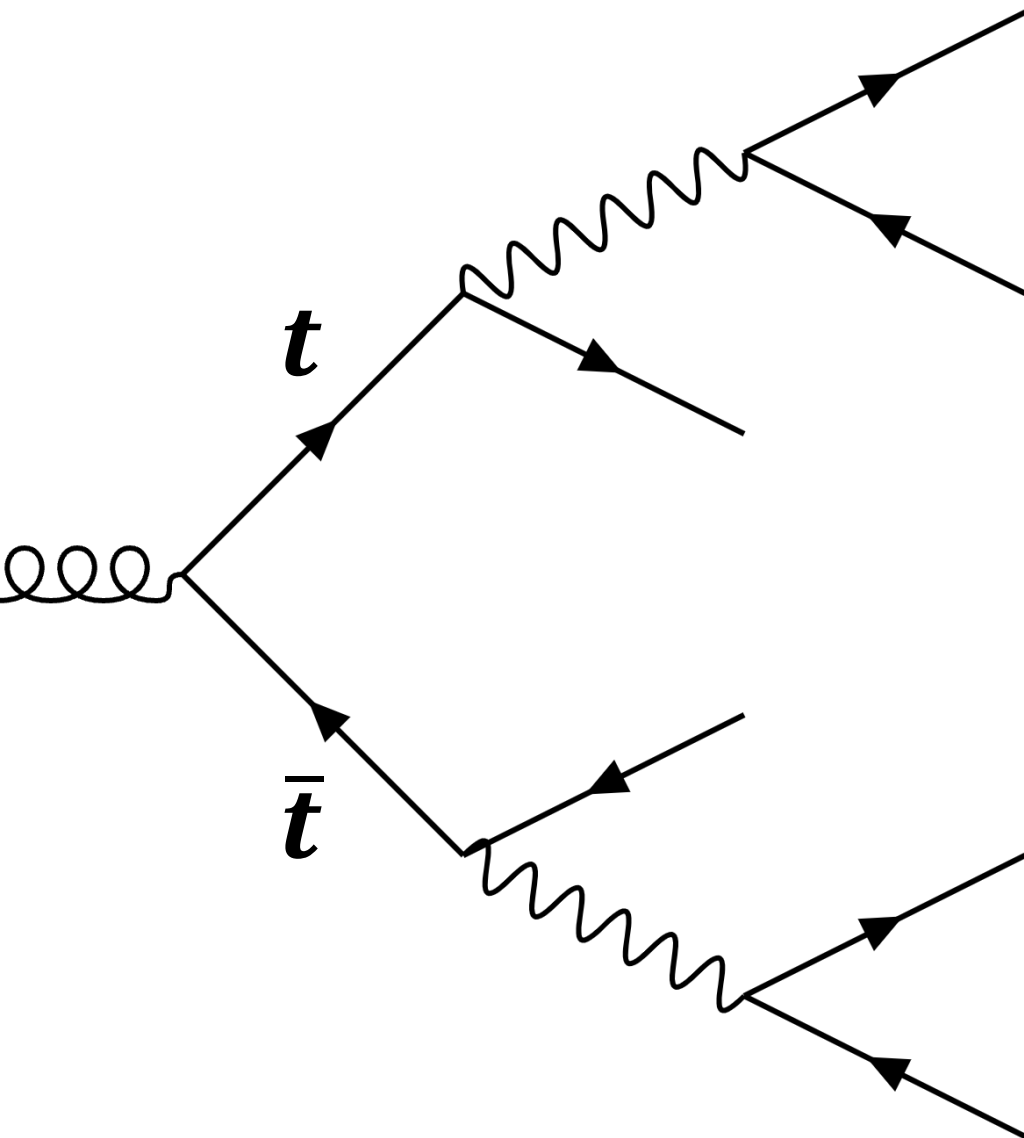
- Good data-to-simulation performance for Muon identification
- Stable isolation performance



## Electron Performance

- Minimal inefficiency in BPix failure region
- A BDT-based (XGBoost) multivariate optimization is performed to discriminate background electrons from signal ones
- The cut-based and MVA-based electron and photon IDs were tuned to ensure a better performance than the Run2





## Jet+MET

- PUPPI as the default algorithm for pileup suppression in Run3
- JES: Excellent and stable performance in the barrel region
- JER: Run3 outperforming Run 2 legacy in  $|\eta| < 2.5$
- Full calibration of flavor-aware pT regression for small-cone jets

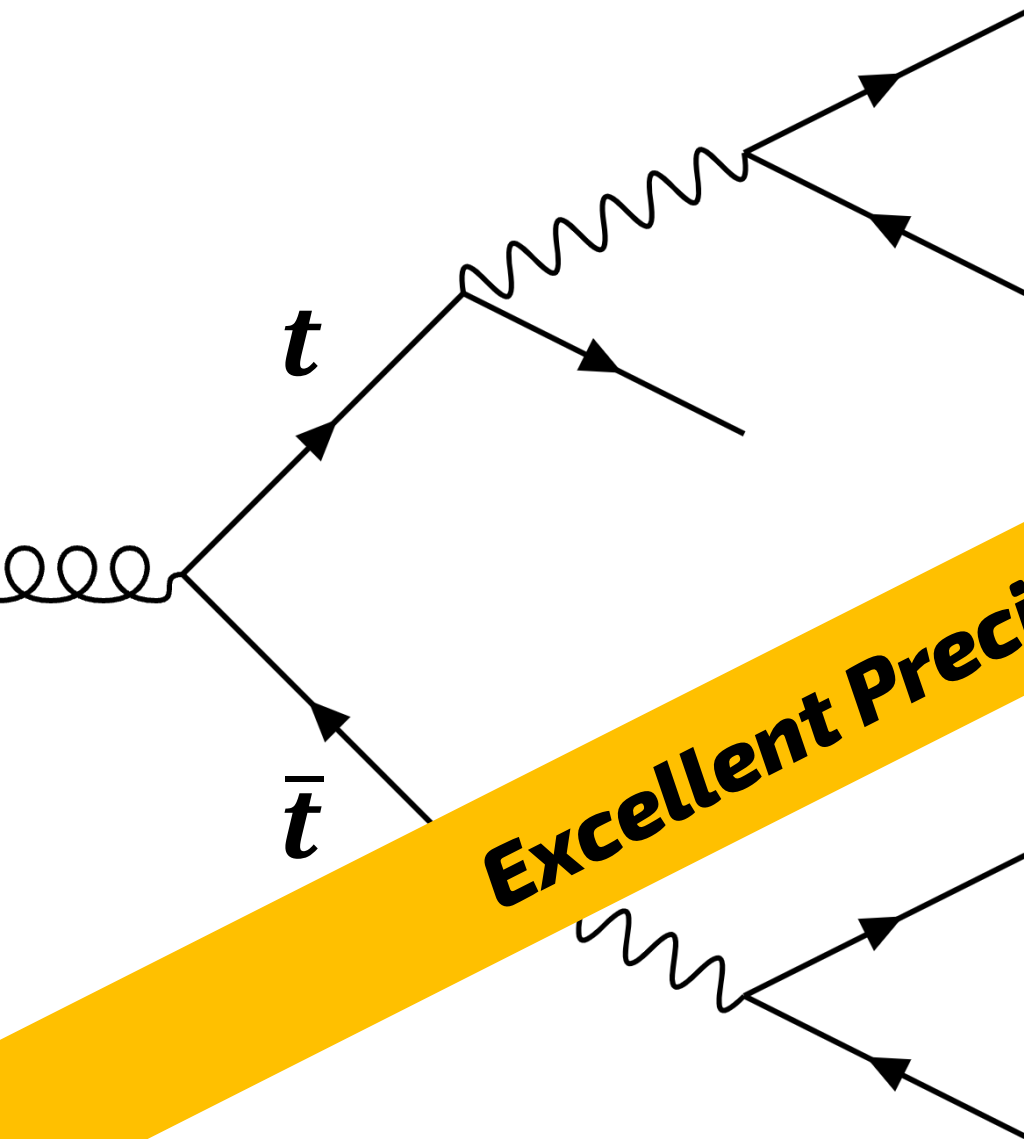
## Heavy flavor tagging

- UParT shows promising performance in Run 3
- 5% better b-jet efficiency than DeepJet in Run2 in the same light jet misidentification rate at 1%
- Good Data vs MC closure after applying SFs

## Leptons

- Good and stable data-simulation performance in Muon
- Better Electron / Photon ID than in Run 2

# Summary



## Jet+MET

- PUPPI as the default algorithm for pile-up mitigation in Run3
- JES: Excellent and stable performance in the barrel region
- JER: Outperforming in Run 3
- Full calibration of jet energy for regression for small-cone jets

## Heavy Flavor Tagging

- Shows promising performance in Run 3
- Better b-jet efficiency than DeepJet in Run2 in the same light jet misidentification rate at 1%
- Good Data vs MC closure after applying SFs

## Leptons

- Good and stable data-simulation performance in Muon
- Better Electron / Photon ID than in Run 2

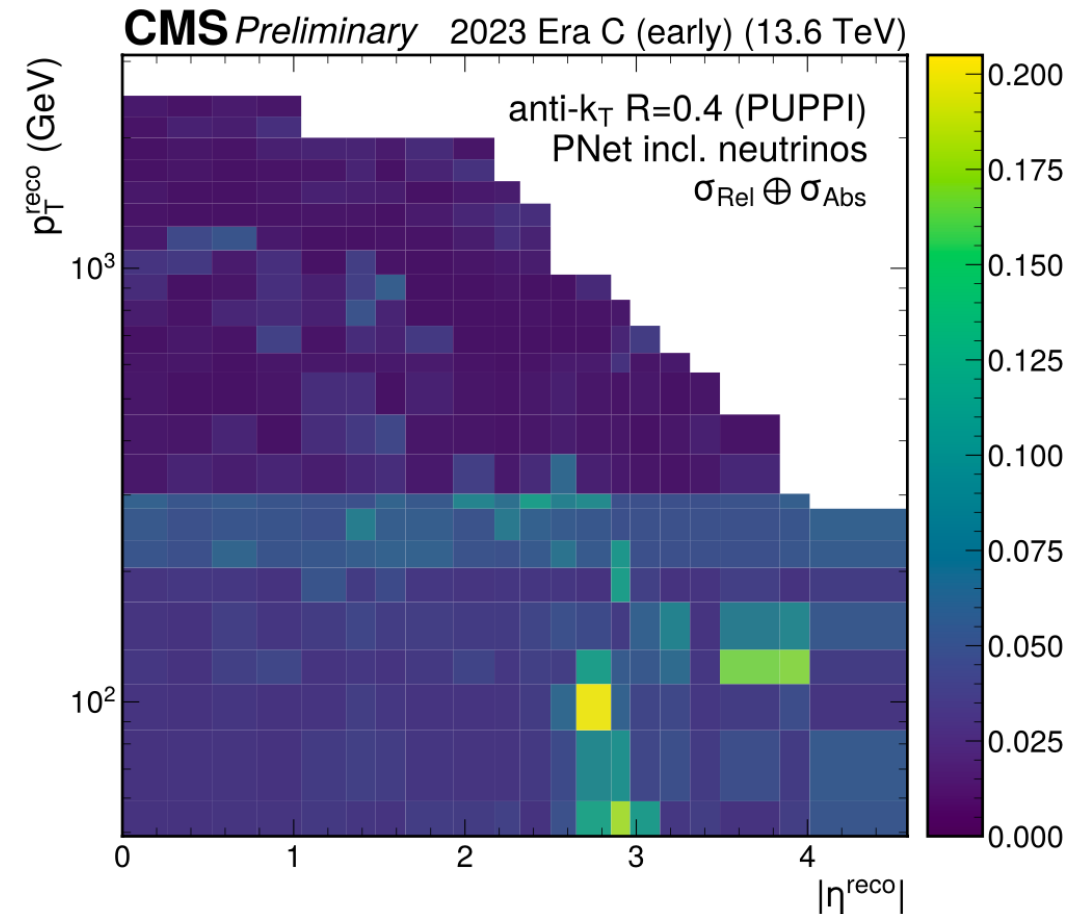
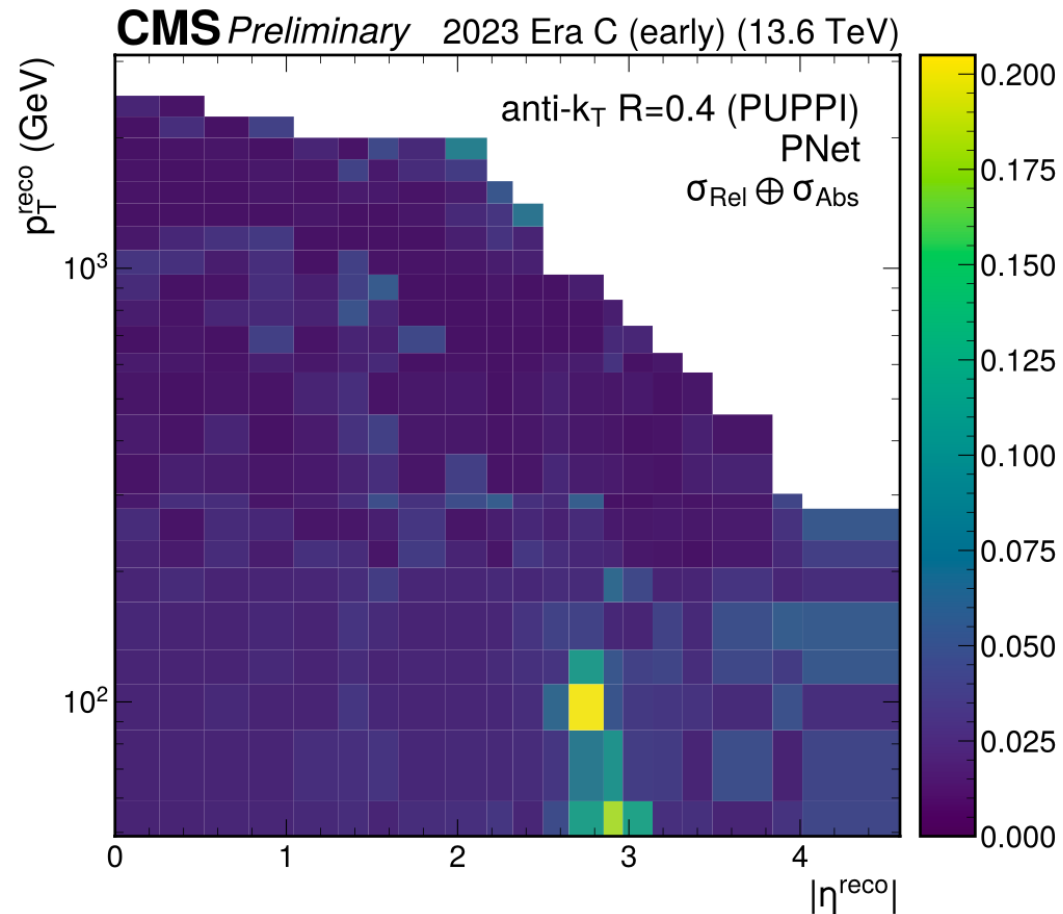
**Excellent Precision for Top measurements!**



# Backup

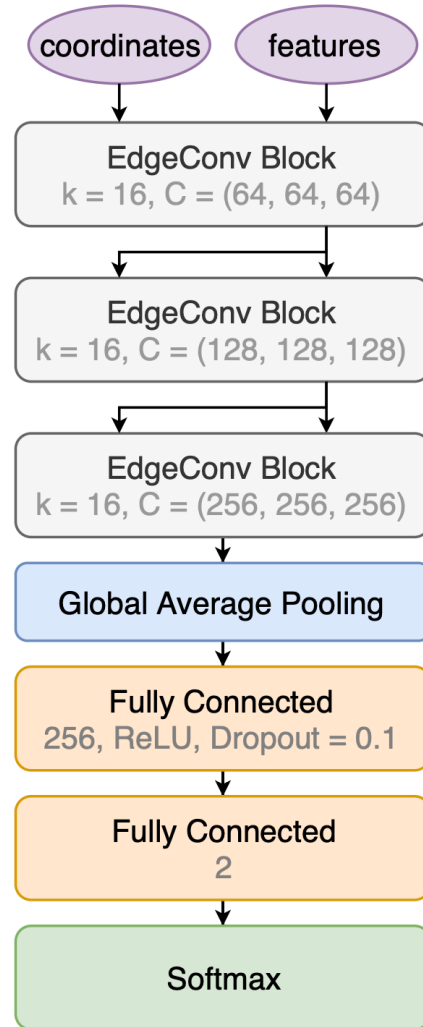
## Residual corrections closure

- Residual corrections are not recomputed for regressed  $p_T$  jets, but the standard ones are applied
- The level of non-closure is approximately up to 2-5% (2-8%) for PNet regressed  $p_T$  without (with) neutrinos in the barrel and higher in the endcap (up to 10-20%)

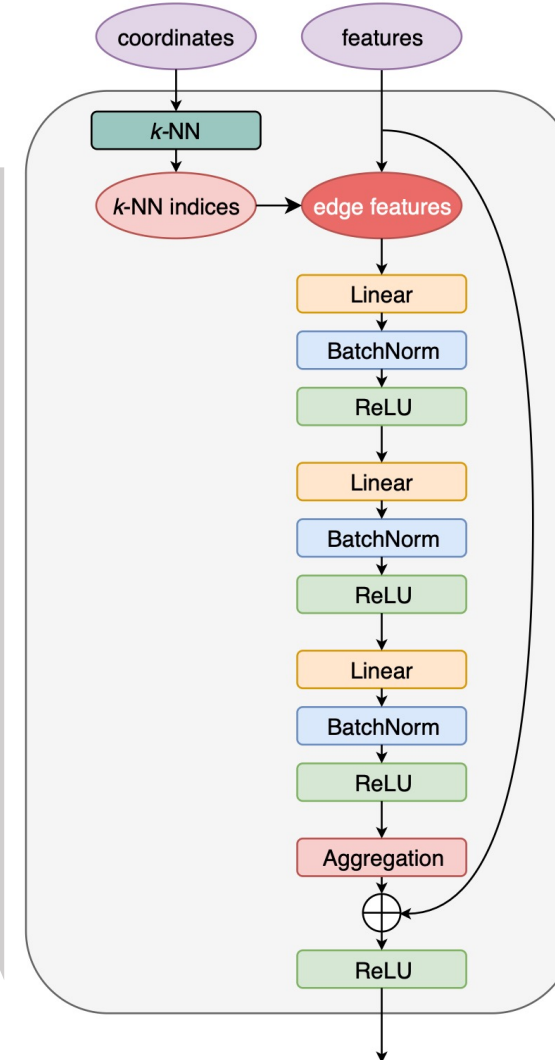


# ML tagger Architectures

## ParticleNet



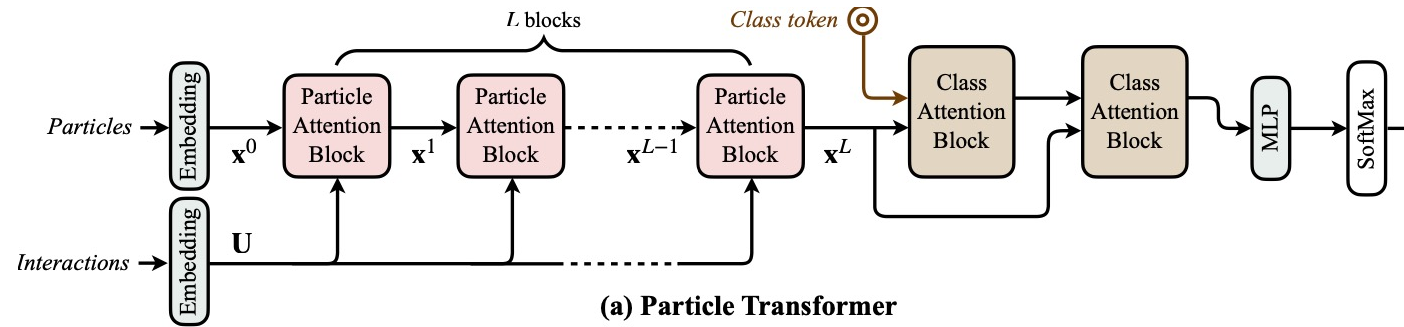
Architectures of the ParticleNet



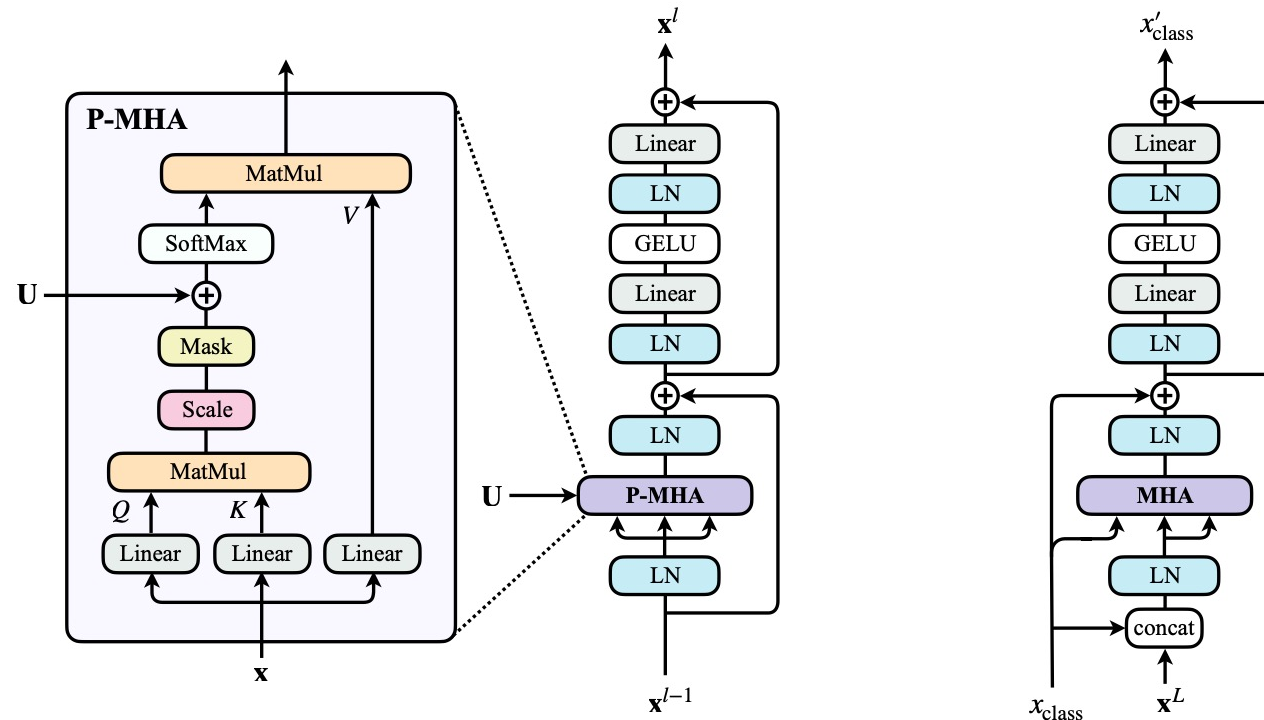
The structure of the EdgeConv block

# ML tagger Architectures

## Particle Transformer



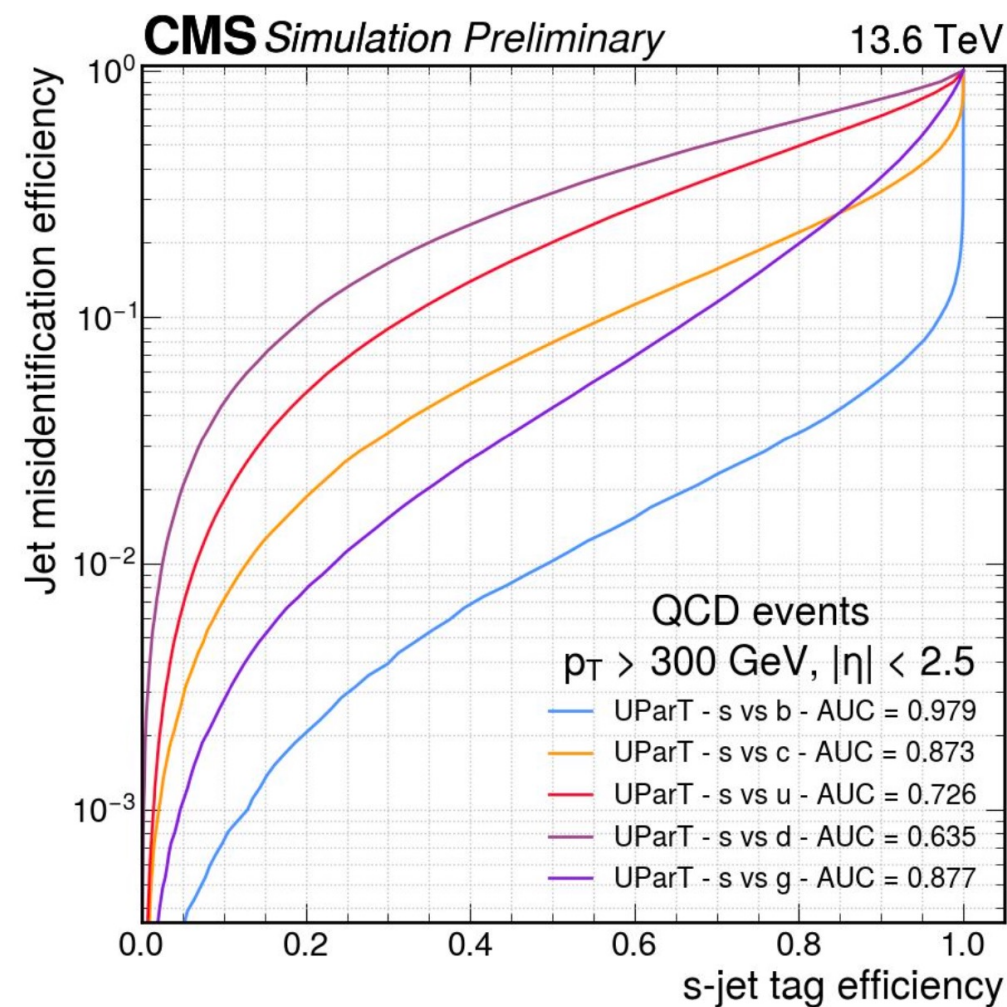
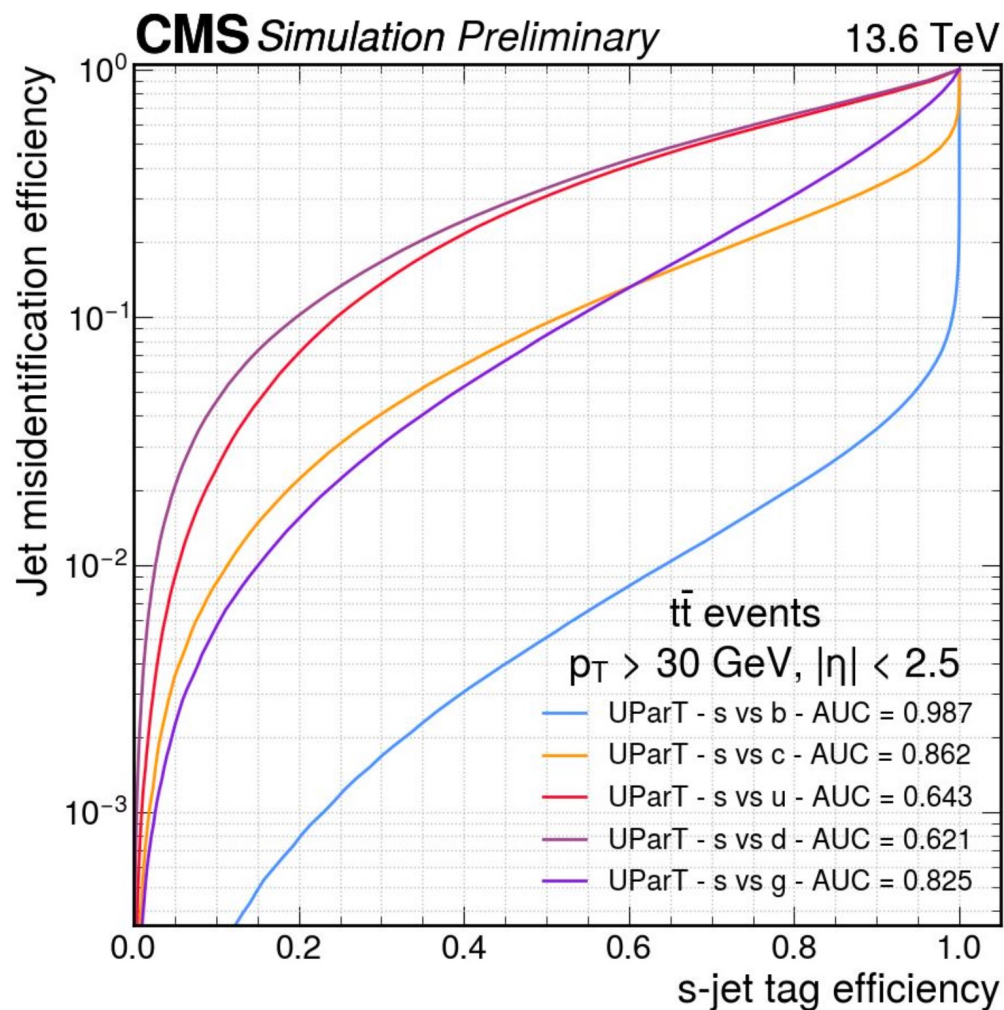
(a) Particle Transformer



(b) Particle Attention Block

(c) Class Attention Block

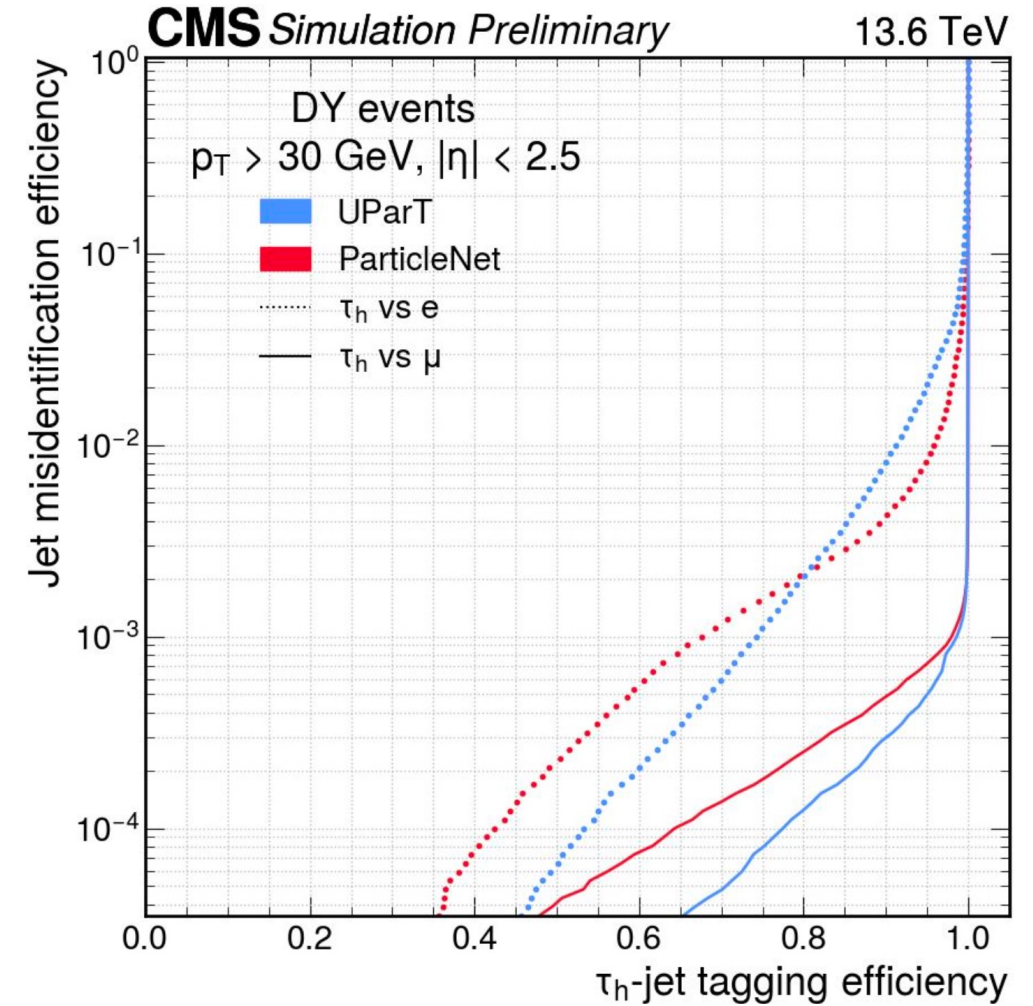
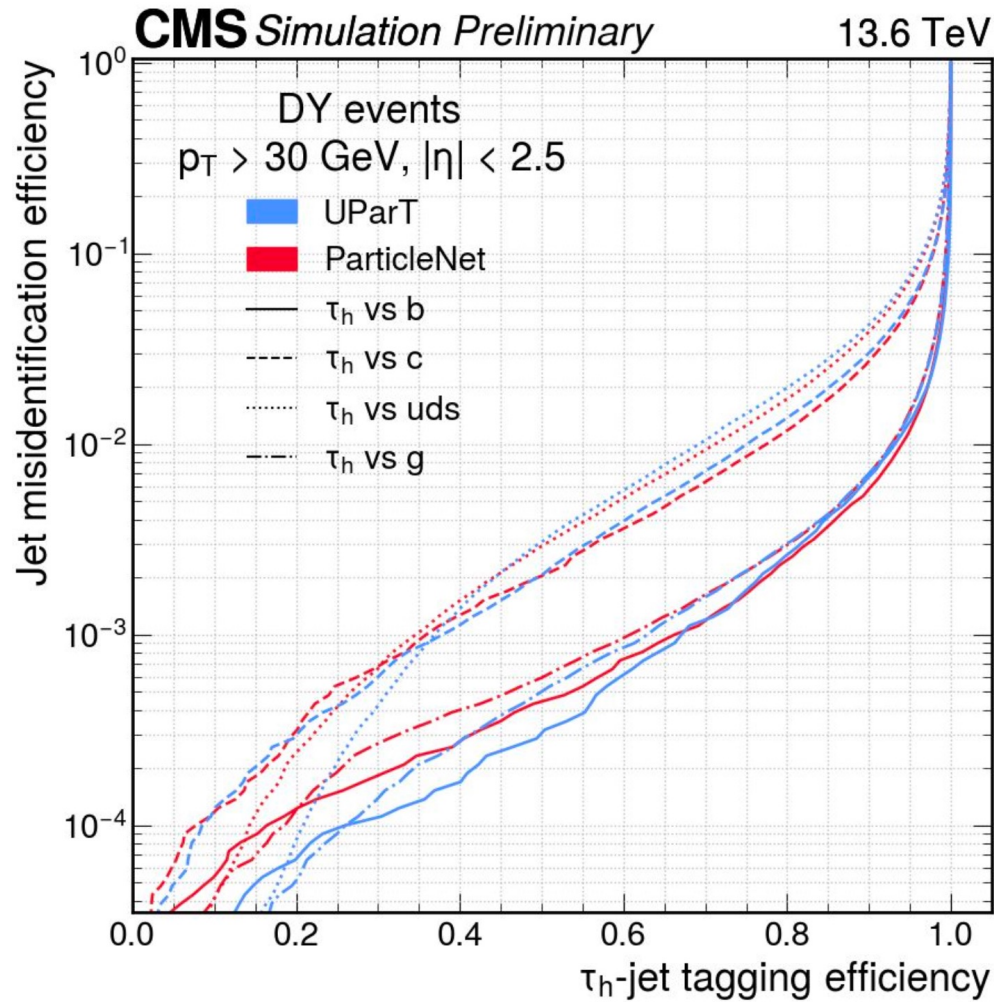
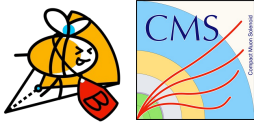
# s-tagging Performance



- The first time a specific s-node added to jet tagging algorithm
- Performance indicates we can achieve a low efficiency s-tagger

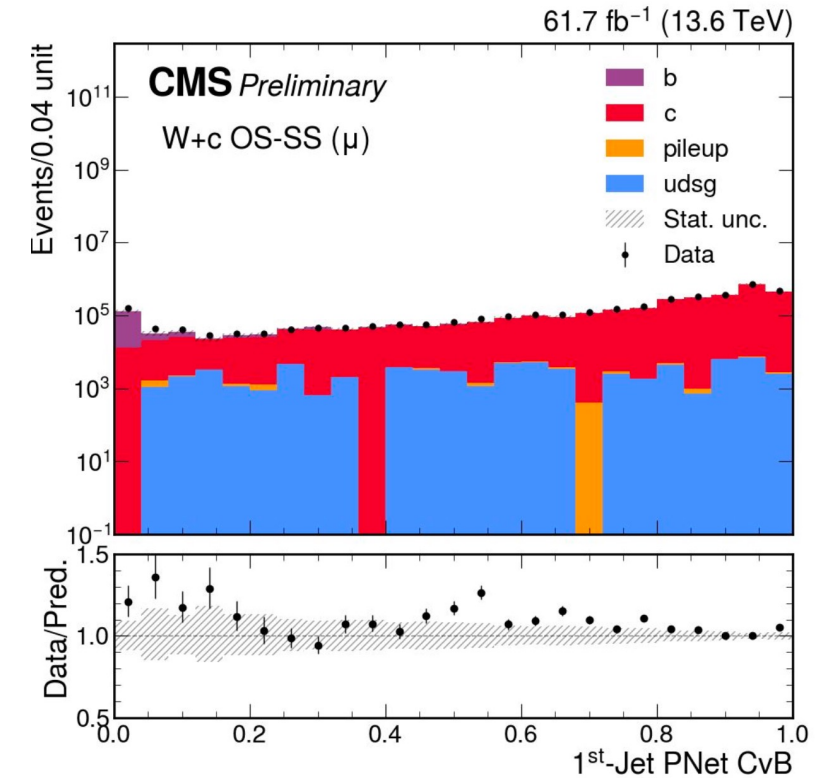
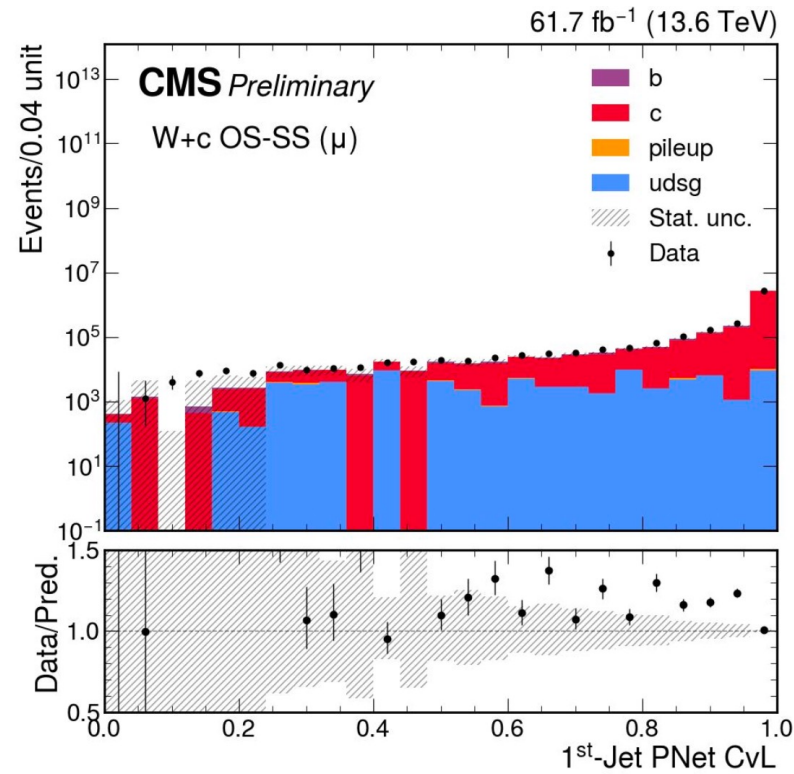
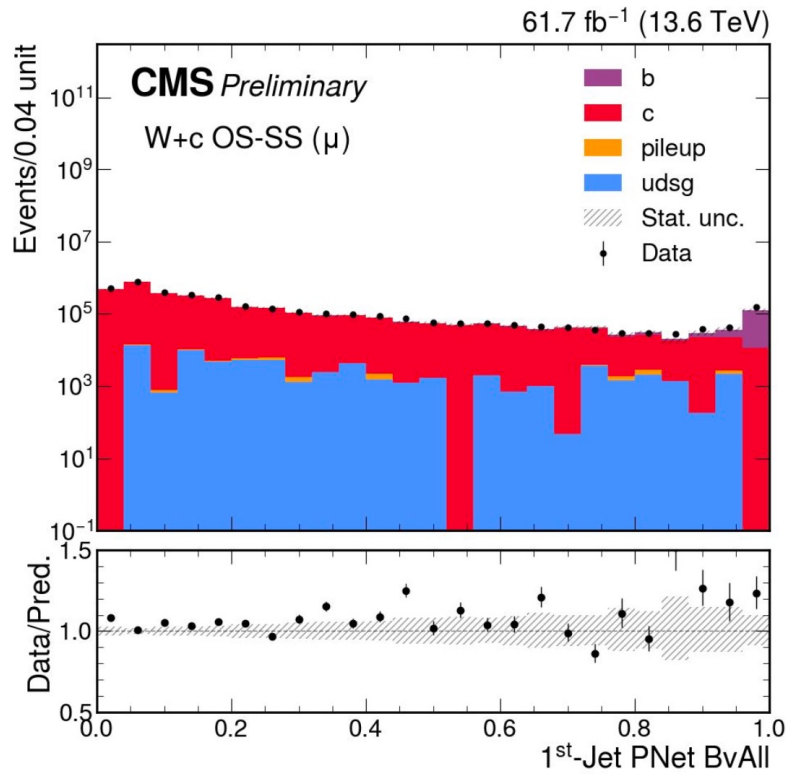
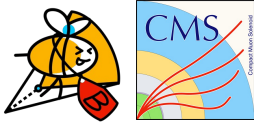


# $\tau_h$ -tagging Performance



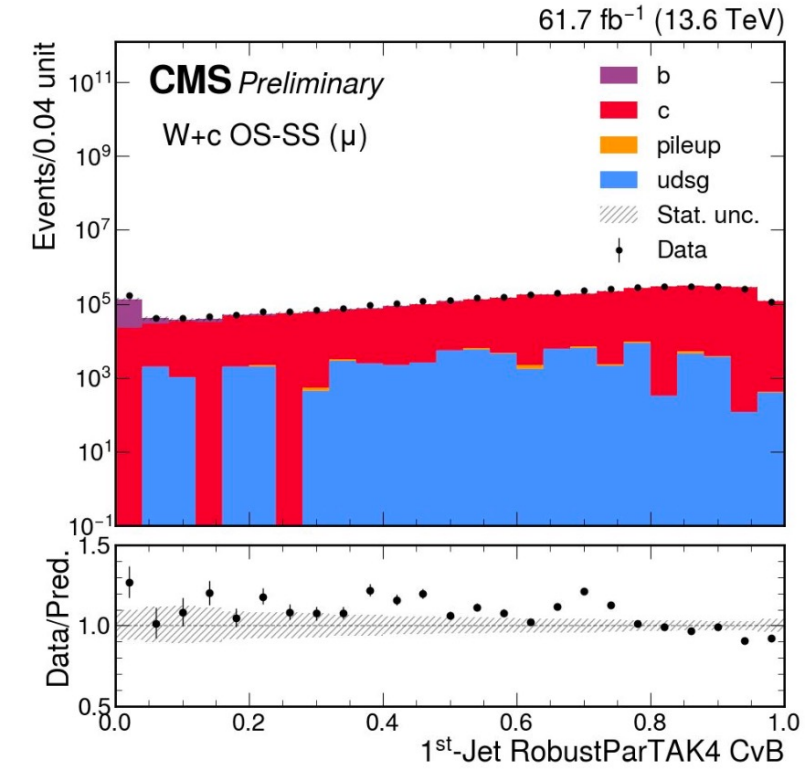
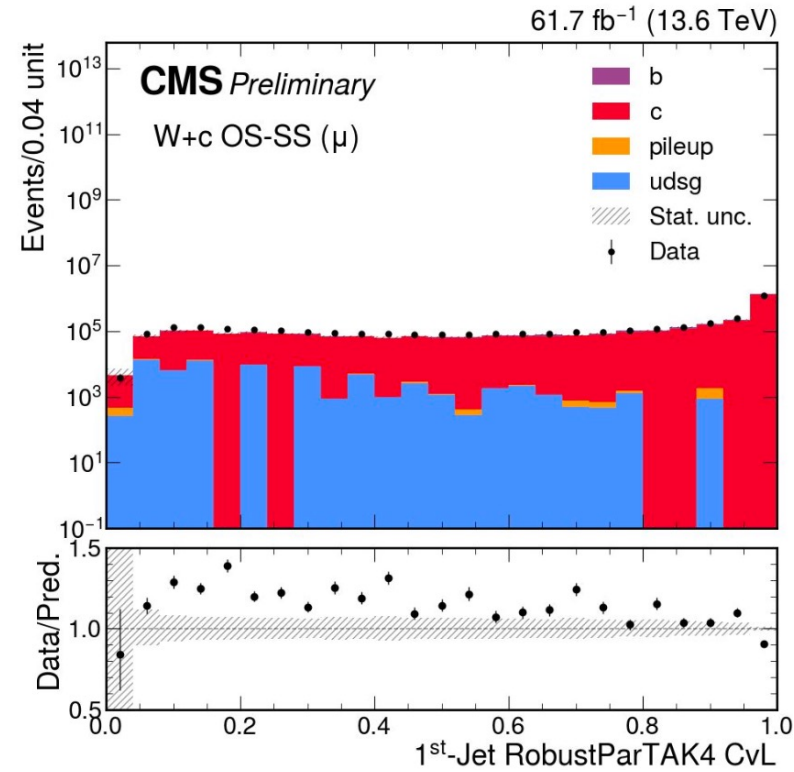
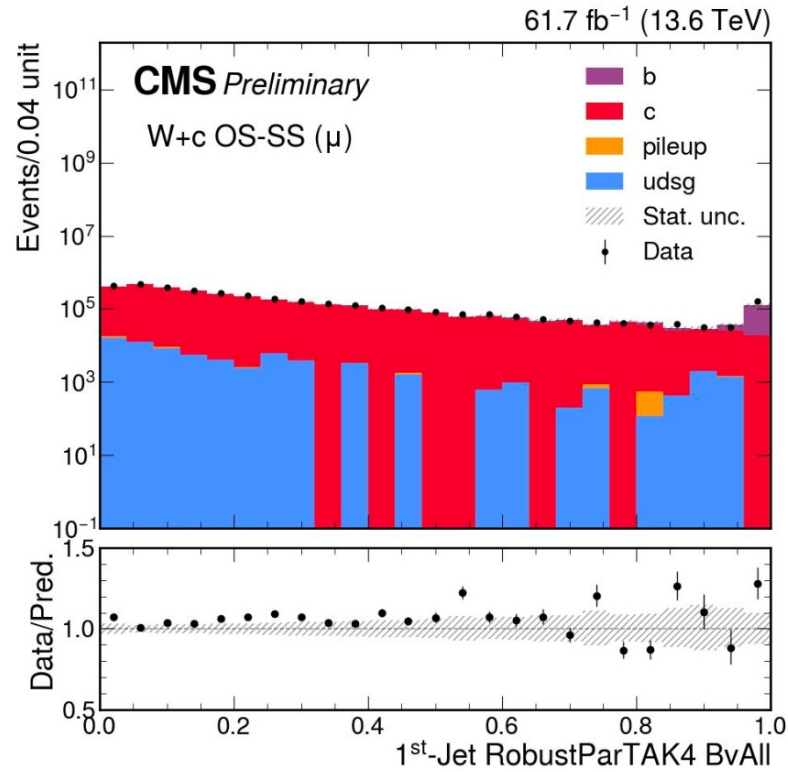
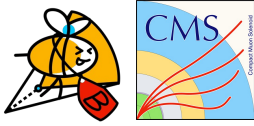
- ParticleNet and UParT show similar performances
- ParticleNet performs better at high misidentification rate and UParT at lower rate

# Commissioning in c-tagging



- Performed in W+c phase space

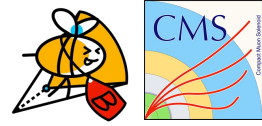
# Commissioning in c-tagging



- Performed in W+c phase space

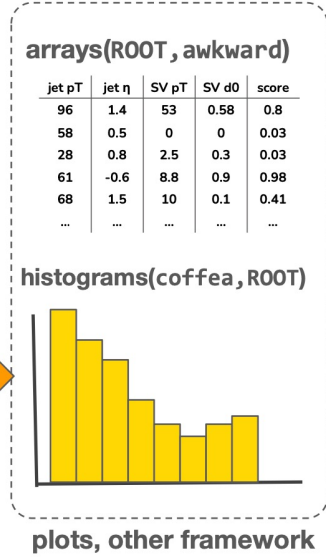
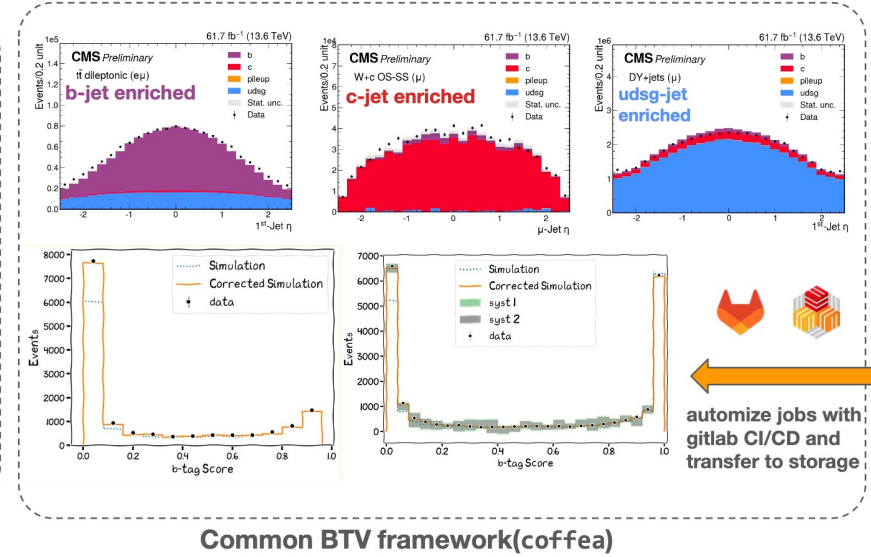
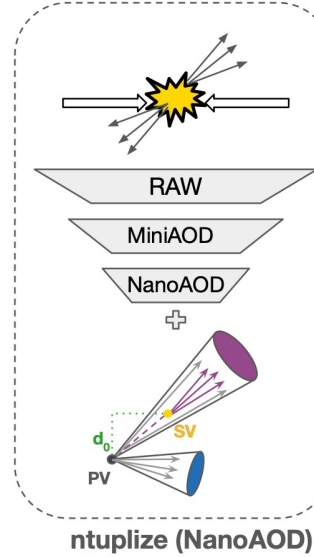


# FTAG Frameworks



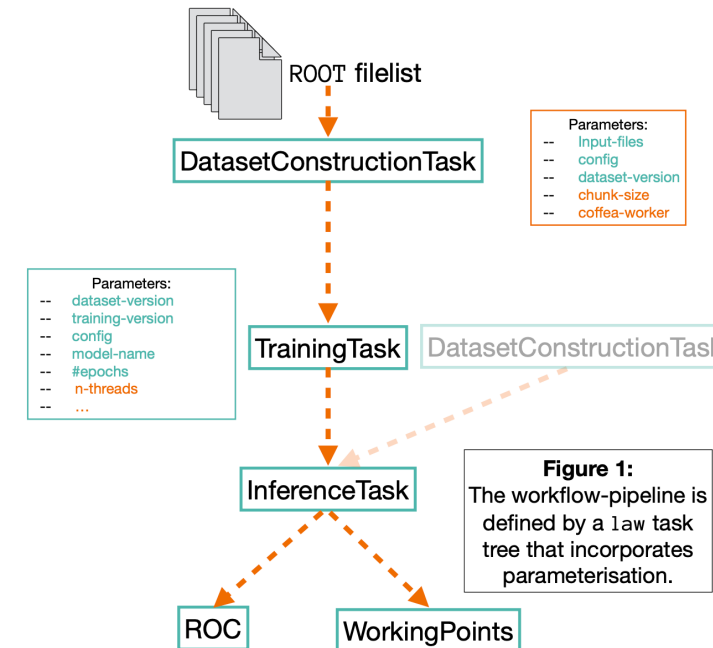
## Commissioning workflows

- BTVNanoCommissioning
  - Fast and efficient
  - Python array based manipulation instead of loops
  - Automated using Gitlab Continuous integration
  - Monitor performance in regular intervals



## Training

- b-hive
  - Framework dedicated for training
  - Easily customizable to introduce your own model



**Figure 1:** The workflow-pipeline is defined by a 1aw task tree that incorporates parameterisation.