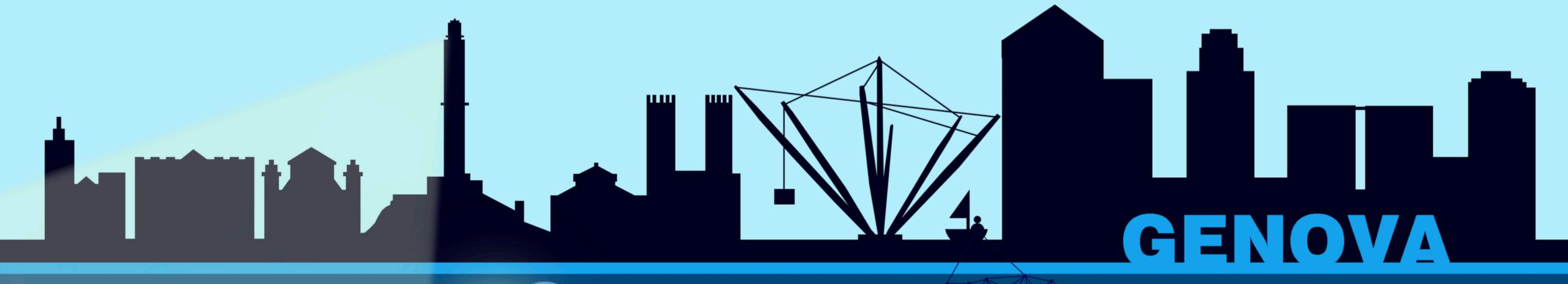
BOOST Experimental SummaryGenova 2024



Jennifer Roloff









DISCLAIMERS

We've come so far, and BOOST is everywhere

Zooming in

It's all about the details

Making the most of it

Using everything we have

Understanding the space

There's so much more to explore

We've come so far, and BOOST is everywhere

Zooming in

It's all about the details

Making the most of it

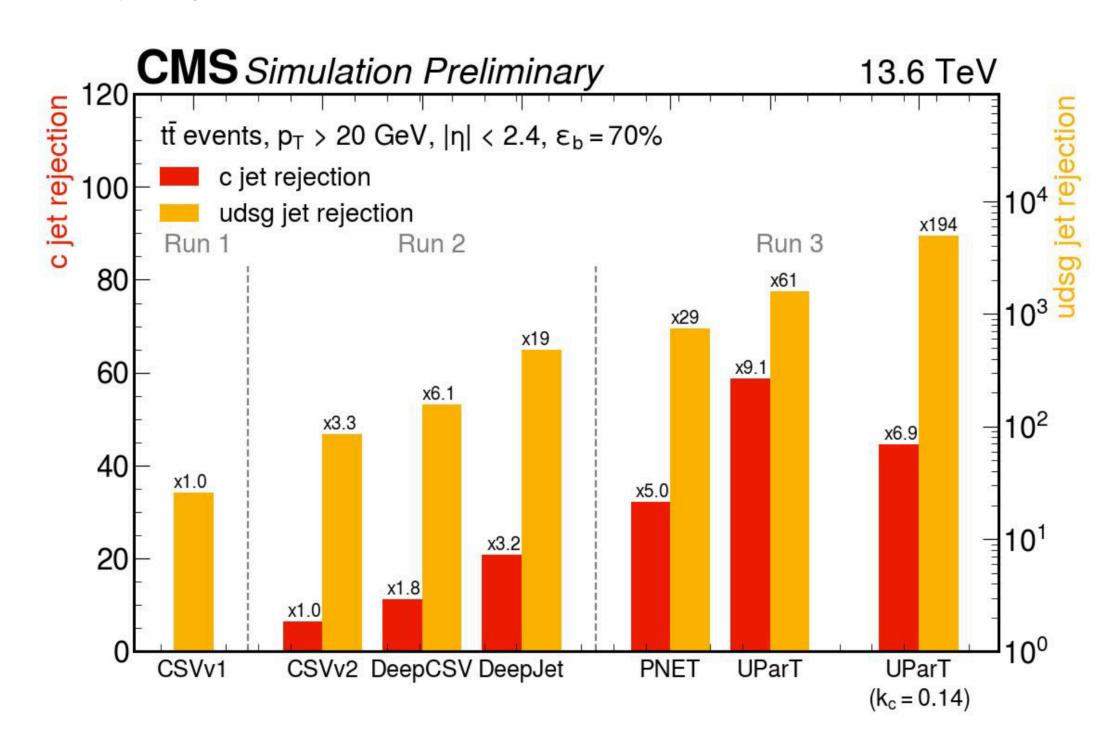
Using everything we have

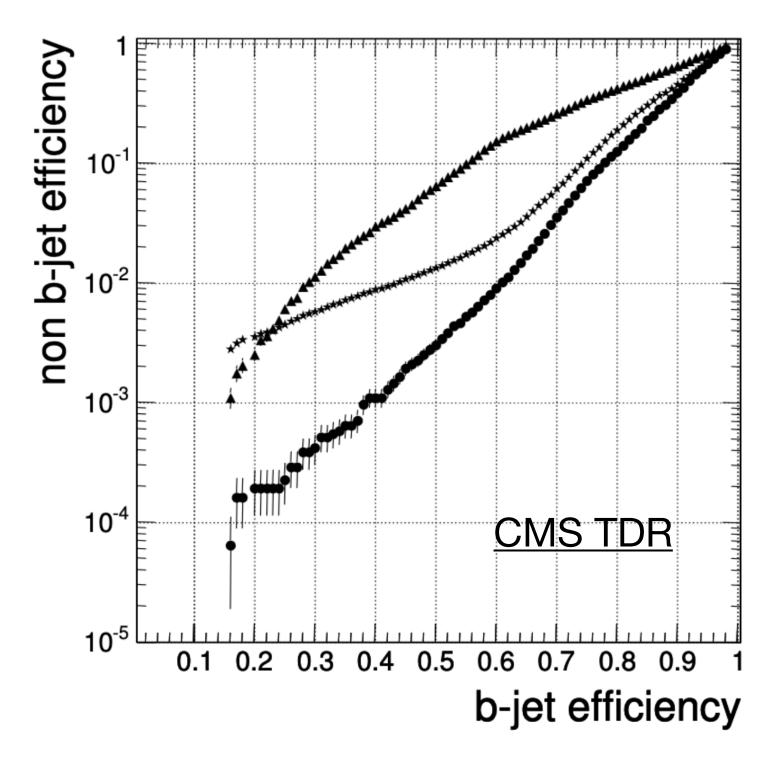
Understanding the space

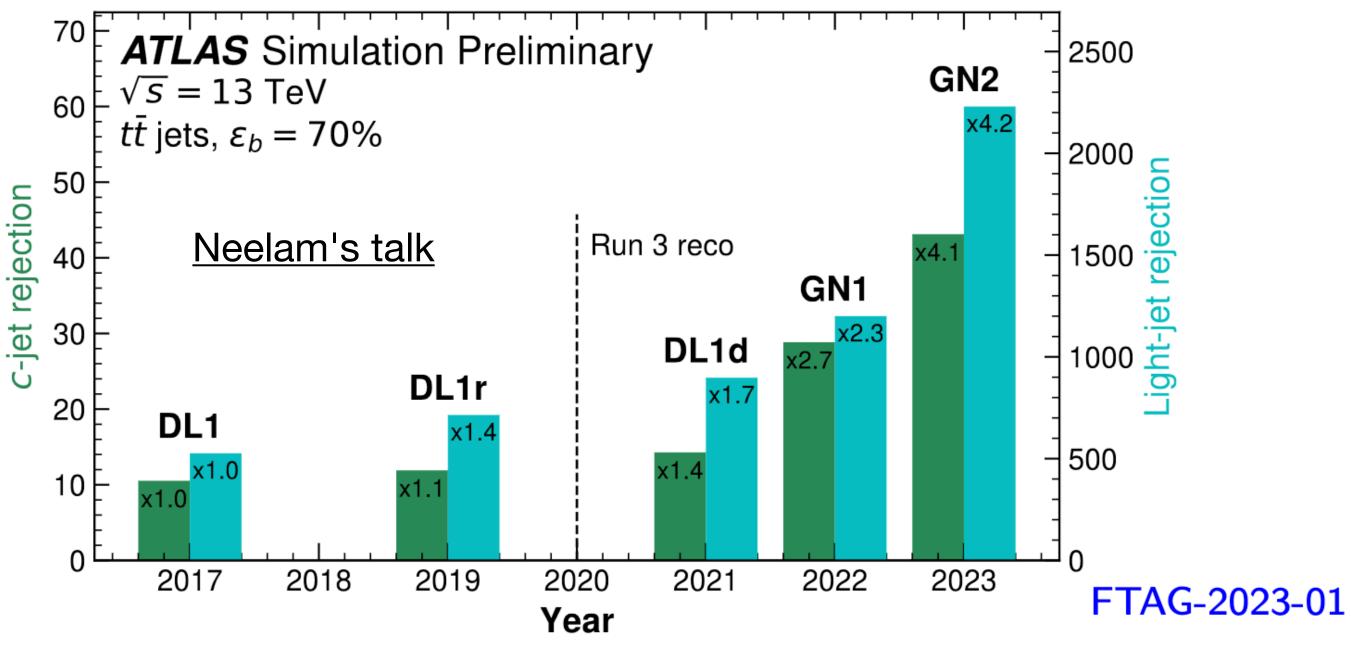
There's so much more to explore

Machine learning is integrated into our conversations

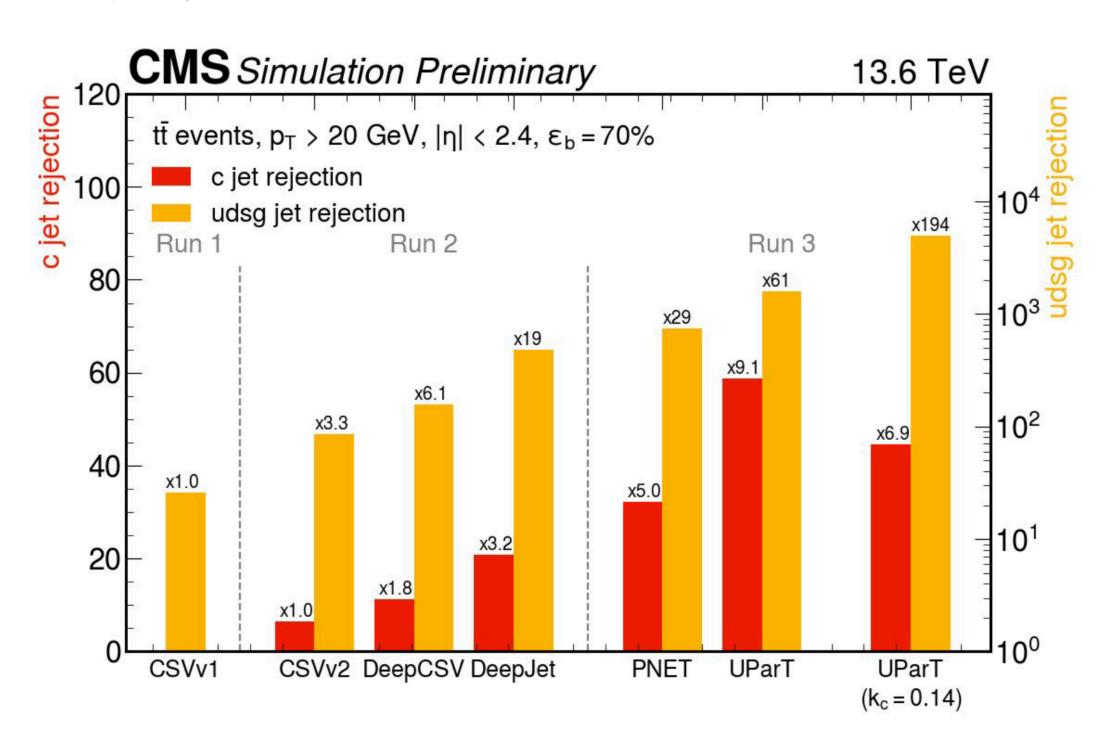
- ATLAS and CMS have made huge gains in their b-jet and c-jet tagging since the start of the LHC
 - ... and even more since the TDR!
- Neural networks are the default, and we are trying to optimize their architectures

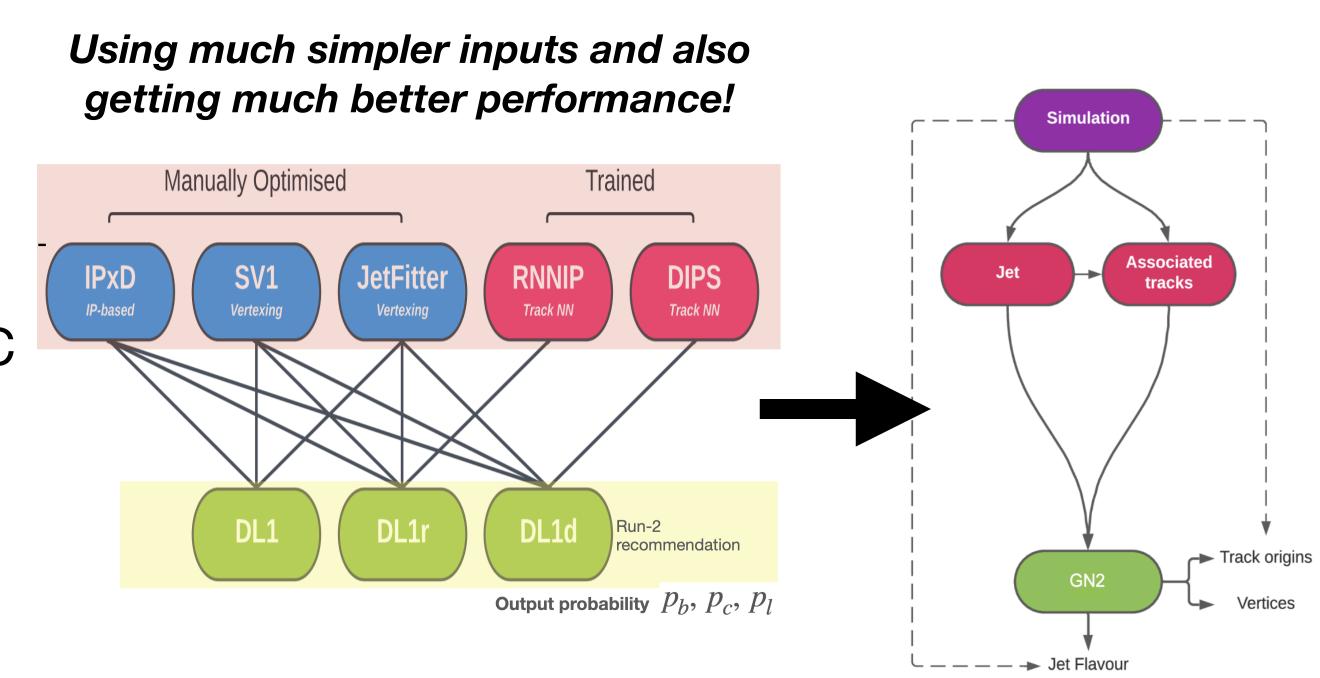


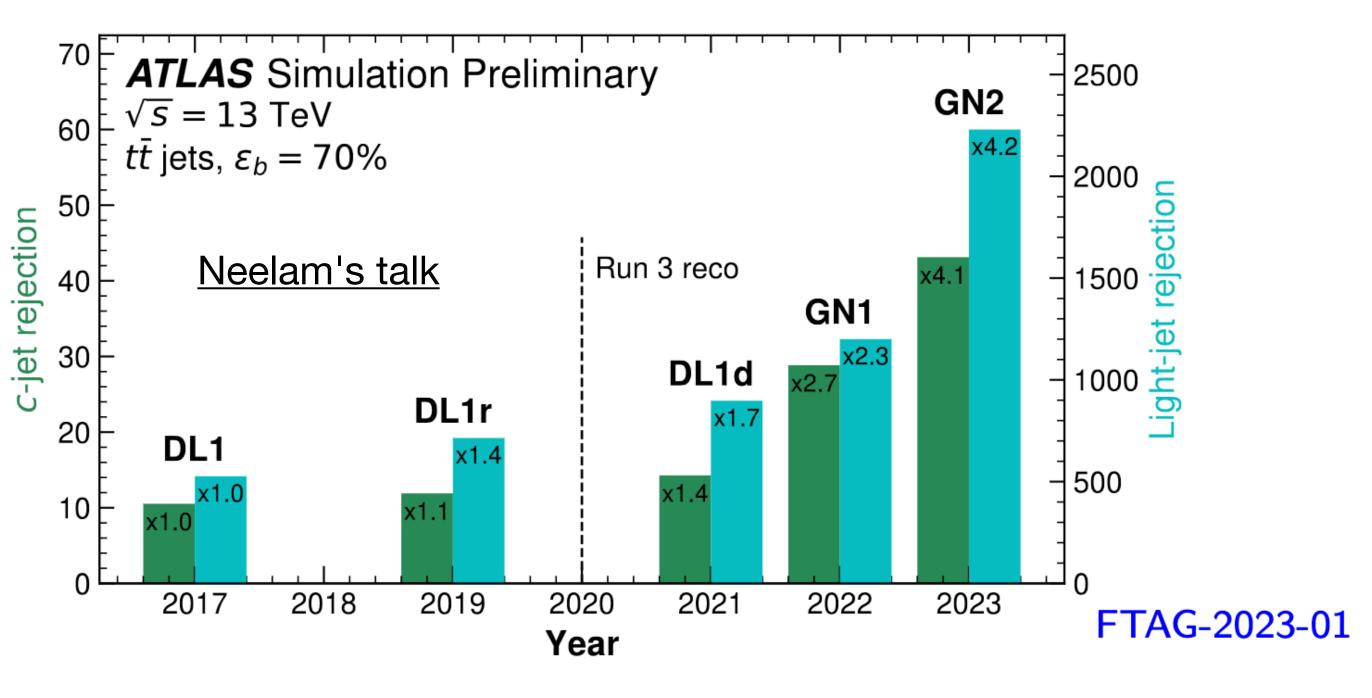


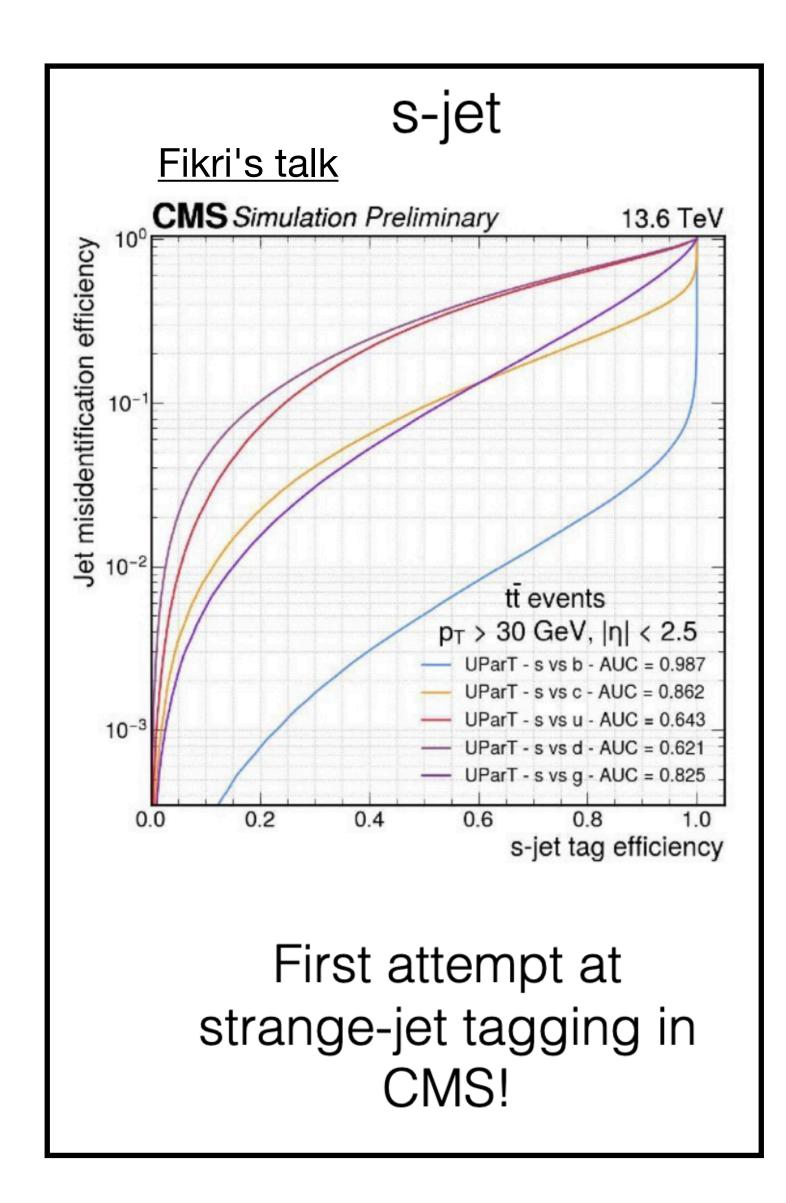


- ATLAS and CMS have made huge gains in their b-jet and c-jet tagging since the start of the LHC
 - ... and even more since the TDR!
- Neural networks are the default, and we are trying to optimize their architectures

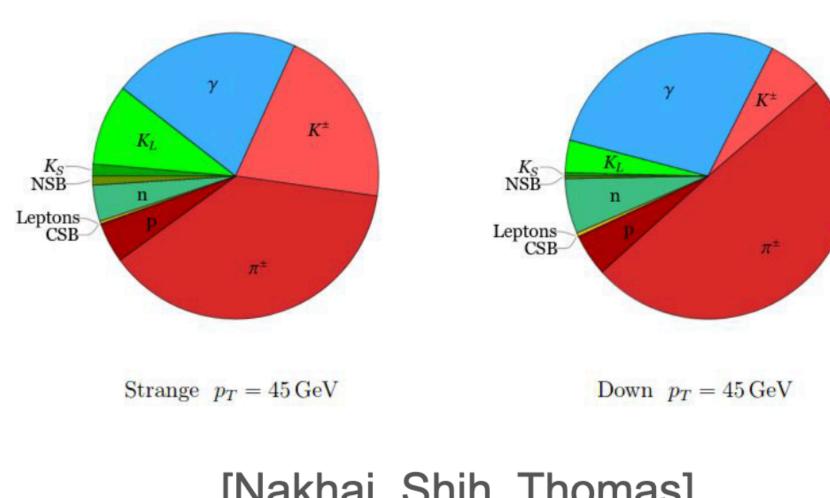








Michele's talk



[Nakhai, Shih, Thomas]

- Heard about potential applications of strange-tagging at future colliders
 - ... and this is already being attempted at the LHC!
- Some handles on the differences in hadron composition, but much harder problem than b- or c-tagging

 π^{\pm}, K^{\pm}

 $K_S \rightarrow \pi^+\pi^-$

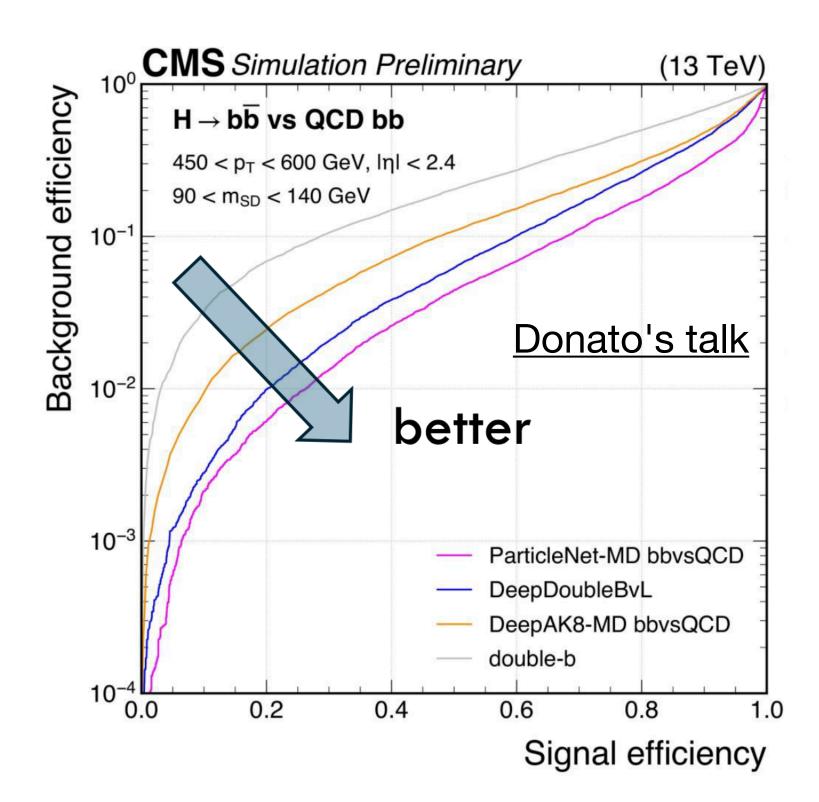
---* KL, KS

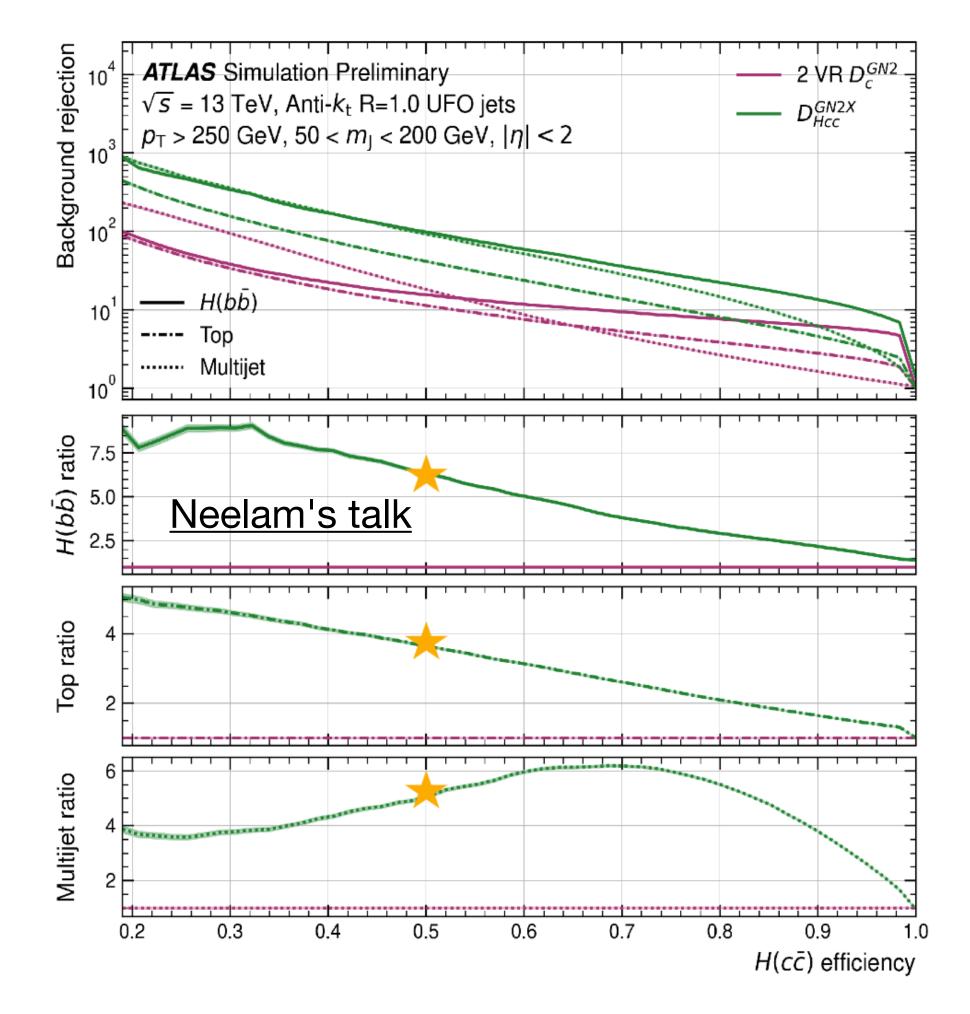
150 cm

ECAL

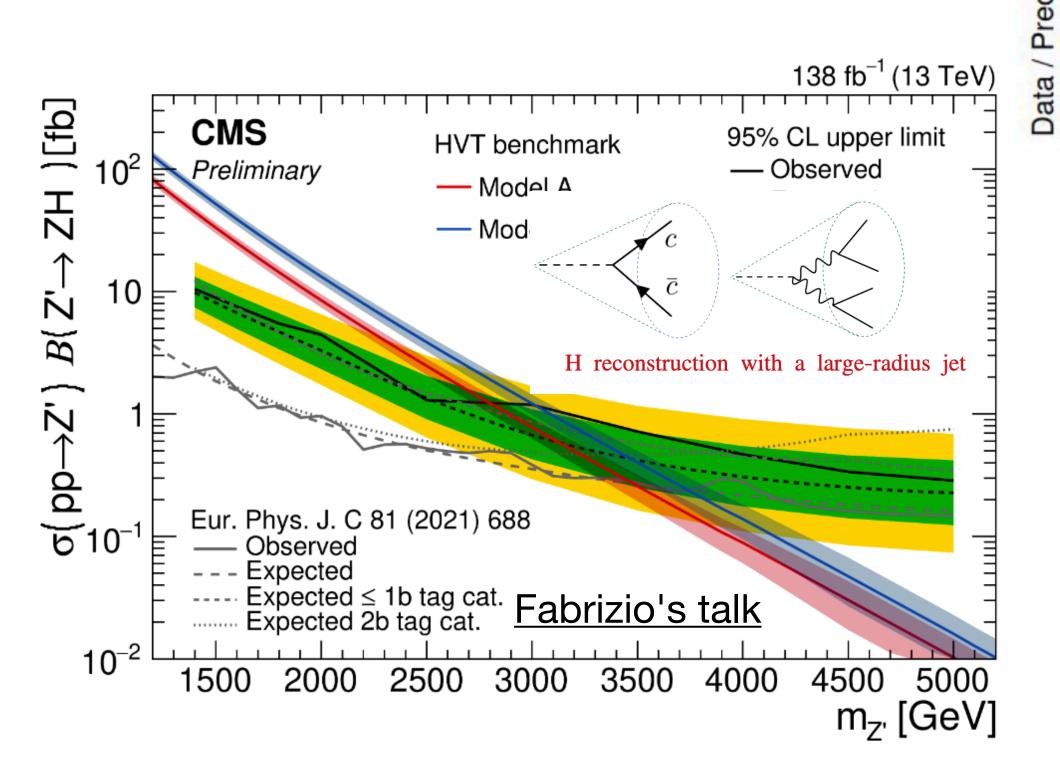
130 cm

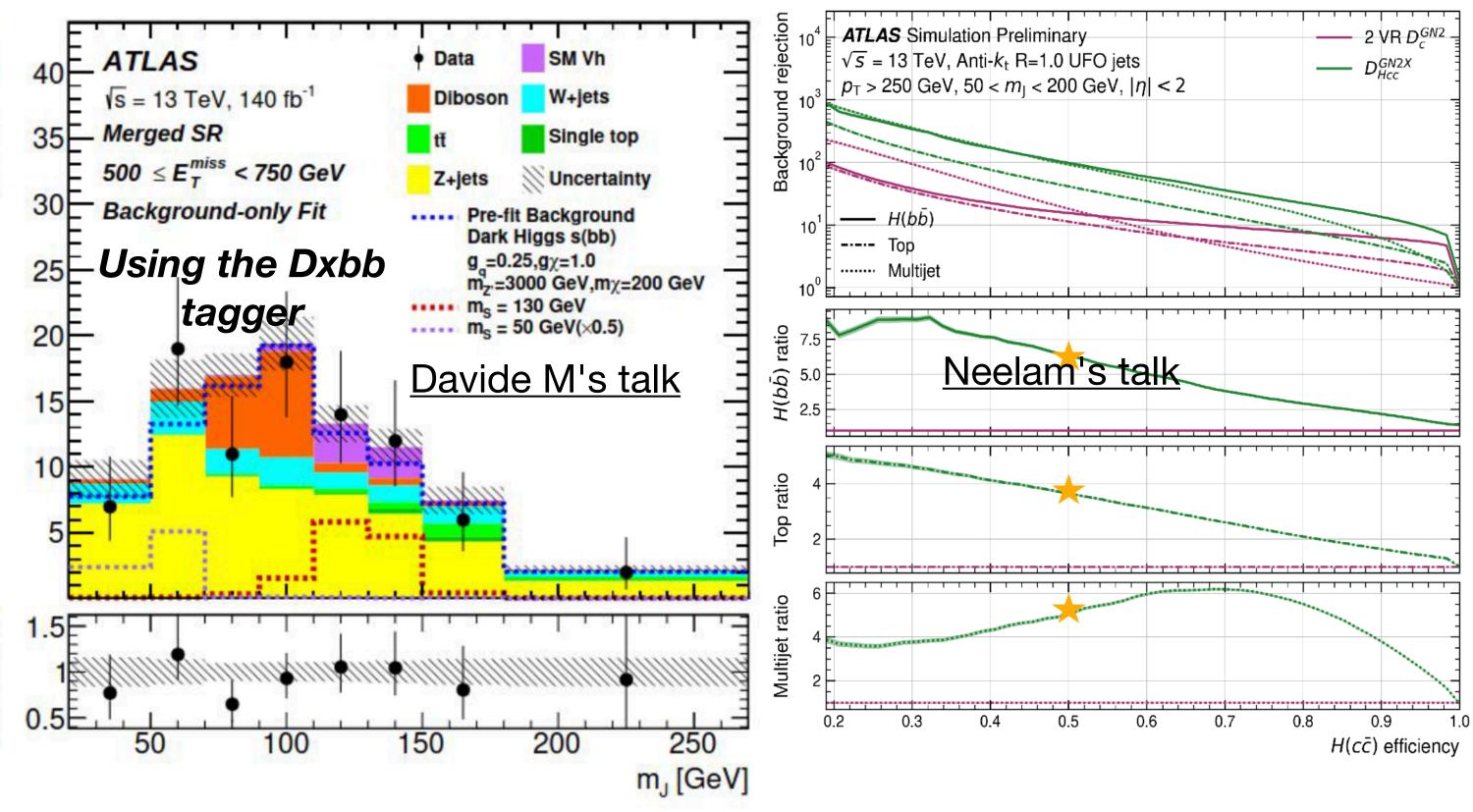
- Double-b and double-c tagging are part of the standard toolkit
 - Still making big gains in the tagging performance, so expect more improvements to come





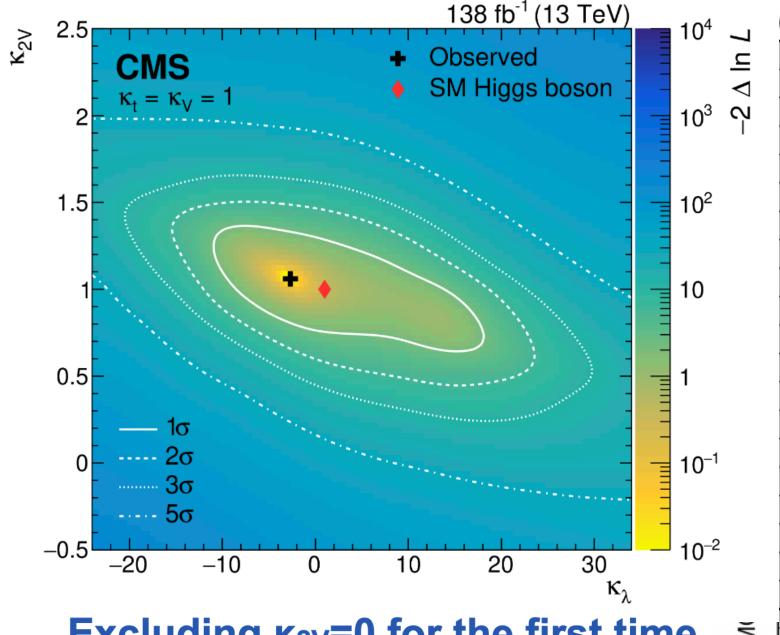
X→bb/cc taggers are used for a wide range of searches



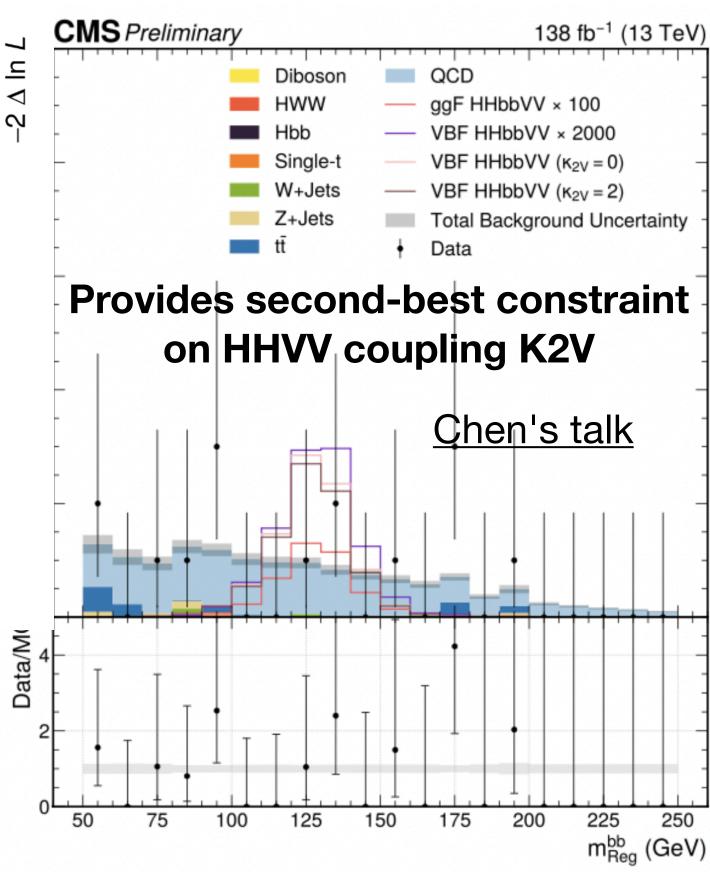


- New taggers take time to optimize and calibrate → can take a few years to go from proof of concept to being used in analysis
 - Can expect more sensitivity in many searches without more data!

Boosted channels are now a staple of the Higgs and di-Higgs physics programs

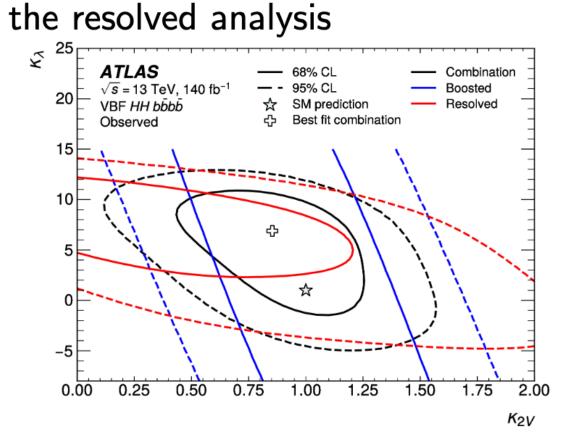


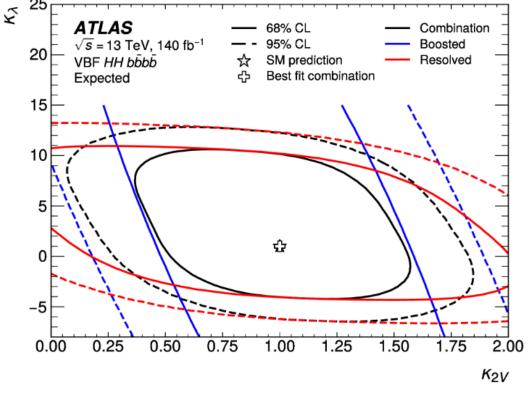




Fabrizio's talk

• Boosted analysis is dominant for κ_{2V} sensitivity while κ_{λ} sensitivity is driven by

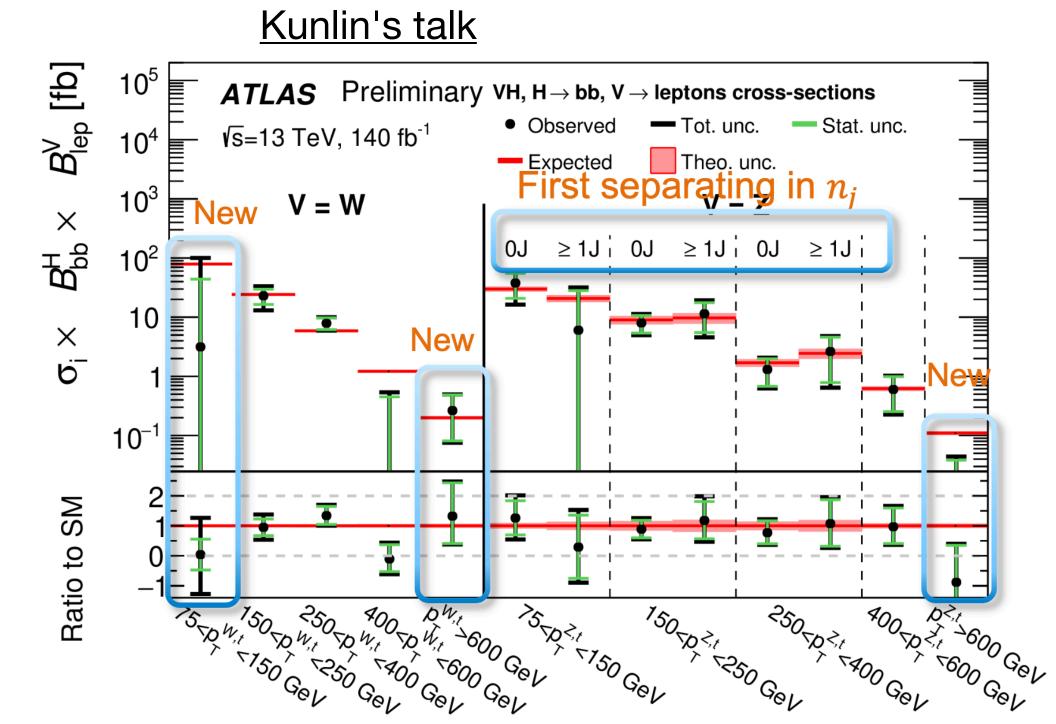


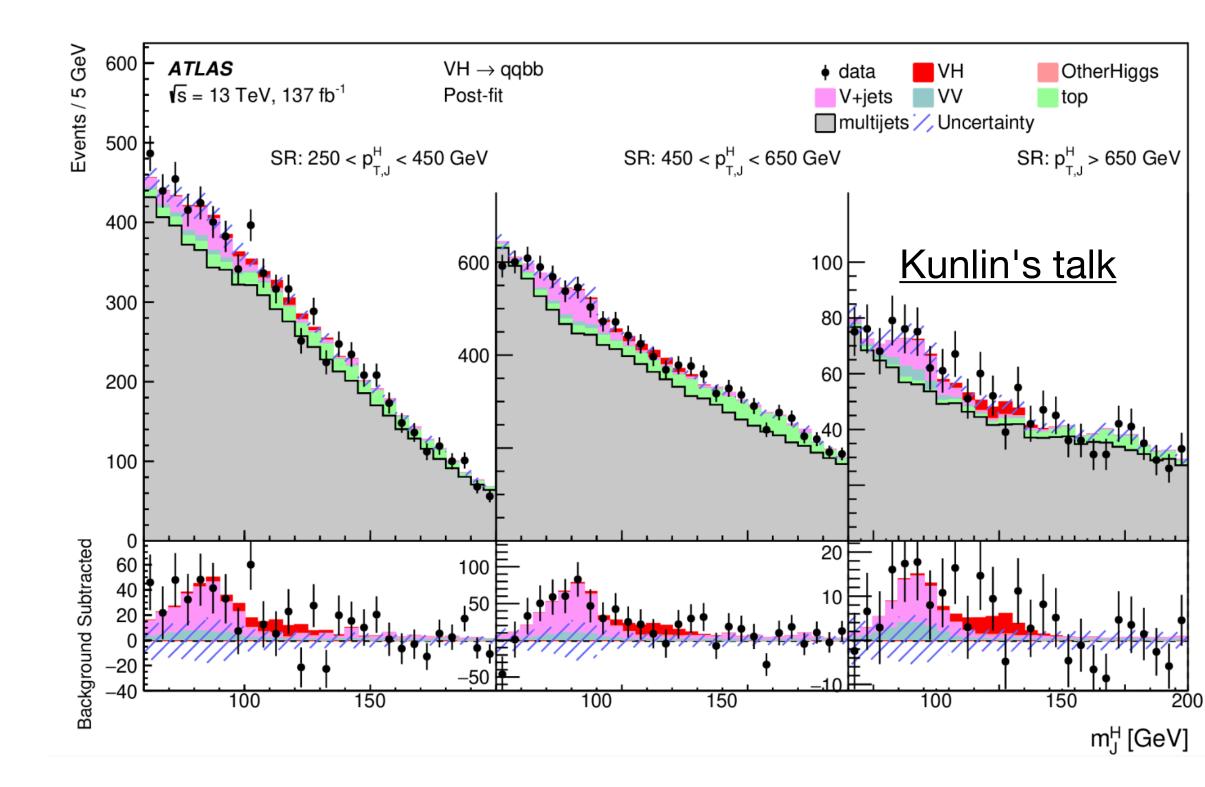


• Results on κ_{2V} analysis are as good as the HL-LHC projections of the previous-best VBF HH analysis (the full Run 2 resolved VBF hh4b).

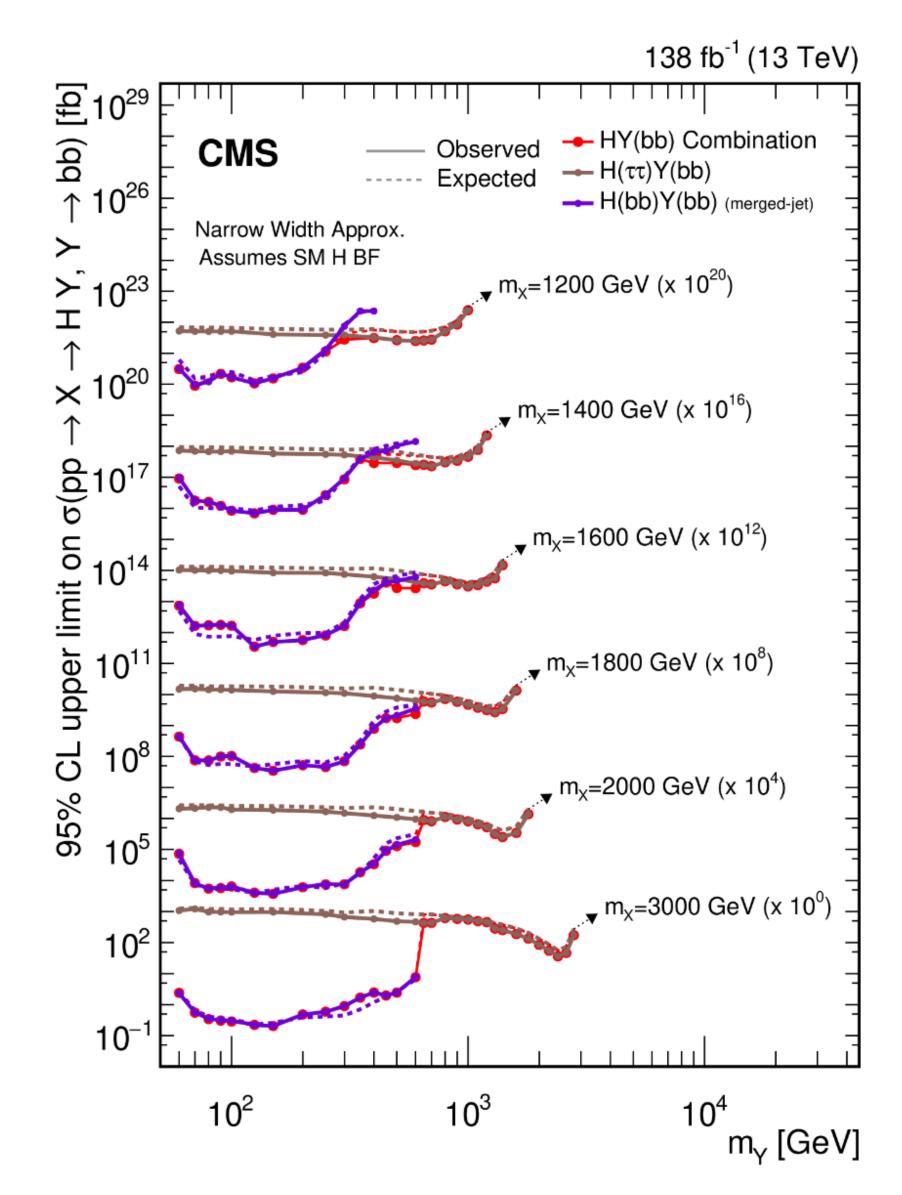
- Rapid tagger improvements mean we are already reaching expectations for the HL-LHC with Run-2 data
 - Expect more improvements to come

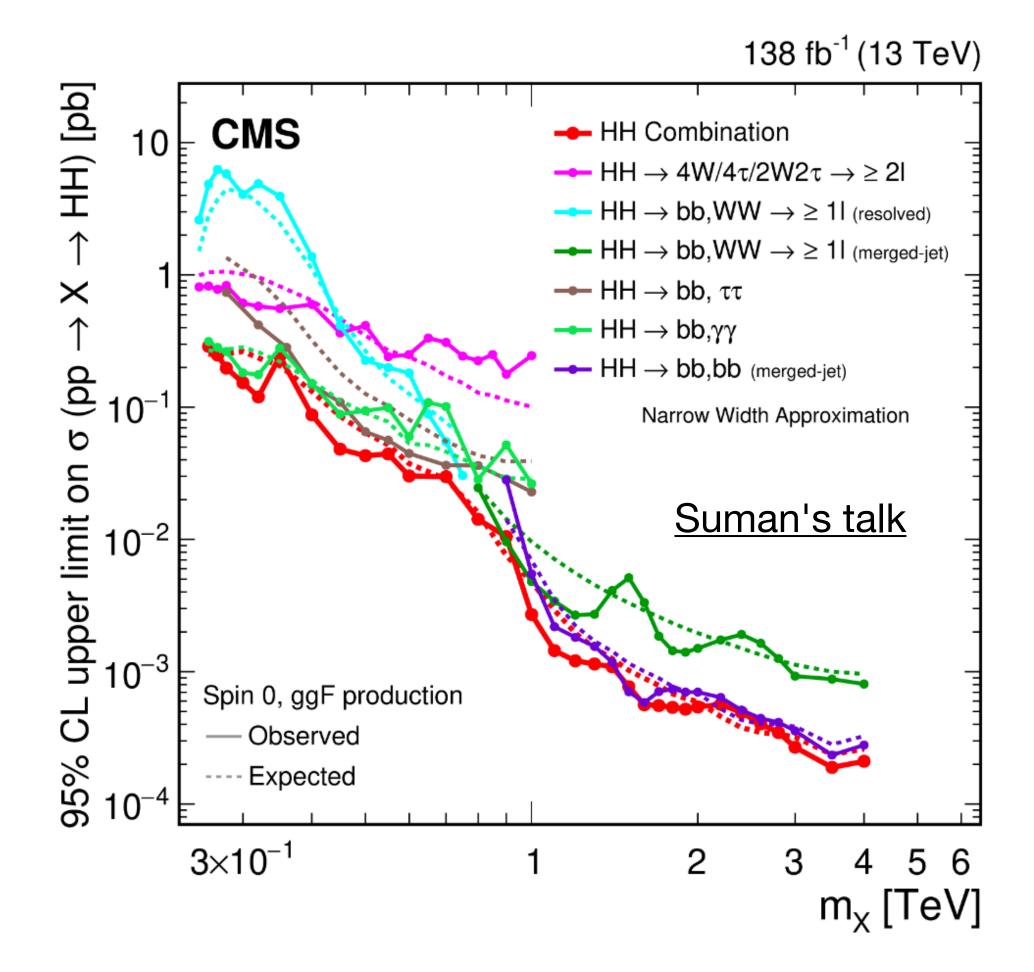
- Starting to use all-hadronic channels to gain access to higher p_T, and with higher statistics (at the cost of higher backgrounds...)
 - Relevant for EFT interpretations, and will benefit from better background modeling and more background reduction





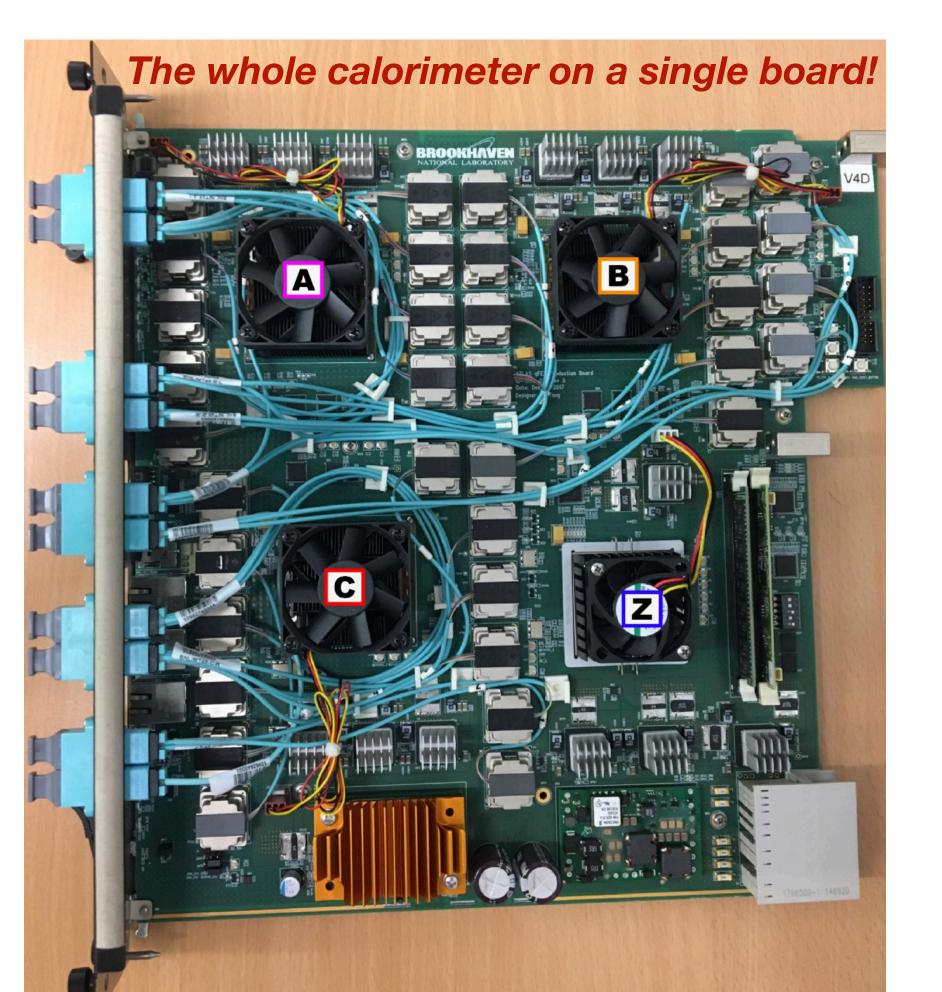
These analyses benefit a lot from the performance work done to optimize these taggers and calibrations!

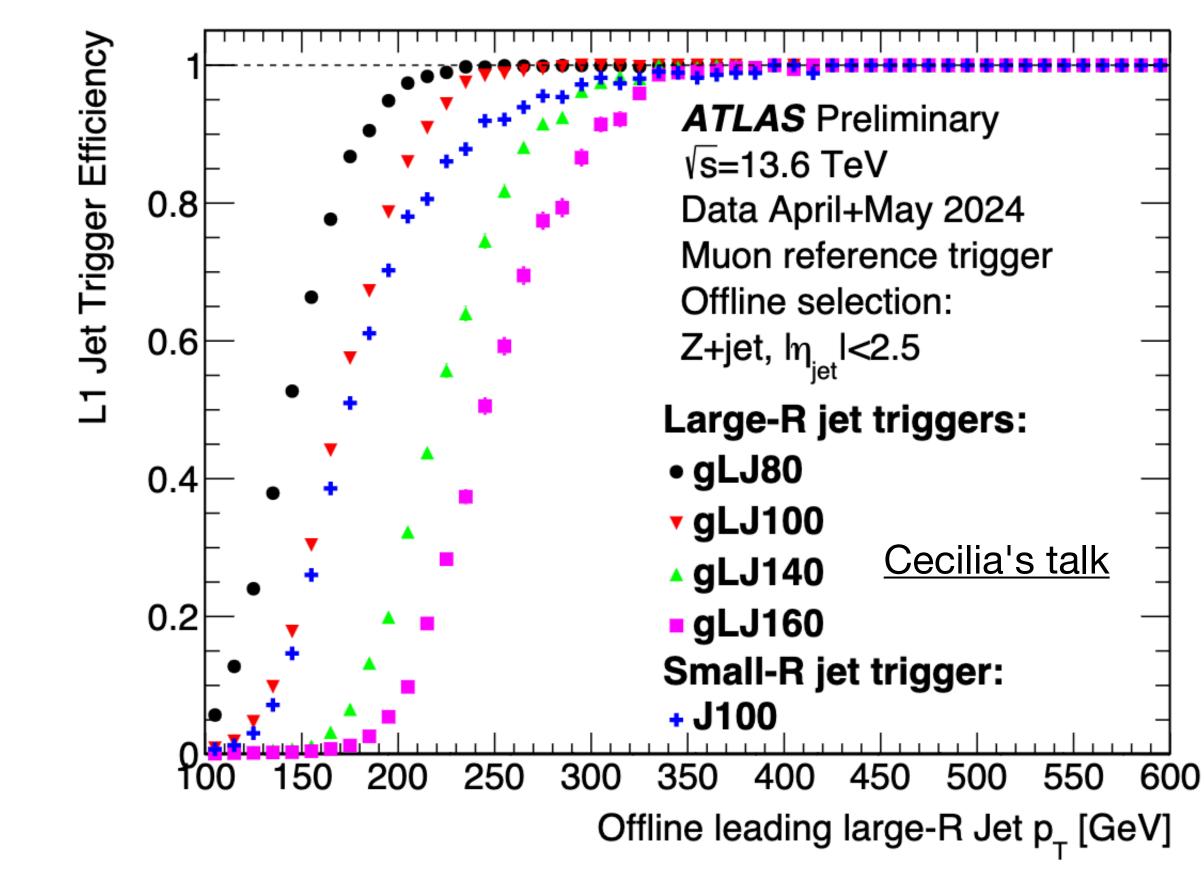




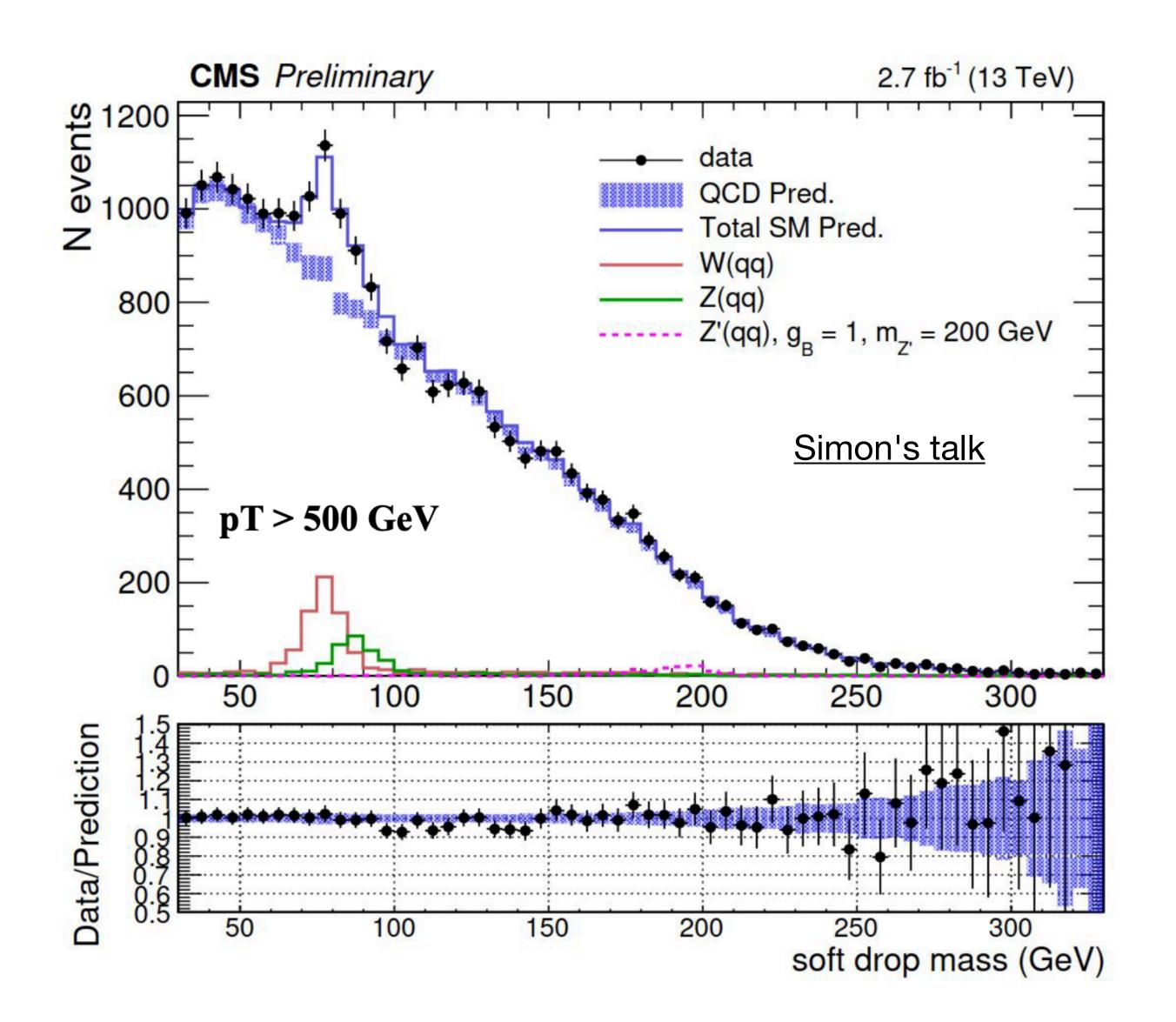
- Boosted jets are used to probe a wide range of models, often bringing more sensitivity than any other channel
 - We are no longer surprised to see searches that use boosted jets as one of the main channels

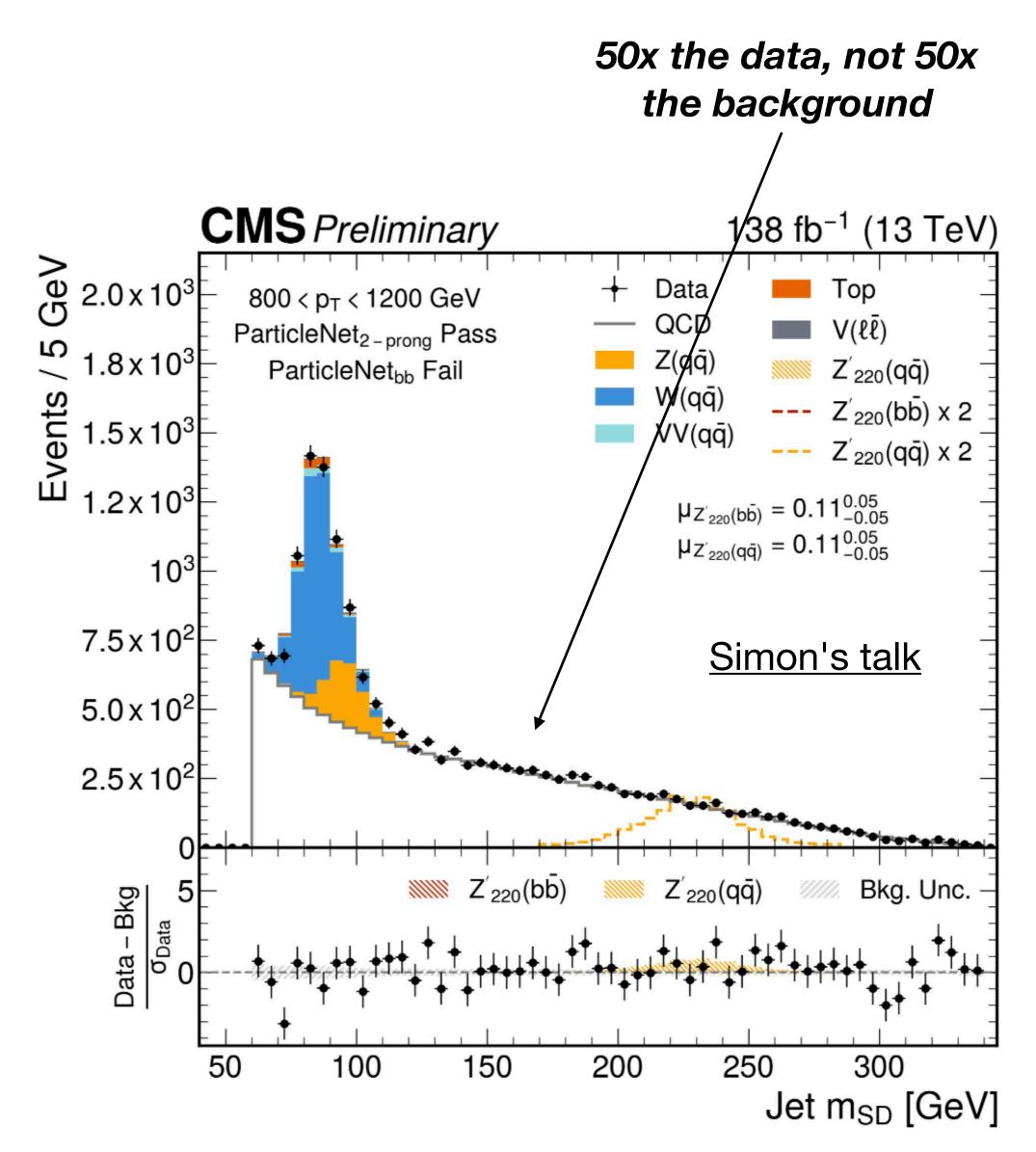
Trigger strategies for single-prong jets do not always apply for multi-prong jets

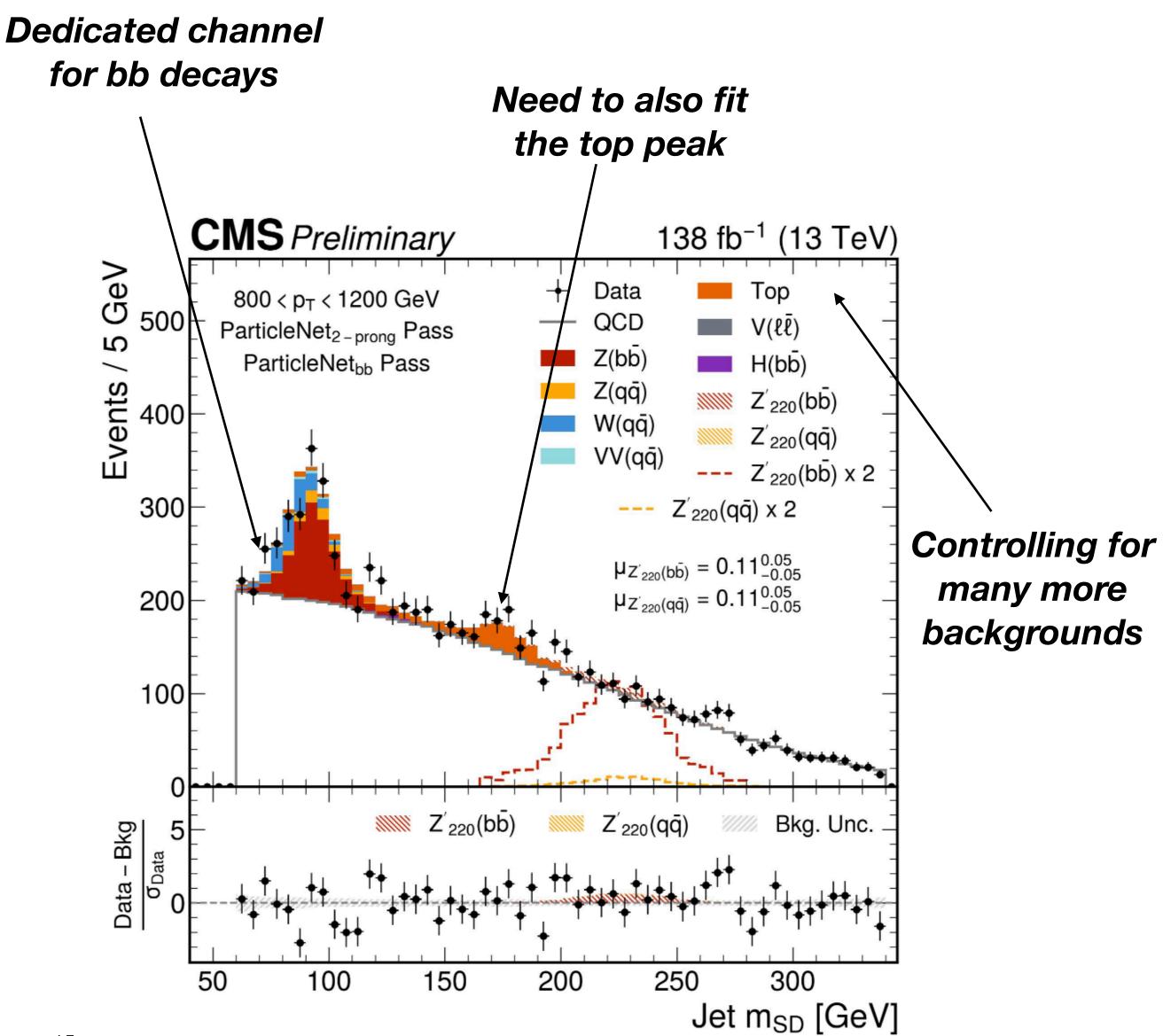




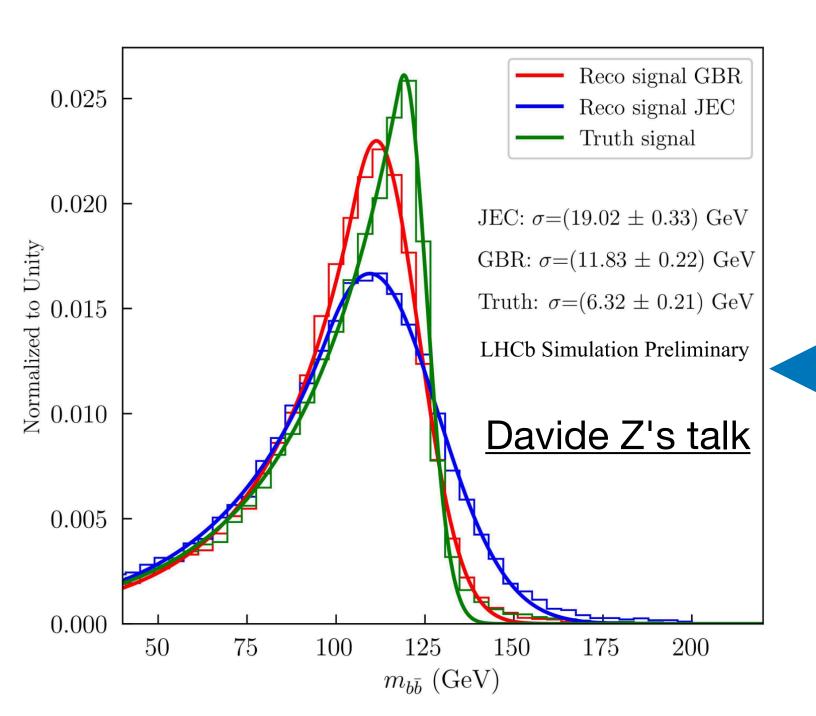
- ATLAS is commissioning a new system that can trigger on multi-prong jets more efficiently
 - Lots of potential for development with new ideas!





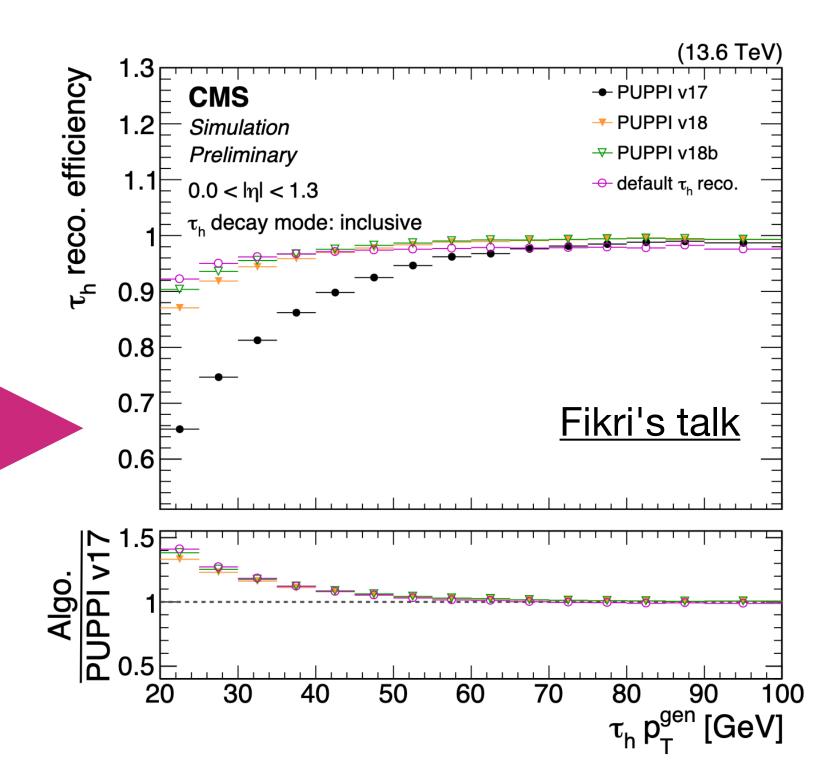


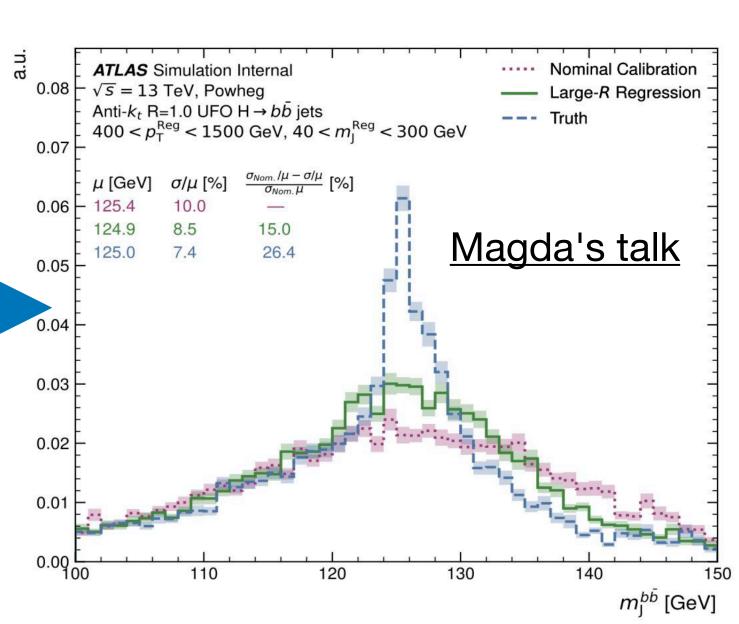
- Low efficiency for τ 's due to the interaction between PUPPI and the association of (displaced) tracks to primary vertices
 - Improved the algorithm to handle these cases, achieving similar or better τ efficiency than the standard reconstruction method



- Can use additional information from b-jets to provide better calibrations
 - LHCb is also getting in on the fun!

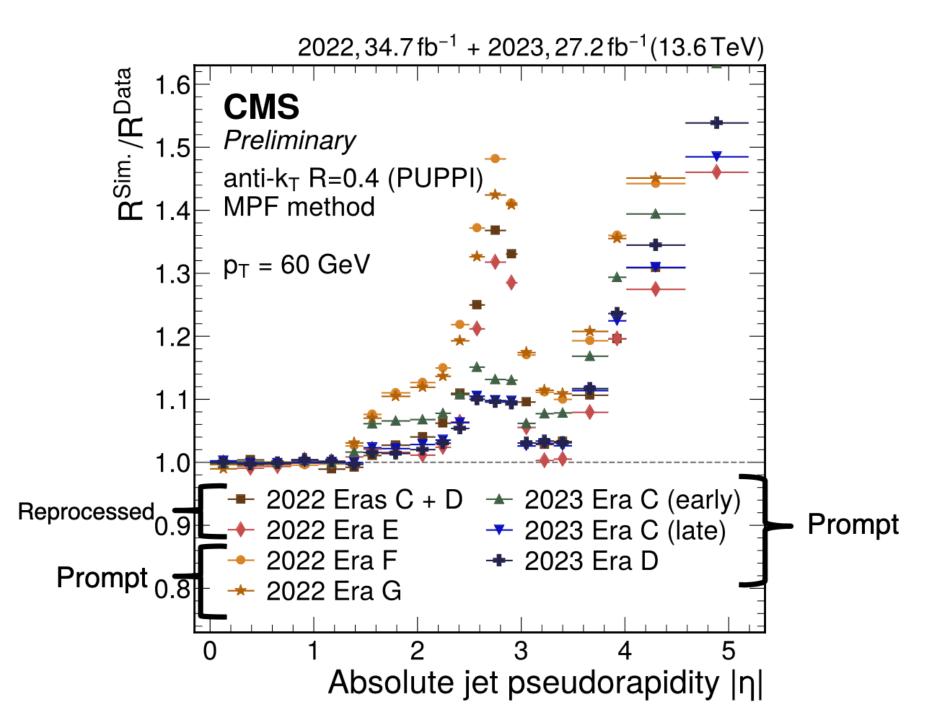
16

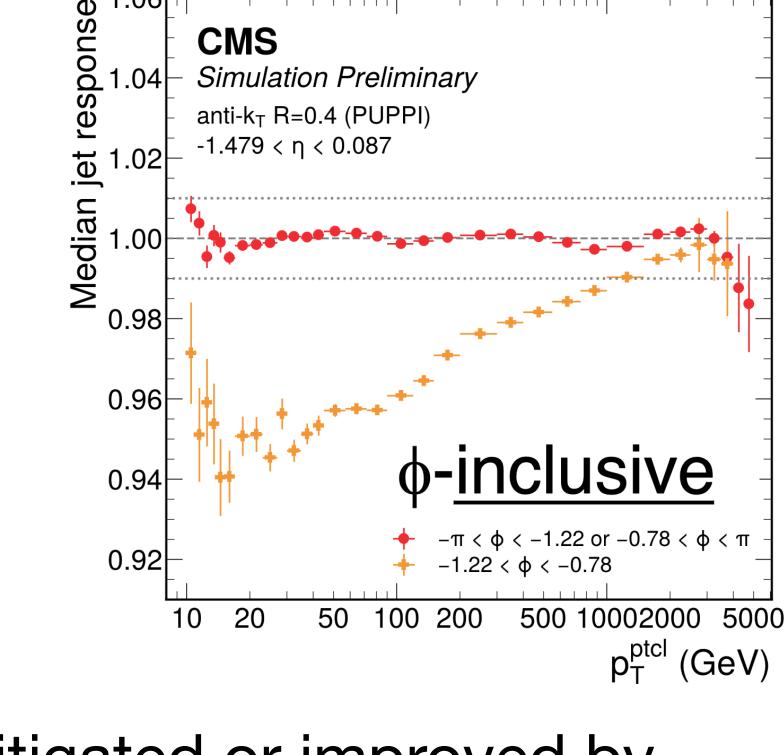




- Detector aging adds challenges to our reconstruction and calibration
 - This will become even more relevant at the HL-LHC

JES residual corrections





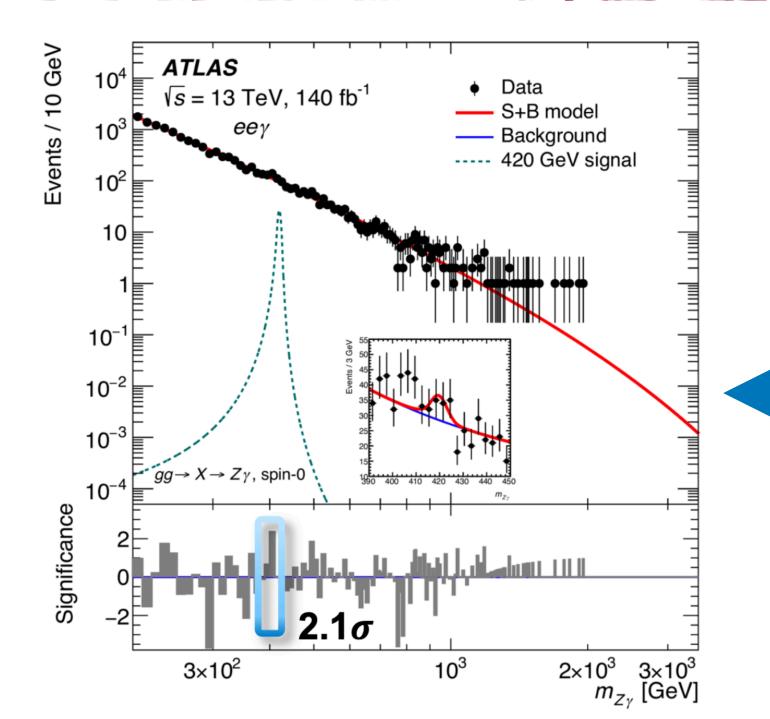
CMS

Simulation Preliminary

anti-k_T R=0.4 (PUPPI)

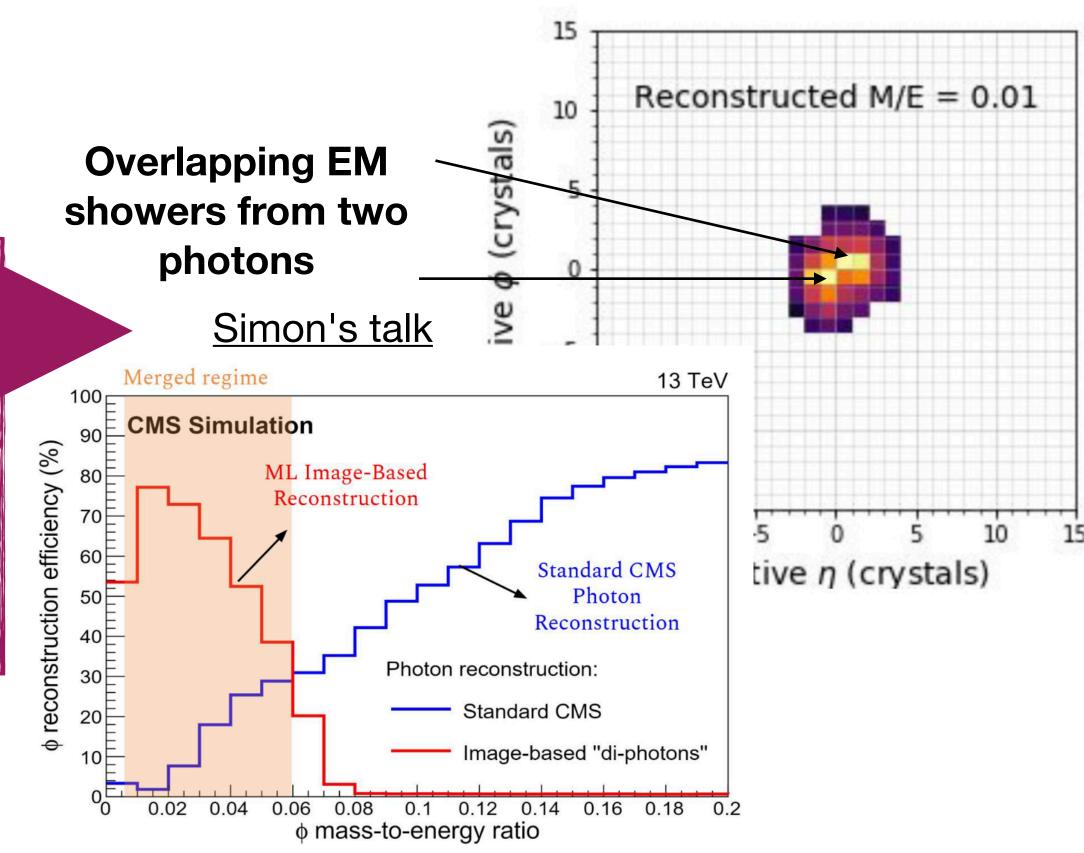
- Many effects can be mitigated or improved by dedicated solutions
 - Need good understanding of detector conditions, and strategies so we can handle these sorts of changes with relative simplicity

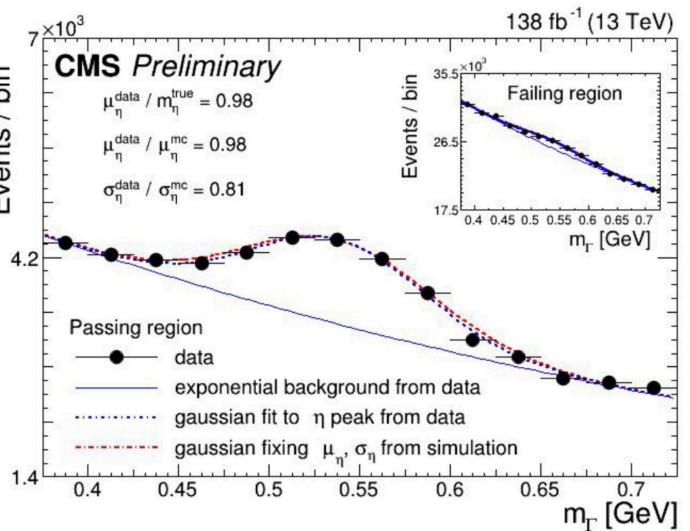
- Jets are not the only thing that can be boosted!
- CMS created a dedicated algorithm to separate showers from boosted di-photons for a BSM search
 - Using hadronic decays within jets to validate the analysis strategy (reconstructing η's)



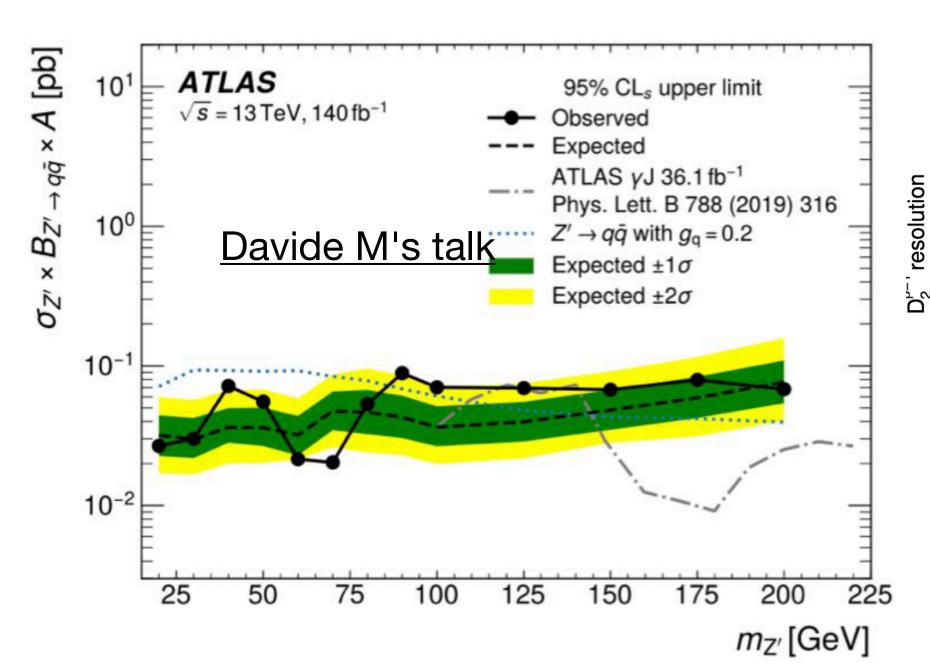
ATLAS has a dedicated algorithm for boosted di-electrons

18

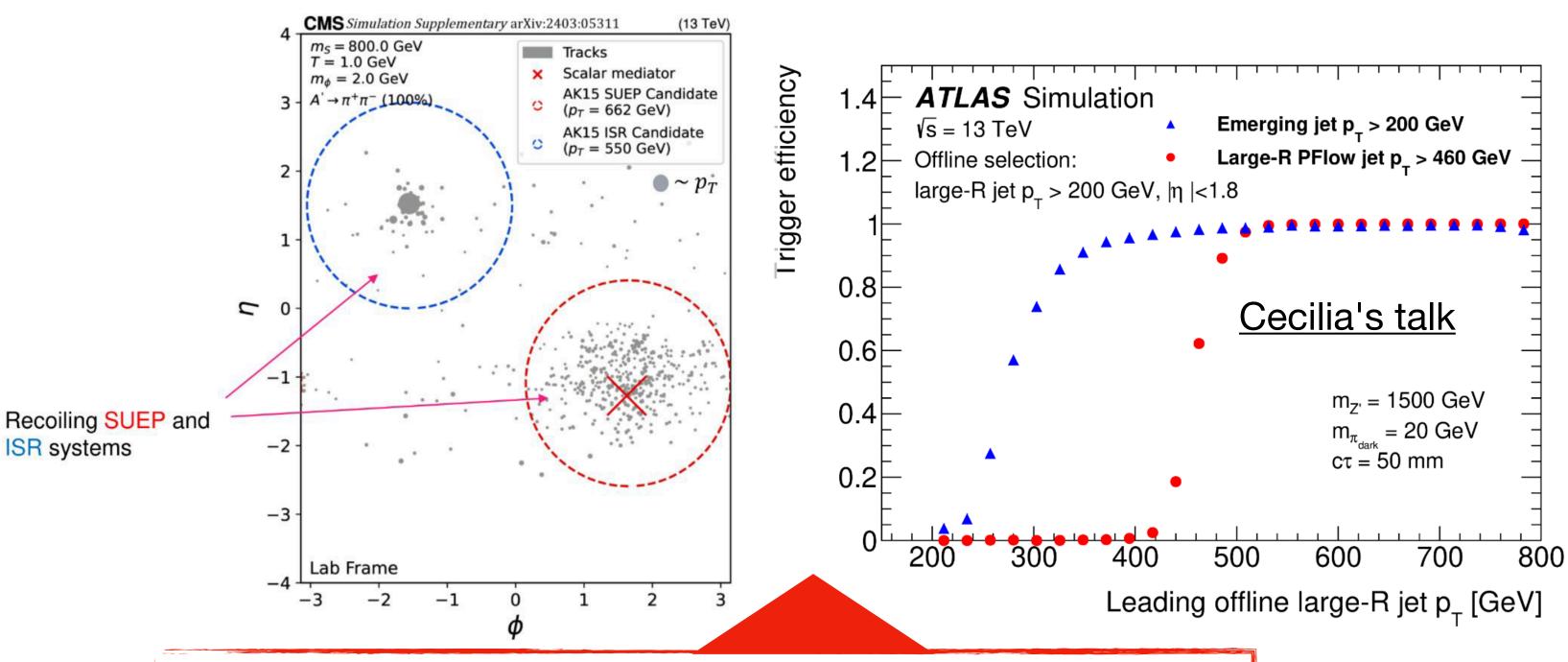




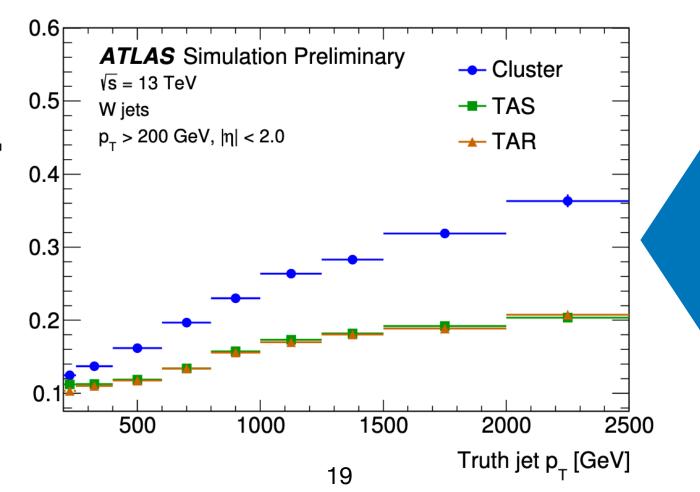
Many searches need dedicated reconstruction and observables to target challenging signatures



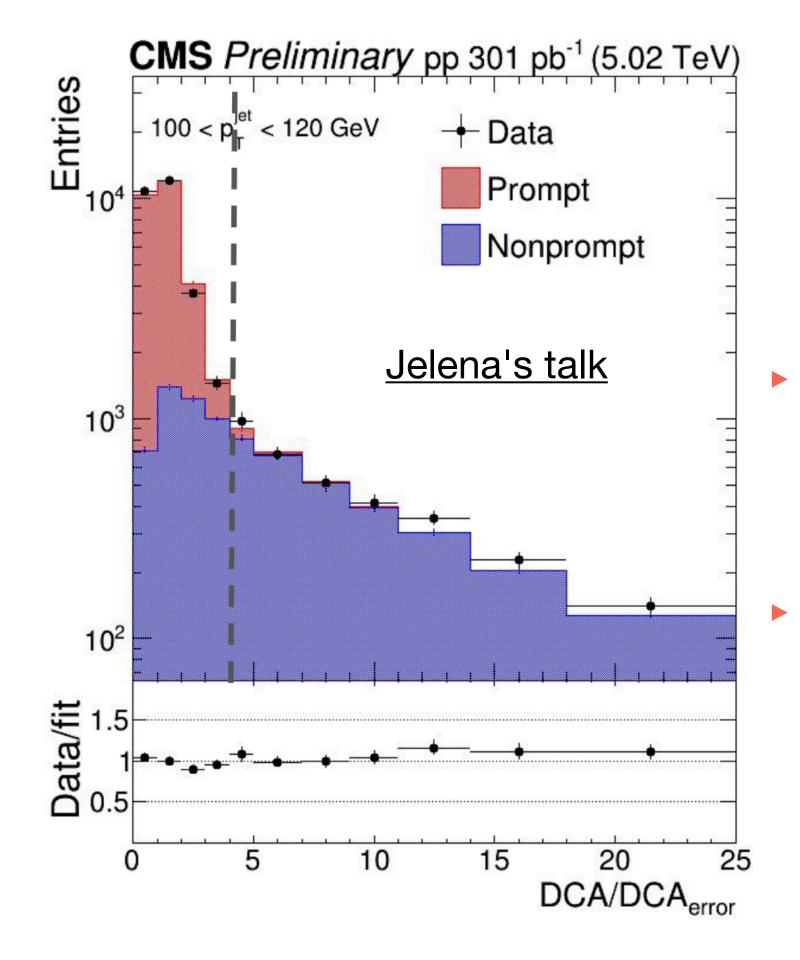
Simon's talk

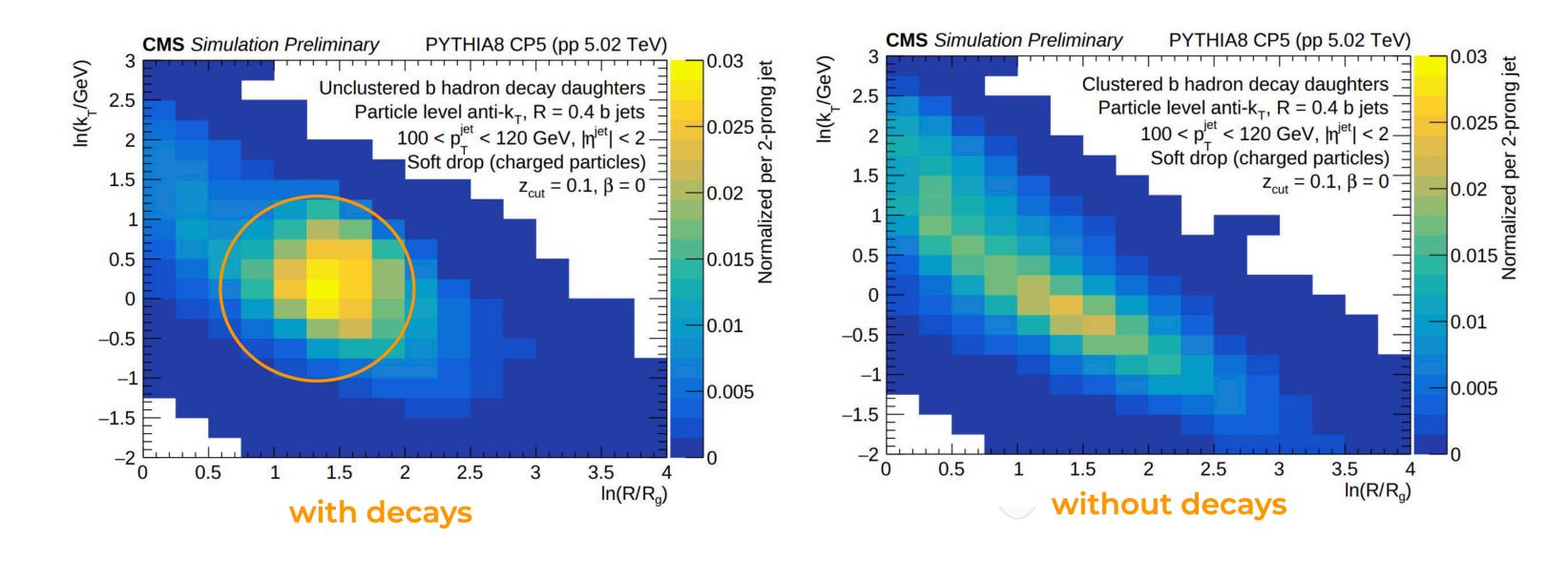


Unusual structure that does not get captured by the standard algorithms



Difficult phase space where standard calibrations do not apply

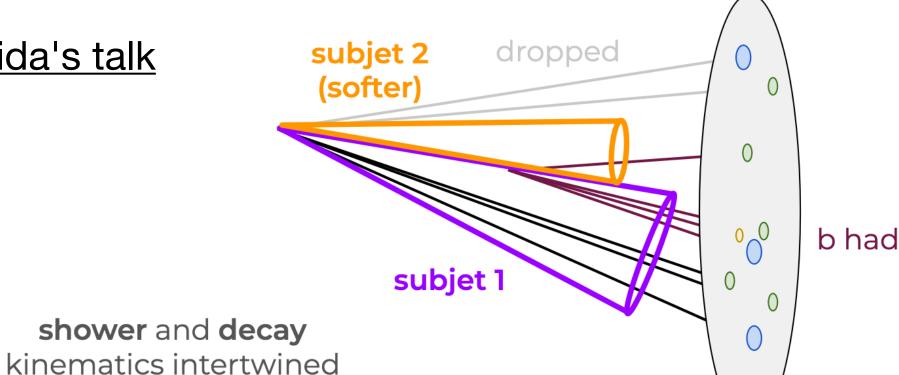




Starting to see more measurements of substructure Lida's talk of heavy flavor jets

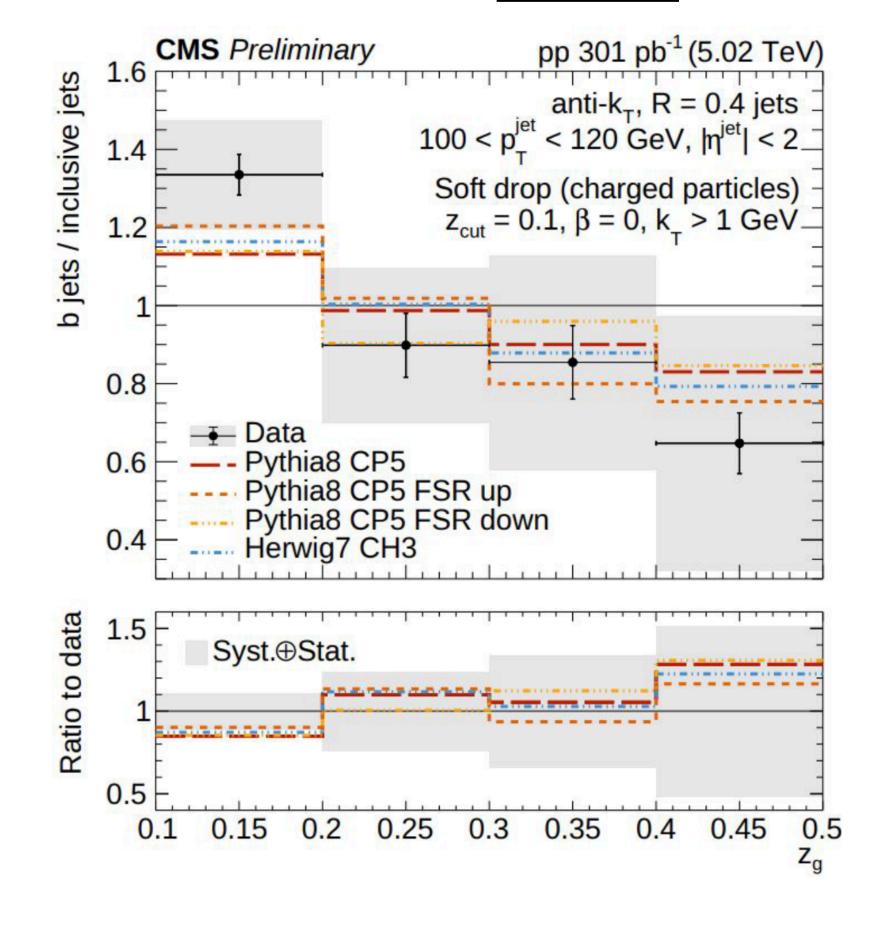
Lots of interesting physics to explore, but also specific experimental challenges

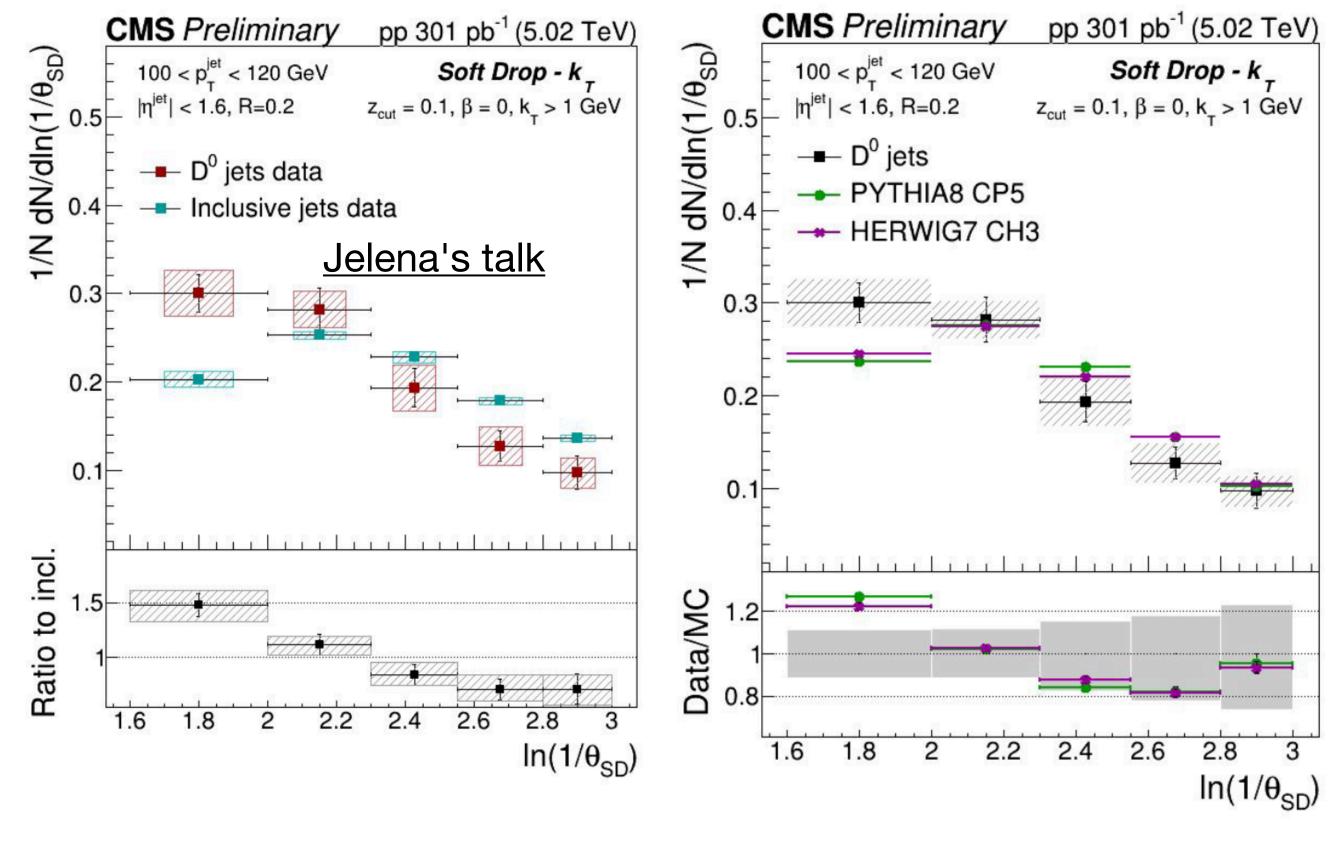
Jet substructure can be spoiled by the B/D meson decays → need to reconstruct these!



Heavy hadron decay daughters do not follow angular ordering

Lida's talk





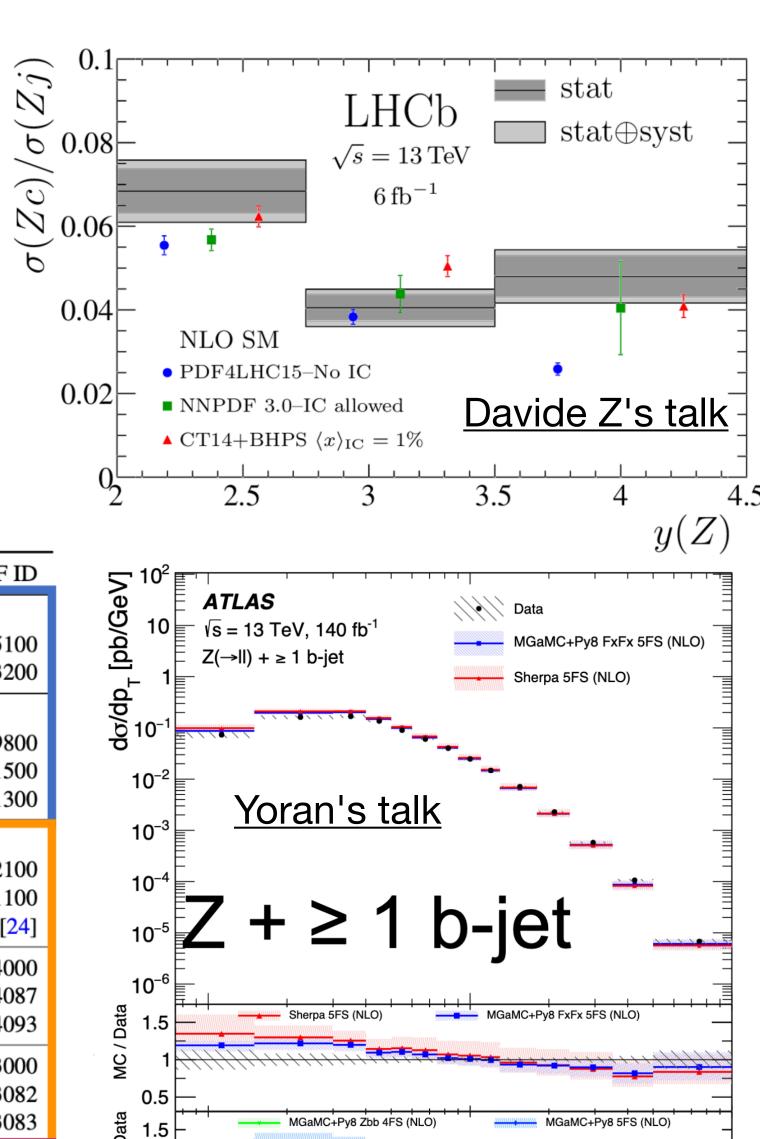
- Clear differences between inclusive and heavy flavor jets
 - More work still needed for full interpretation
 - ... but have potential to provide sensitivity to dead cone effects

Measurements of final states with heavy-flavor jets can be used to test model, and give sensitivity to the presence of intrinsic charm

Factorized changes to the modeling enable detailed studies of the importance of different effects!

Main MC samples		'Generator/settings	Flav. scheme	PDF	LHAPDF ID	
Sherpa 2.2.11 5FS NNPDF3.0 (NNLO) 303200		Main MC samples				
Predictions to test various flavour schemes		MGaMC+Py8 FxFx	5FS	NNPDF3.1 (NNLO) LuxQED	325100	
MGAMC+Py8 5FS NNPDF2.3 (NLO) 229800 MGAMC+Py8 Zbb 4FS NNPDF3.1 (NLO) pch 321500 MGAMC+Py8 Zcc 3FS NNPDF3.1 (NLO) pch 321300 Intrinsic charm (IC) predictions NNPDF4.0 (NNLO) pch (no IC) 332100 NNPDF4.0 (NNLO) 331100 NNPDF4.0 (NNLO) EMC+LHCbZc - [24] MGAMC+Py8 FxFx 5FS CT18 (NNLO) (no IC) 14000 CT18FC - CT18 BHPS3 14087 14093 CT14 (NNLO) (no IC) 13000 13000 CT14 (NNLO)IC - BHPS1 13082 CT14 (NNLO)IC - BHPS2 13083 Fixed-order predictions [3] NLO 5FS PDF4LHC21 93000		Sherpa 2.2.11	5FS	NNPDF3.0 (NNLO)	303200	
MGAMC+Py8 Zbb MGAMC+Py8 Zcc 4FS 3FS NNPDF3.1 (NLO) pch NNPDF3.1 (NLO) pch 321500 321300 Intrinsic charm (IC) predictions NNPDF4.0 (NNLO) pch (no IC) NNPDF4.0 (NNLO) 332100 331100 NNPDF4.0 (NNLO) EMC+LHCbZc − [24] MGAMC+Py8 FxFx 5FS CT18 (NNLO) (no IC) CT18FC − CT18 BHPS3 CT18FC − CT18 MCM-E 14000 14093 CT14 (NNLO) (no IC) CT14 (NNLO) (no IC) CT14 (NNLO)IC − BHPS1 CT14 (NNLO)IC − BHPS2 13083 Fixed-order predictions [3] NLO 5FS PDF4LHC21 93000		Predictions to test various flavour schemes				
MGAMC+Py8 Zcc 3FS NNPDF3.1 (NLO) pch 321300 Intrinsic charm (IC) predictions NNPDF4.0 (NNLO) pch (no IC) 332100 NNPDF4.0 (NNLO) pch (no IC) 331100 NNPDF4.0 (NNLO) EMC+LHCbZc -[24] MGAMC+Py8 FxFx 5FS CT18 (NNLO) (no IC) 14000 CT18FC - CT18 BHPS3 14087 CT18FC - CT18 MCM-E 14093 CT14 (NNLO) (no IC) 13000 CT14 (NNLO)IC - BHPS1 13082 CT14 (NNLO)IC - BHPS2 13083 Fixed-order predictions [3] NLO 5FS PDF4LHC21 93000		MGaMC+Py8	5FS	NNPDF2.3 (NLO)	229800	
Intrinsic charm (IC) predictions NNPDF4.0 (NNLO) PCH (no IC) 332100 NNPDF4.0 (NNLO) 331100 NNPDF4.0 (NNLO) EMC+LHCbZc - [24]		MGaMC+Py8 Zbb	4FS	NNPDF3.1 (NLO) PCH	321500	
NNPDF4.0 (NNLO) PCH (no IC) 332100 NNPDF4.0 (NNLO) S31100 NNPDF4.0 (NNLO) EMC+LHCbZc - [24] MGAMC+Py8 FxFx 5FS CT18 (NNLO) (no IC) 14000 CT18FC - CT18 BHPS3 14087 CT18FC - CT18 MCM-E 14093 CT14 (NNLO) (no IC) 13000 CT14 (NNLO)IC - BHPS1 13082 CT14 (NNLO)IC - BHPS2 13083 Fixed-order predictions [3] NLO 5FS PDF4LHC21 93000		MGaMC+Py8 Zcc	3FS	NNPDF3.1 (NLO) PCH	321300	
NNPDF4.0 (NNLO) 331100 NNPDF4.0 (NNLO) EMC+LHCbZc - [24] MGAMC+Py8 FxFx 5FS CT18 (NNLO) (no IC) 14000 14087 14087 14087 14093 14087 14093 14087 14093 14087 14093 14087 14093 14087 14093 14087 14093 14093 14087 14093 14087 14093 1409						
NNPDF4.0 (NNLO) EMC+LHCbZc				NNPDF4.0 (NNLO) PCH (no IC)	332100	
MGAMC+Py8 FxFx 5FS CT18 (NNLO) (no IC) 14000 CT18FC - CT18 BHPS3 14087 CT18FC - CT18 MCM-E 14093 CT14 (NNLO) (no IC) 13000 CT14 (NNLO)IC - BHPS1 13082 CT14 (NNLO)IC - BHPS2 13083 Fixed-order predictions [3] NLO 5FS PDF4LHC21 93000				NNPDF4.0 (NNLO)	331100	
MGAMC+PY8 FXFX SFS CT18FC - CT18 BHPS3 14087 14093 14093				NNPDF4.0 (NNLO) EMC+LHCbZc	- [24]	
CT18FC - CT18 BHPS3 14087 14093 CT18FC - CT18 MCM-E 14093 CT14 (NNLO) (no IC) 13000 CT14 (NNLO)IC - BHPS1 13082 CT14 (NNLO)IC - BHPS2 13083 Fixed-order predictions [3] NLO 5FS PDF4LHC21 93000 93000 PDF4LHC21 PDF4		MC MC D-0 F-F-	500	CT18 (NNLO) (no IC)	14000	
CT14 (NNLO) (no IC) 13000 CT14 (NNLO)IC – BHPS1 13082 CT14 (NNLO)IC – BHPS2 13083 SFixed-order predictions [3] NLO 5FS PDF4LHC21 93000 93000 PDF4LHC21 93000 PDF4LHC21 93000 PDF4LHC21 93000 PDF4LHC21 93000 PDF4LHC21 93000 PDF4LHC21 PDF		MGAMC+PY8 FXFX	5F8	CT18FC – CT18 BHPS3	14087	
CT14 (NNLO)IC - BHPS1 13082 13083 CT14 (NNLO)IC - BHPS2 13083 CT14 (NNLO)IC - BHPS2 13083 NLO 5FS PDF4LHC21 93000 93000 PDF4LHC21 93000 PDF4LHC21				CT18FC – CT18 MCM-E	14093	
CT14 (NNLO)IC - BHPS1 13082 13083 CT14 (NNLO)IC - BHPS2 CT				CT14 (NNLO) (no IC)	13000	
Fixed-order predictions [3] NLO 5FS PDF4LHC21 93000				, , , ,	13082	
NLO 5FS PDF4LHC21 93000				CT14 (NNLO)IC – BHPS2	13083	
NLO 5FS PDF4LHC21 93000		Fixed-order predictions [3]				
NNLO 5FS PDF4LHC21 93000		NLO			93000	
		NNLO	5FS	PDF4LHC21	93000	

Test multiple theoretical predictions



NLO F.O.

10²

 2×10^{2}

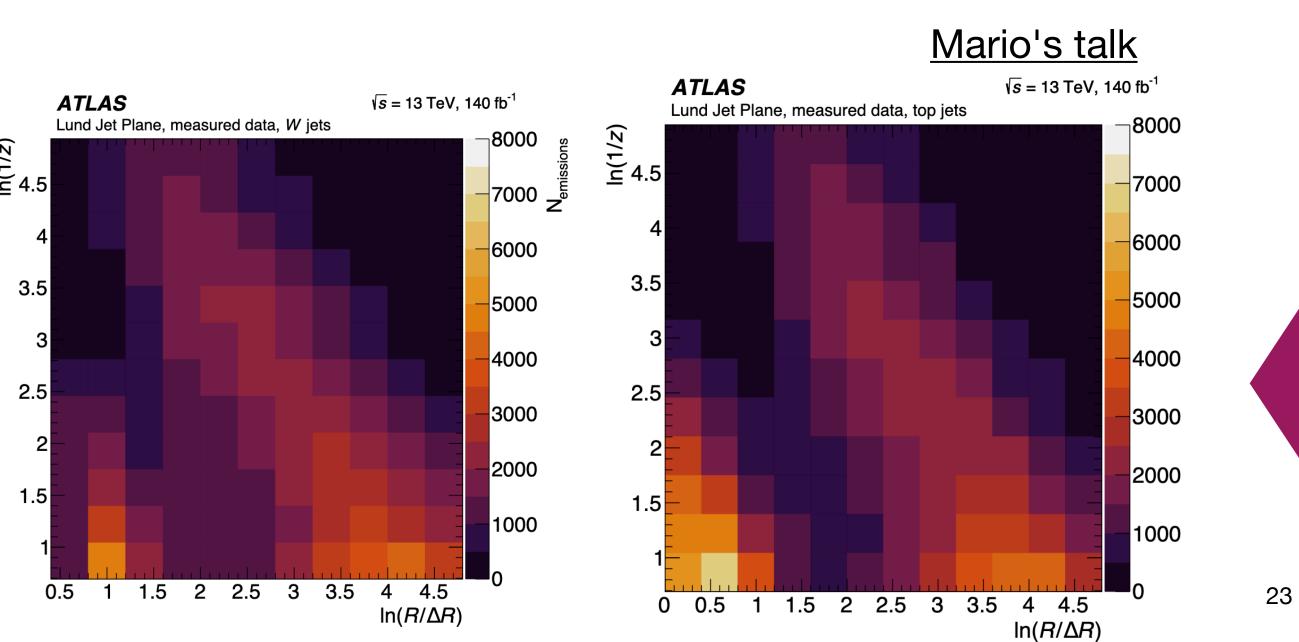
 $p_{T}(Z)$ [GeV]

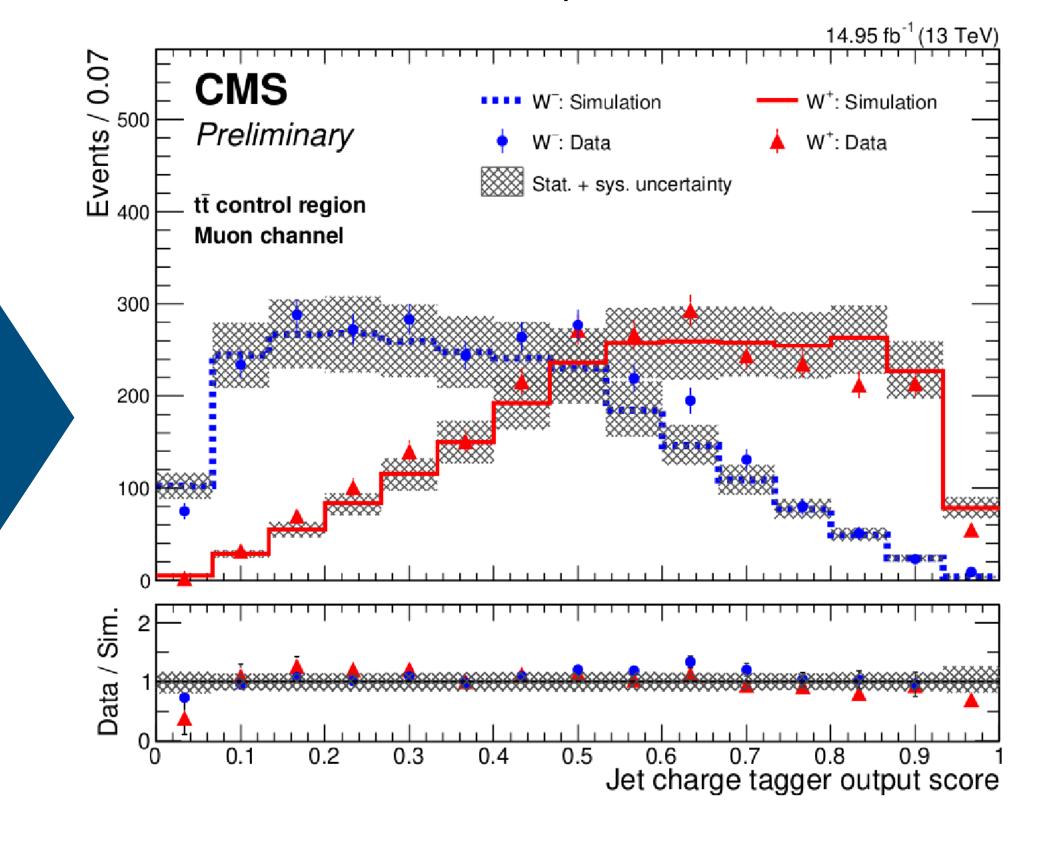
MC / Data

8910

20 30 40

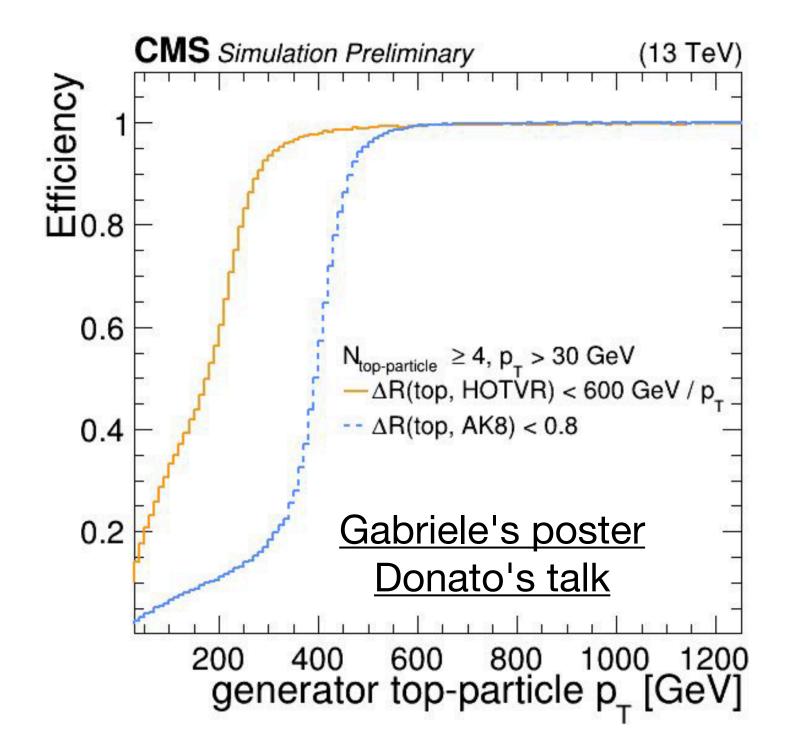
- Not just aiming to identify 2-prong structure → aiming at distinguishing between different hadrons (and their charges)
 - Opens up more possibilities for electroweak measurements in semi- or fully-hadronic channels



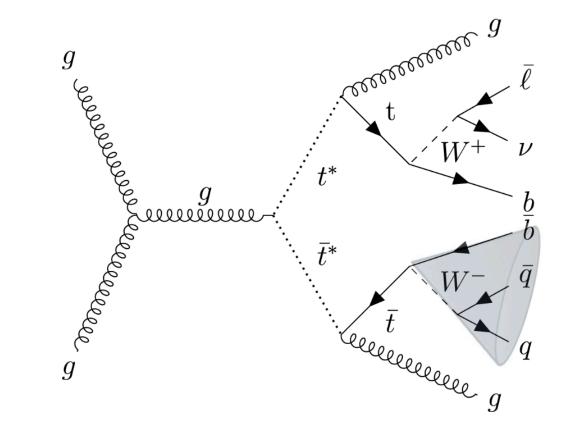


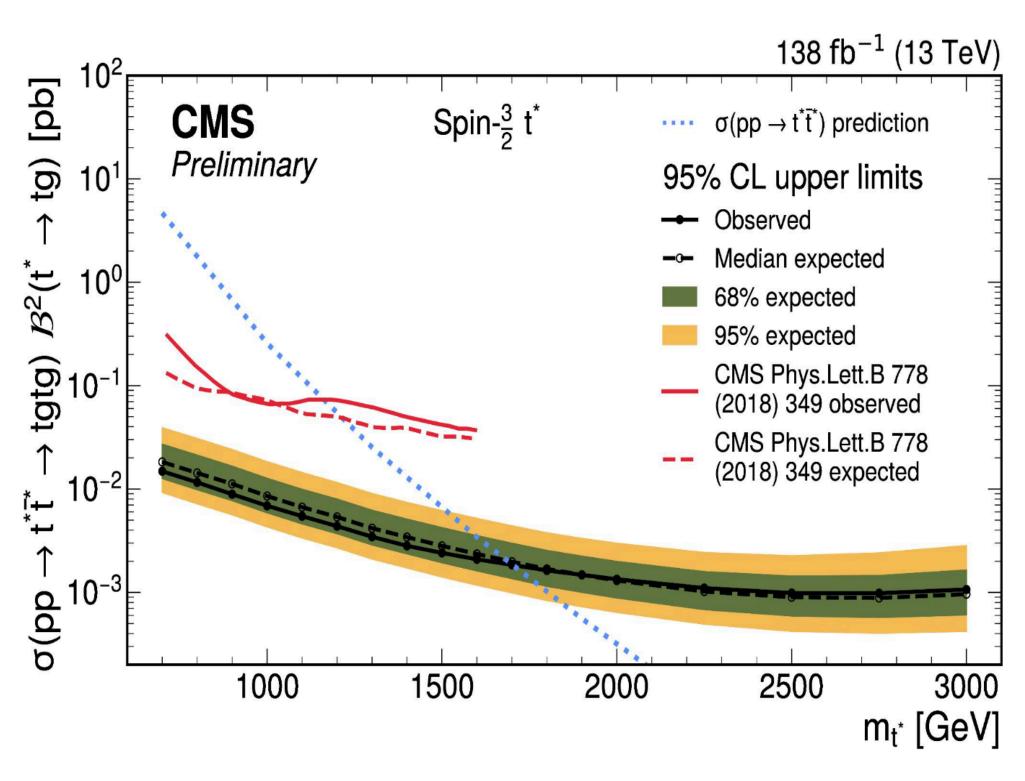
- Dedicated measurements of the substructure of boosted top jets
- Much more complex than q/g jets, but can give insight into the details of their jet formation

- Transition between resolved an boosted channels is difficult to cover efficiently
 - Can instead use variable radius jets to increase the reach of boosted jets while minimizing sensitivity to other effects



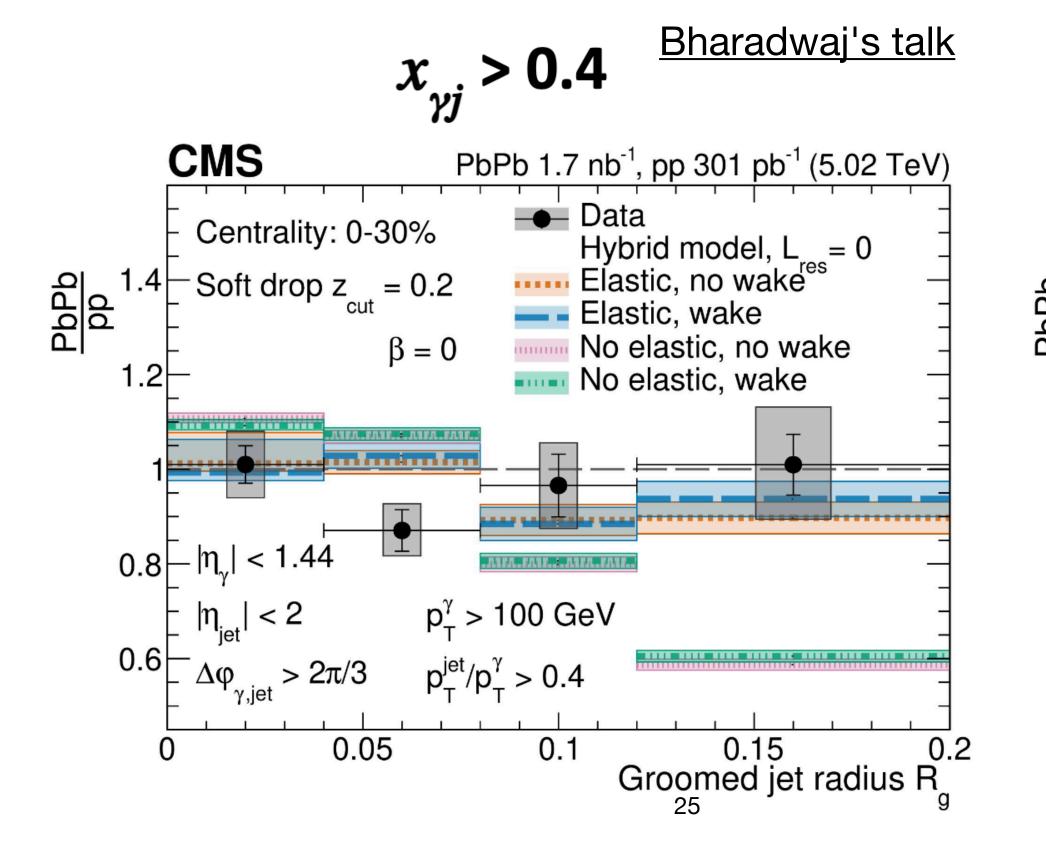
HOTVR has been used by CMS to enhance sensitivity to models like t*

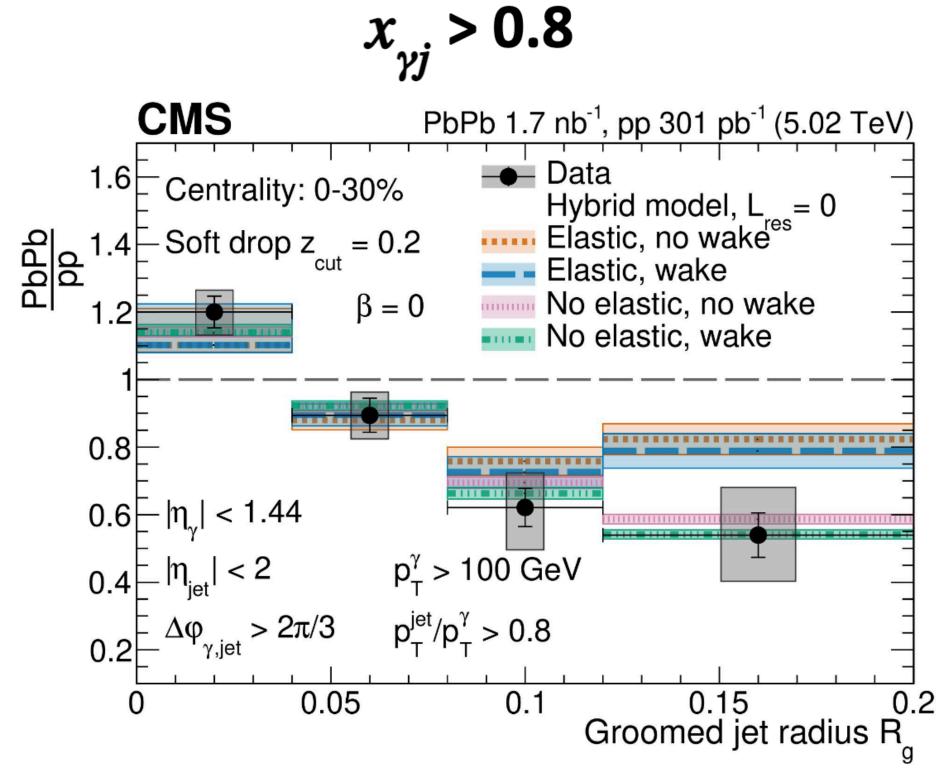




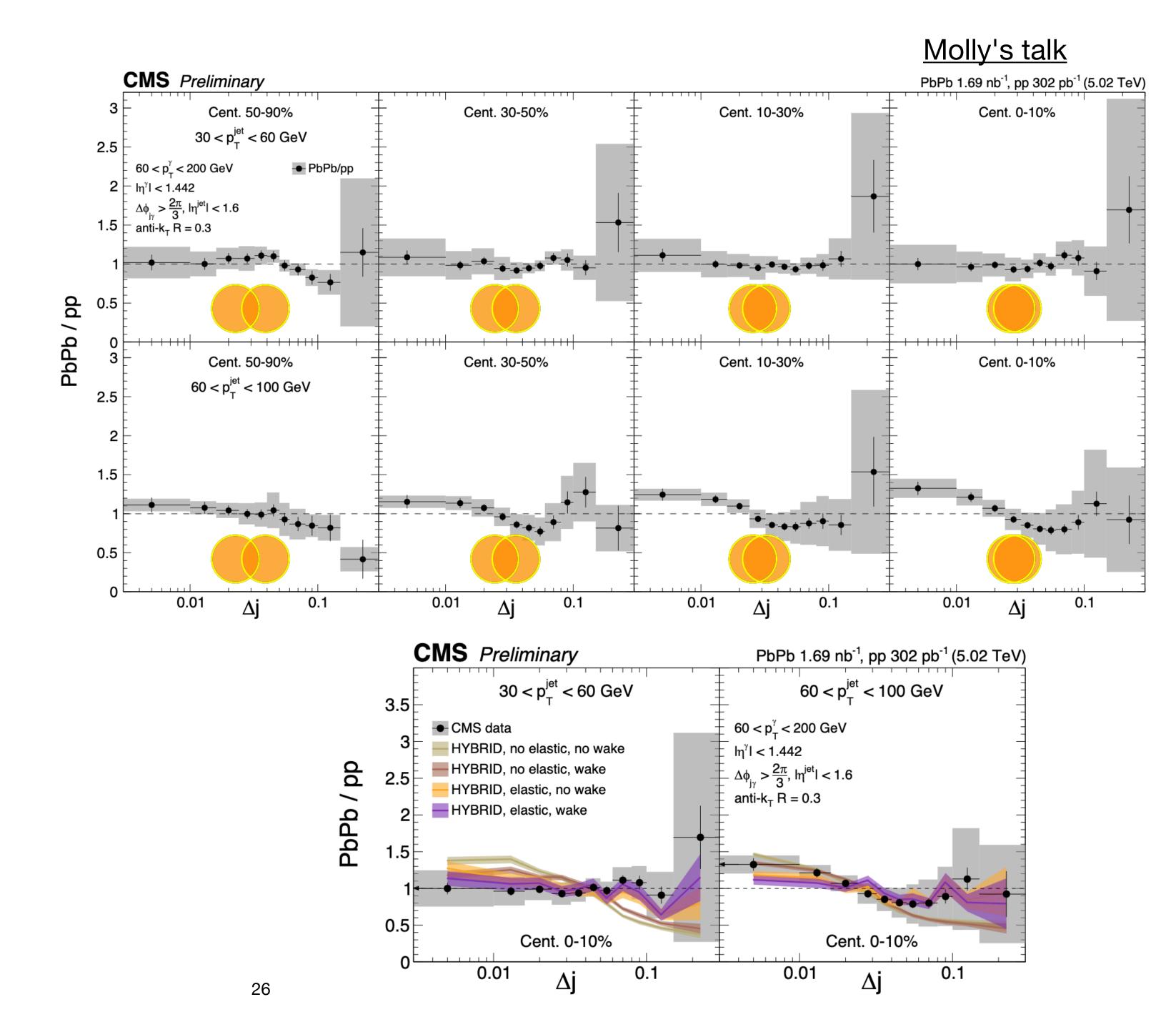
- Using photons as a colorless probe to tag jets
 - Don't expect it to be impacted by the medium → easier interpretation
- Different behavior depending on the p_T balance → see less suppression for more imbalanced events
- No single model describes all of the effects

- Aiming to understand the mechanism of energy loss in heavy ion collisions
- No single model describes all of the effects

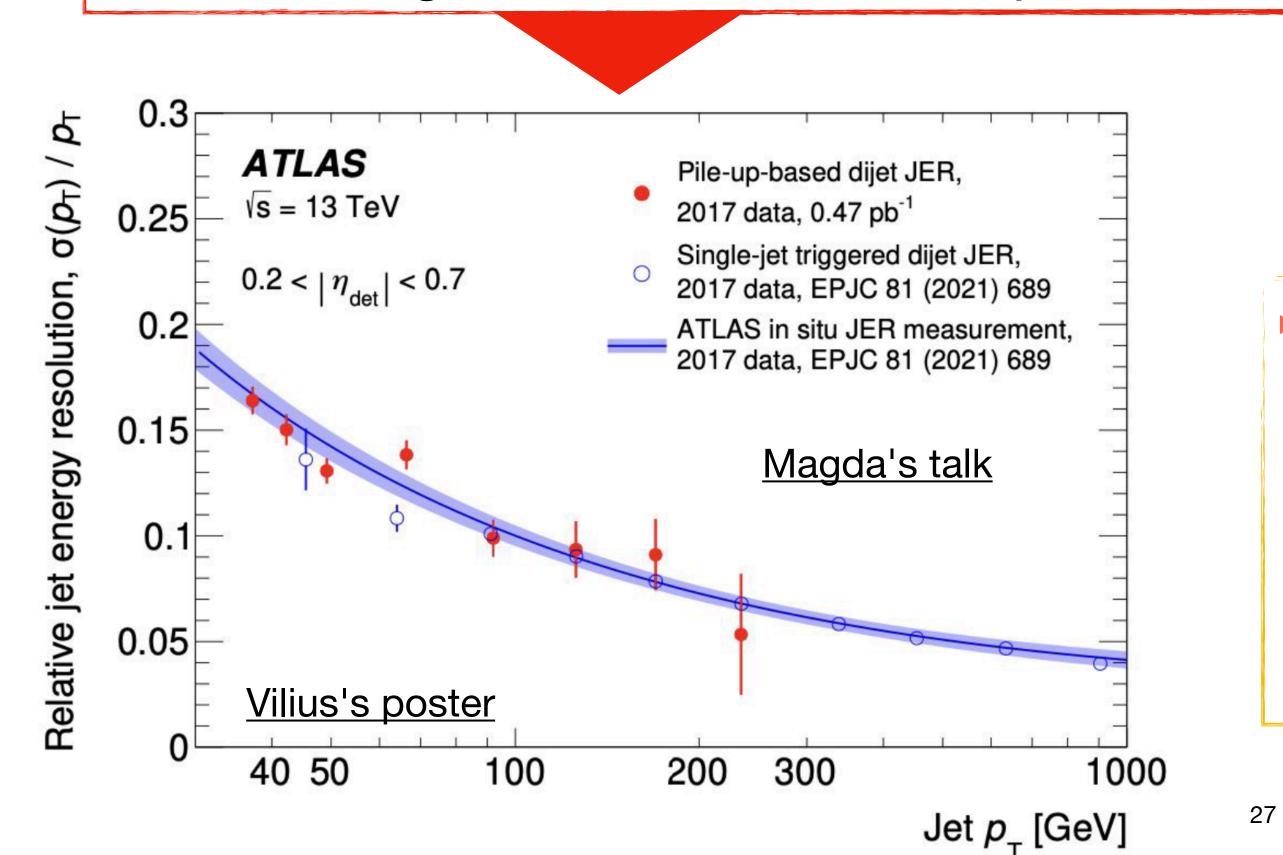


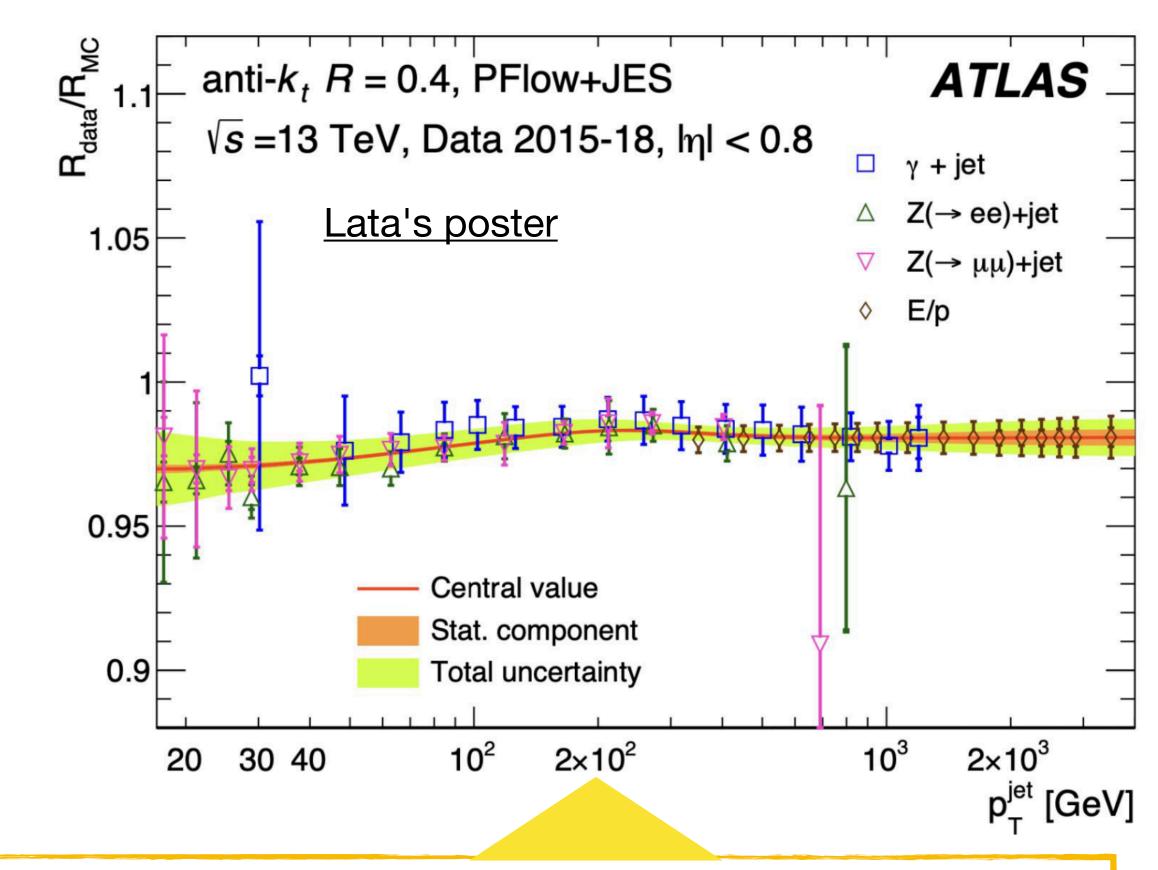


- Also using photon-tagged jets to study axis decorrelation (difference between winner-take-all and E-scheme axes)
- Some indication that wake effects could be important
 - More measurements needed to understand all the effects



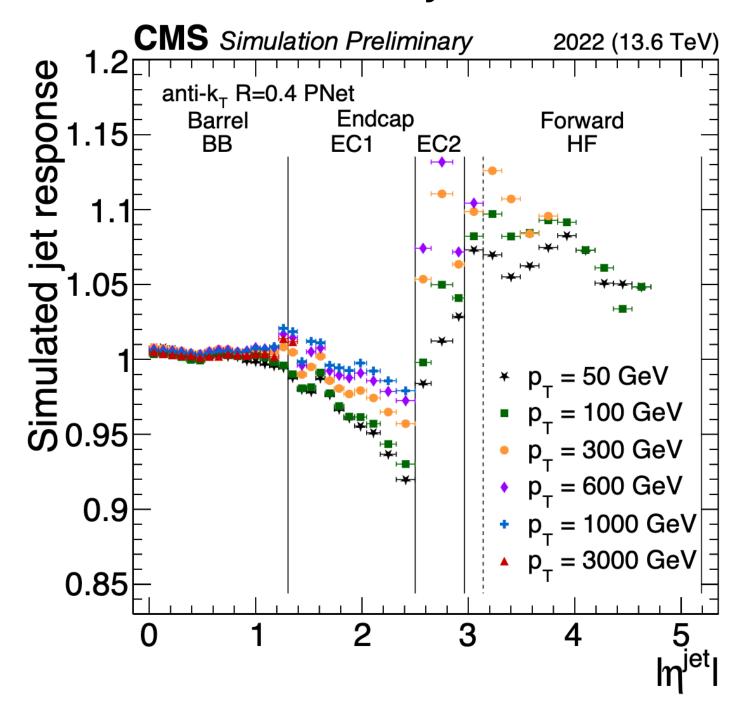
- Using pile-up as a component of the ATLAS jet calibration
 - Not just a nuisance anymore → using it to beat large scale factors at low p_T

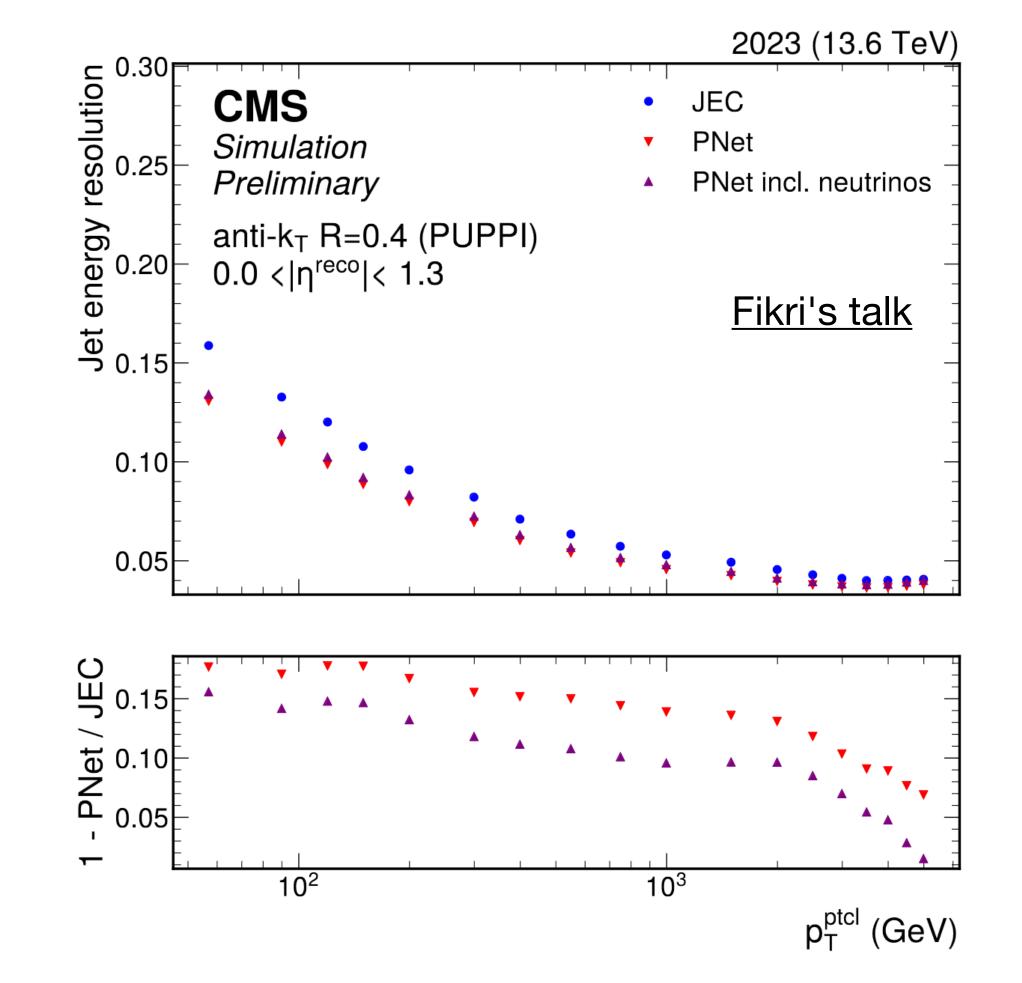




- Using single particle deconvolution to reduce jet uncertainties at high p_T
 - Applications to deriving in-situ jet calibrations with fewer steps for certain uses

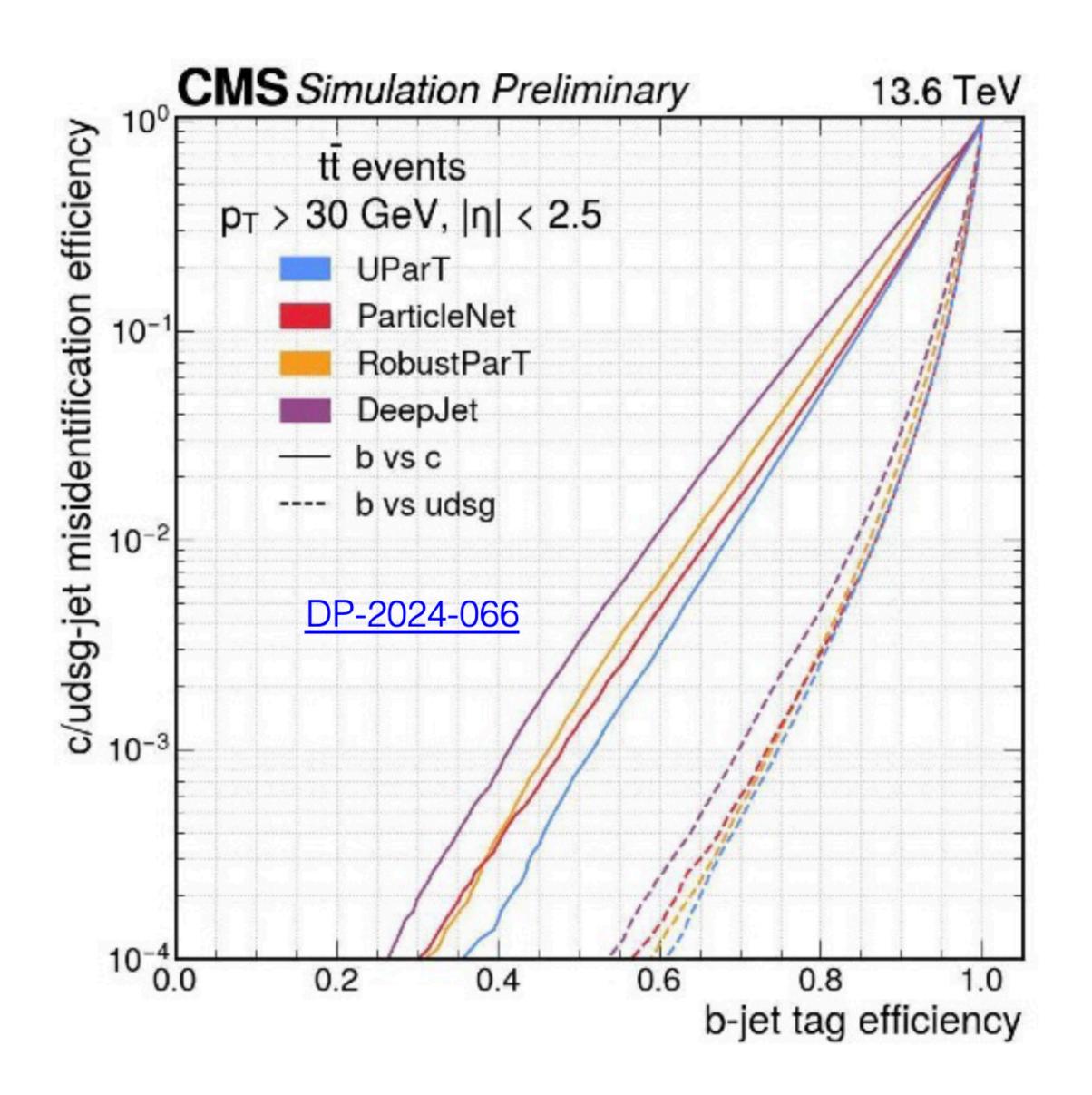
- CMS is using particle net to provide a regression on the jet energy scale
 - Significant improvement to the jet energy resolution!
 - More information about the jet constituents provides more accurate jet corrections





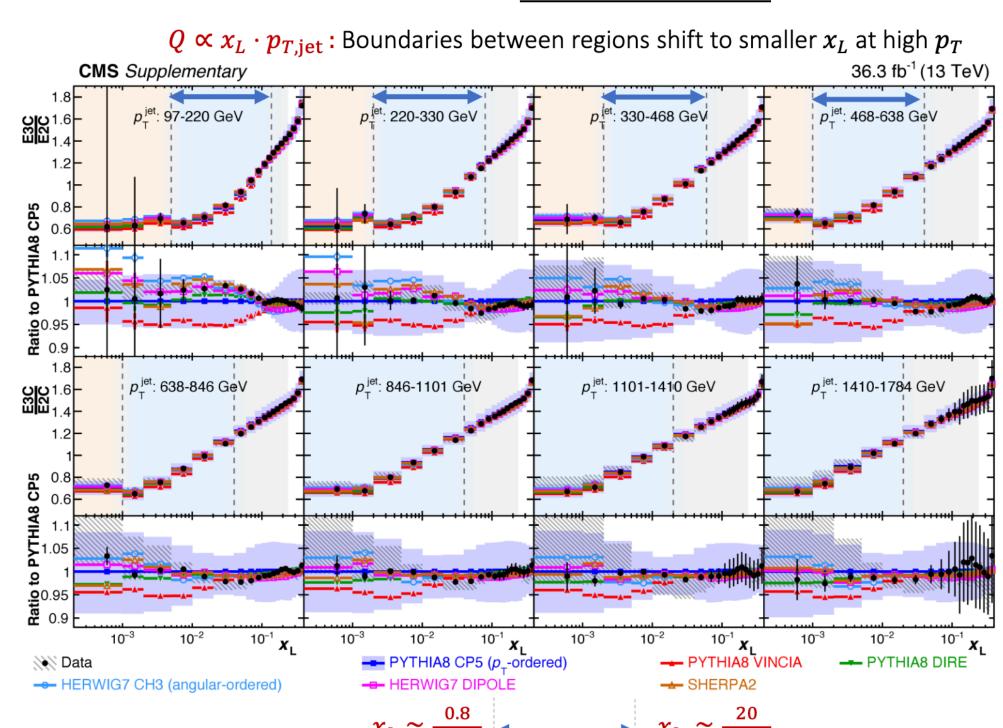
Challenging to make this work over the full phase space, but has potential to be used for the full MC calibration

- UParT uses more output nodes to enhance tagger performance
 - Teaching the network more about the inputs provides better sensitivity
- Also includes flavor-aware jet energy and resolution regression!
- Still a particle-net variant, but still finding ways to enhance the performance



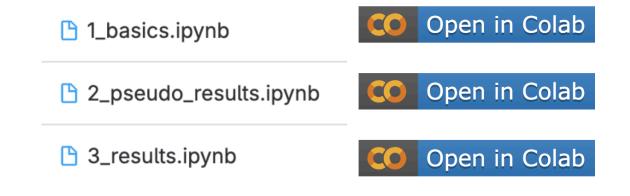
 Unbinned multidifferential unfolding techniques enable new types of open data

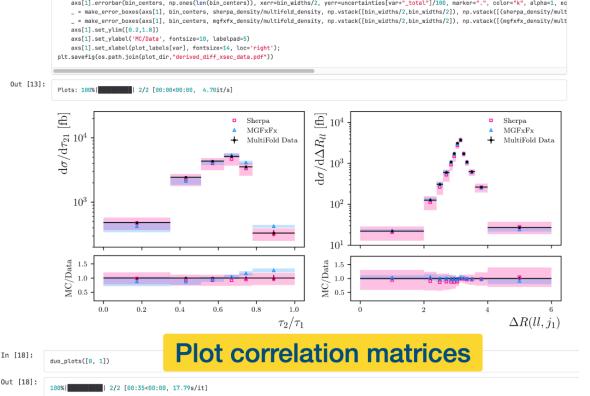
Kaustuv's talk



Dataset & Jupyter Notebooks

- Datasets:
 - https://zenodo.org/records/11507450
- · Codebase:
 - https://gitlab.cern.ch/atlas-physics/public/ sm-z-jets-omnifold-2024
- Notebooks:



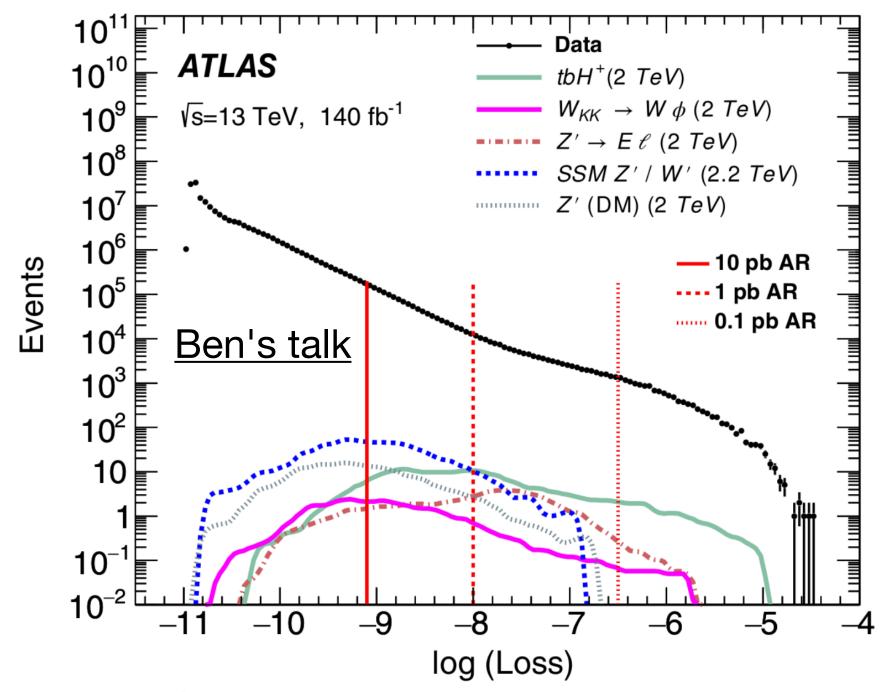


Construct derived observables



Challenging to measure experimentally, but many potential applications!

- Anomaly detection could help us find physics that we wouldn't have seen with a traditional search (or wouldn't have dedicated a search to)
- Many open questions about how we would interpret what we see (but this would be a good place to be)



Florian's talk

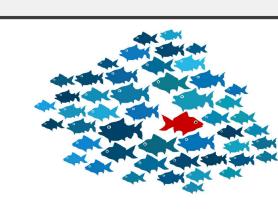
Traditional search

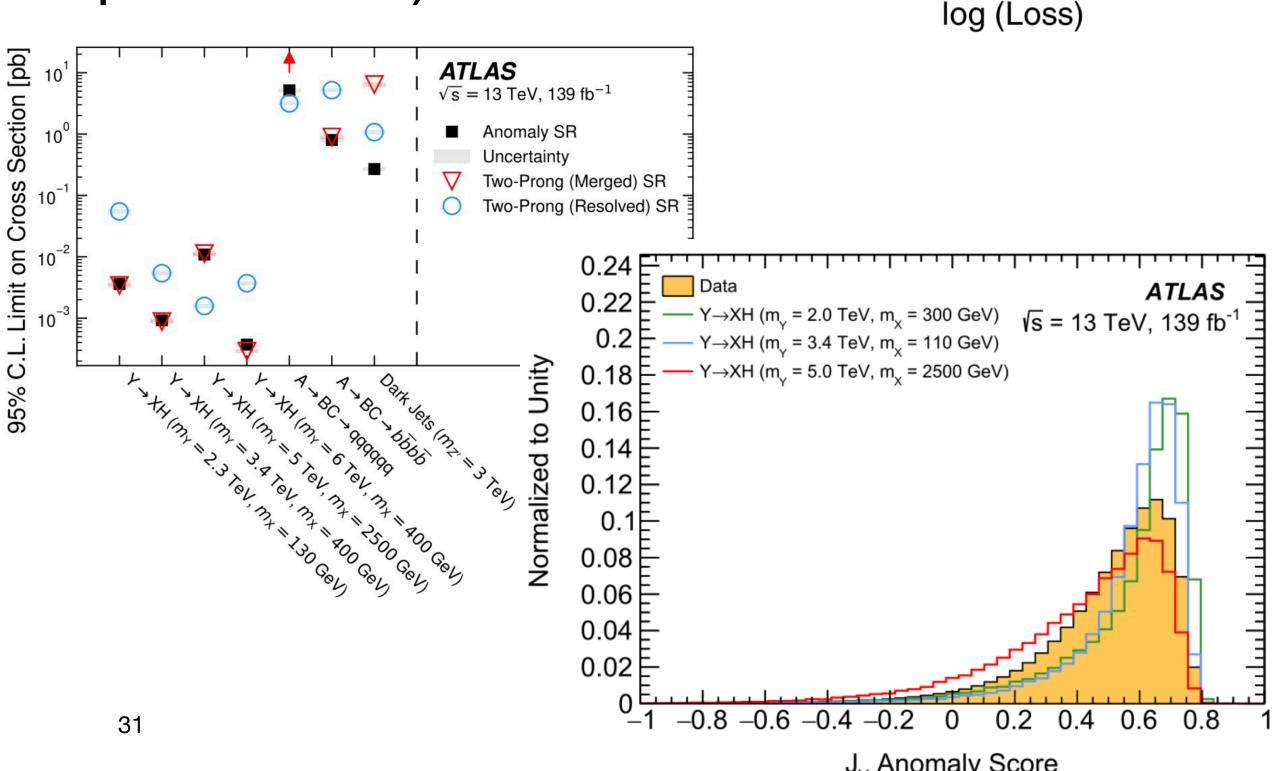
- Targets a specific new physics signal model
- Maximum sensitivity to this signal
- Potentially very little sensitivity to different experimental signatures



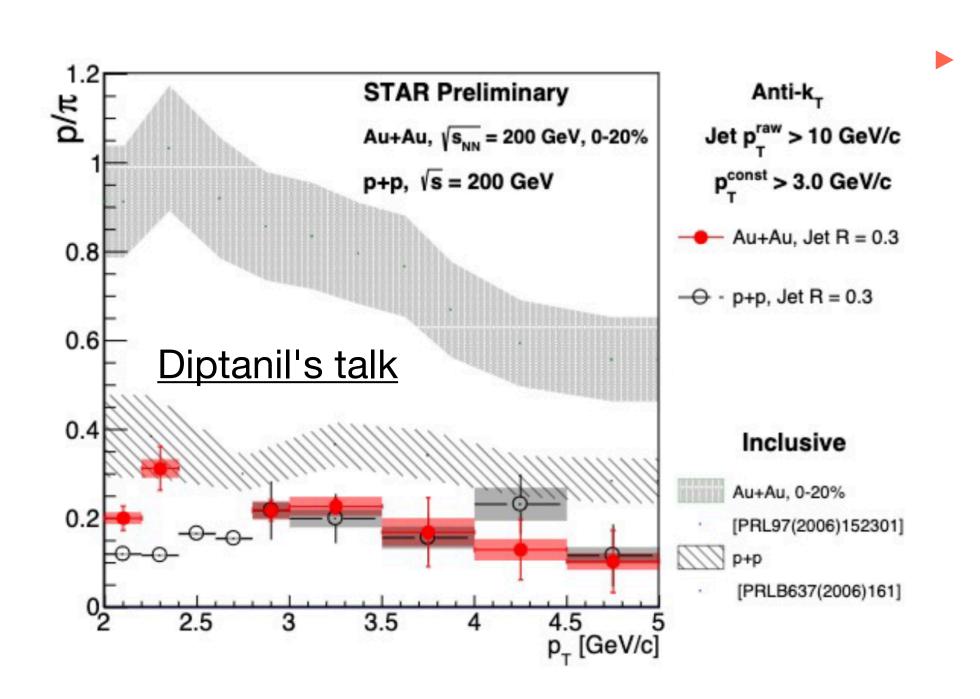
Anomaly detection

- Makes no/few assumptions about the new physics
- Smaller sensitivity compared to traditional search for the target signal
- Sensitive to a wide range of new physics scenarii!

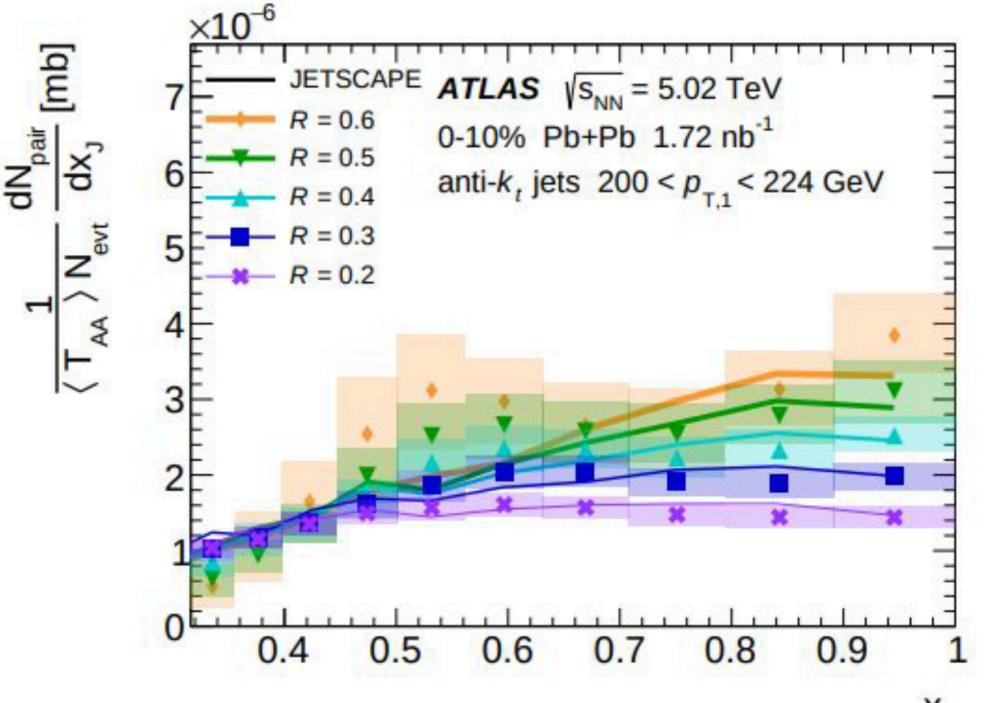


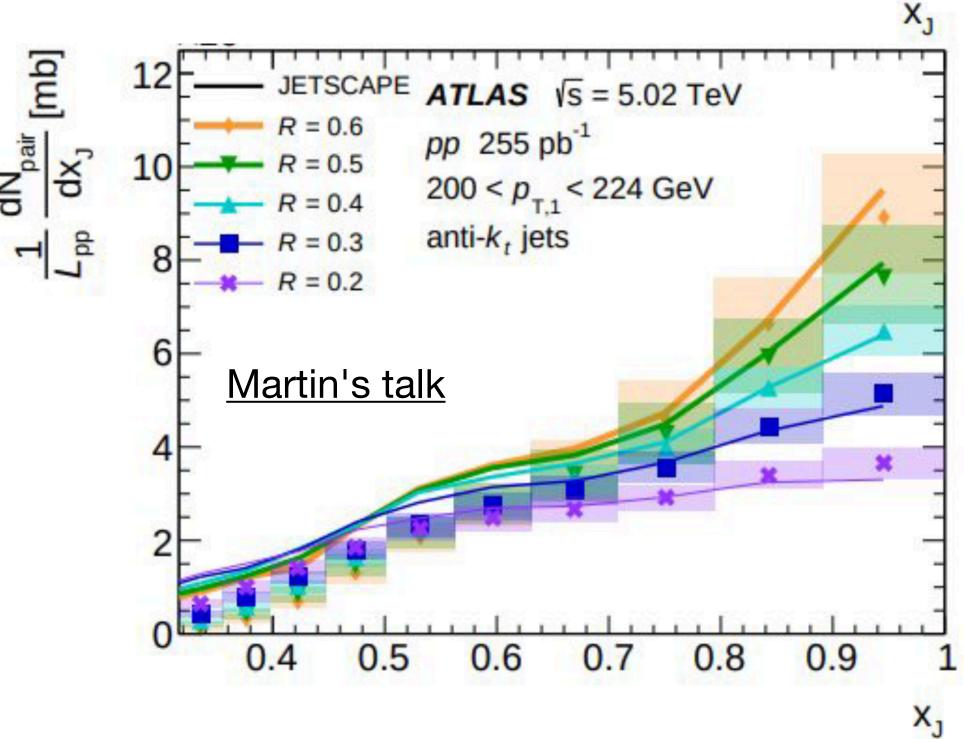


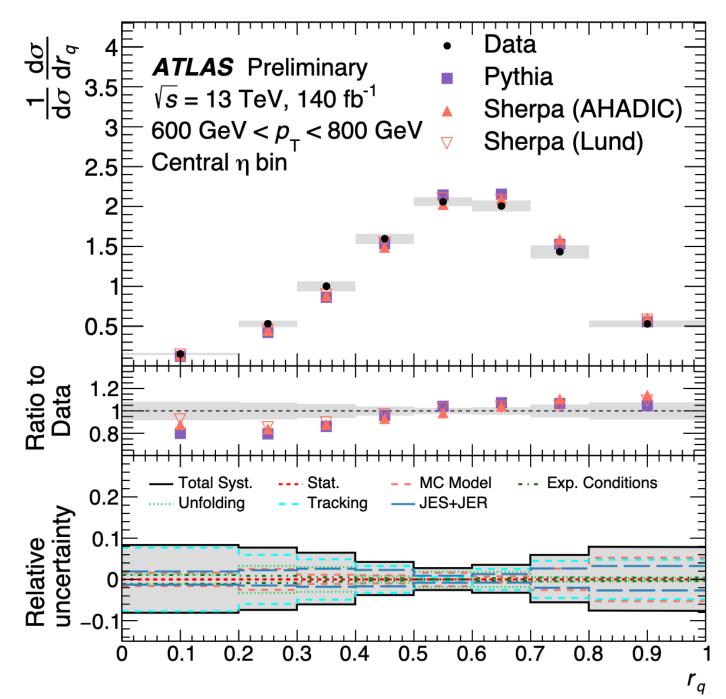
- Energy loss in heavy ion collisions depends on the jet radius as well as the dijet balance
- Clear differences with respect to pp collisions



Also trying to understand which hadrons are produced in jets, and how this changes with heavy ion collisions



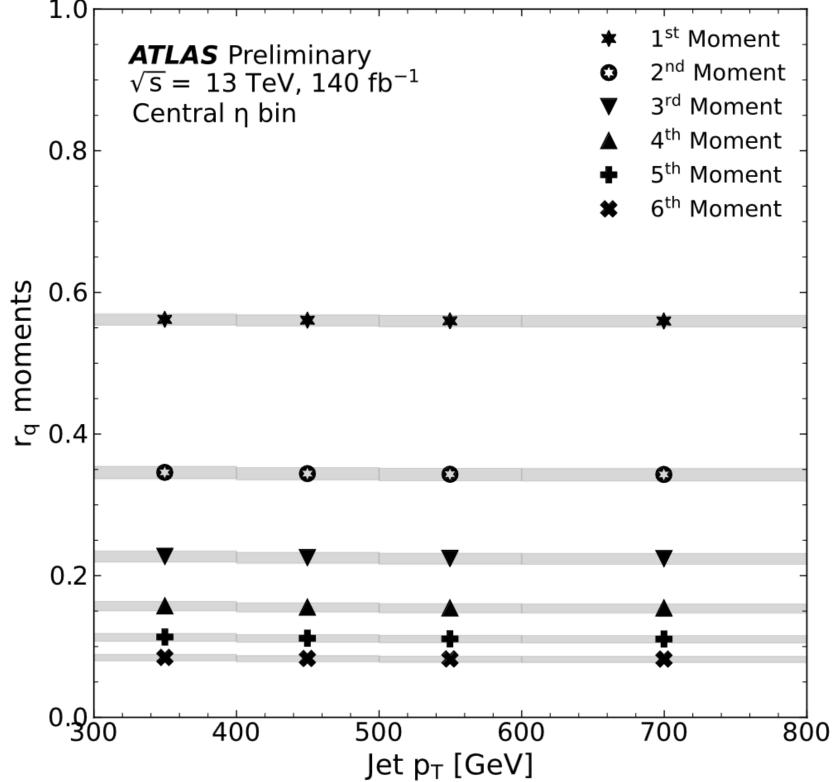


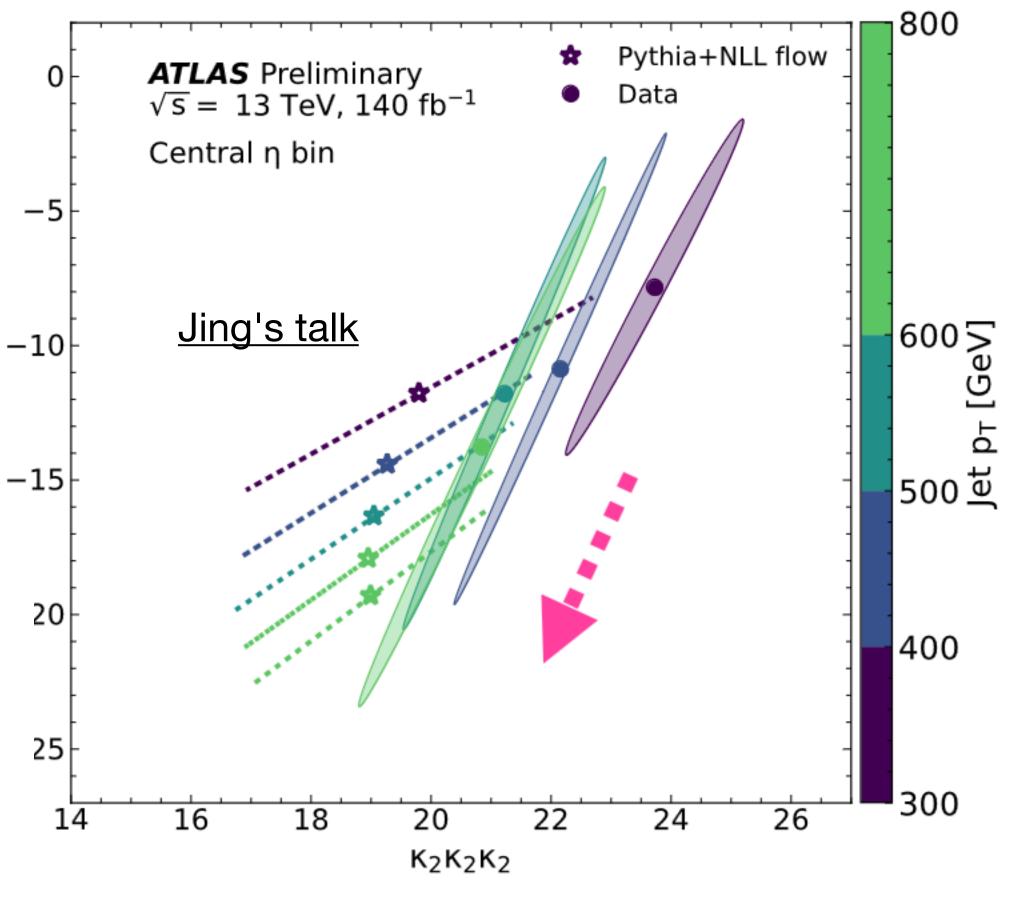


Charged fraction of a jet seems like a simple fundamental quantity



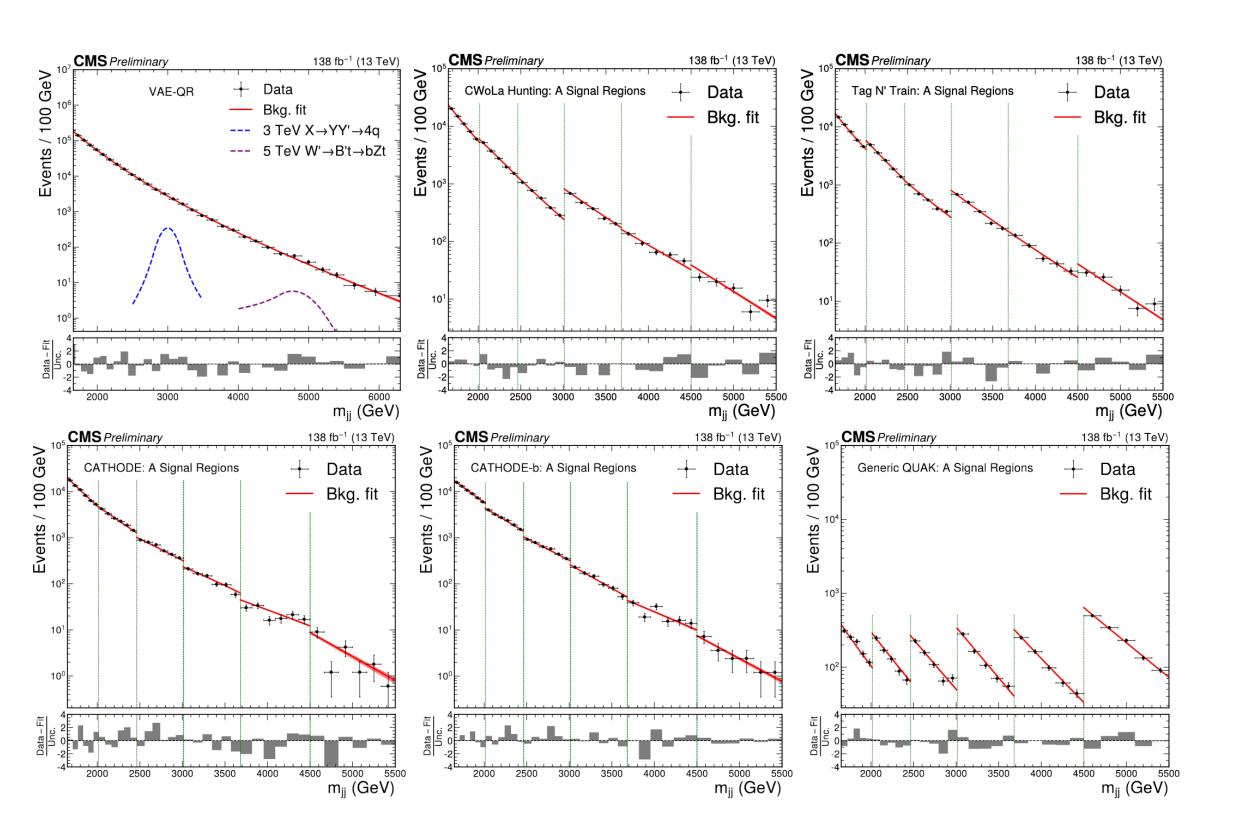
Jet p₁ [GeV]

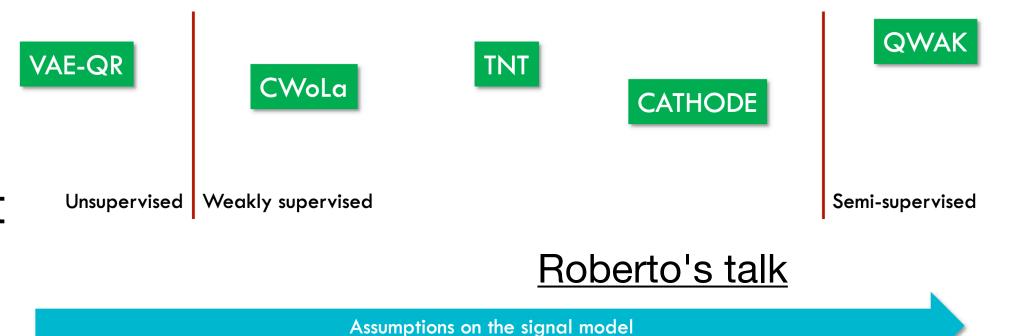




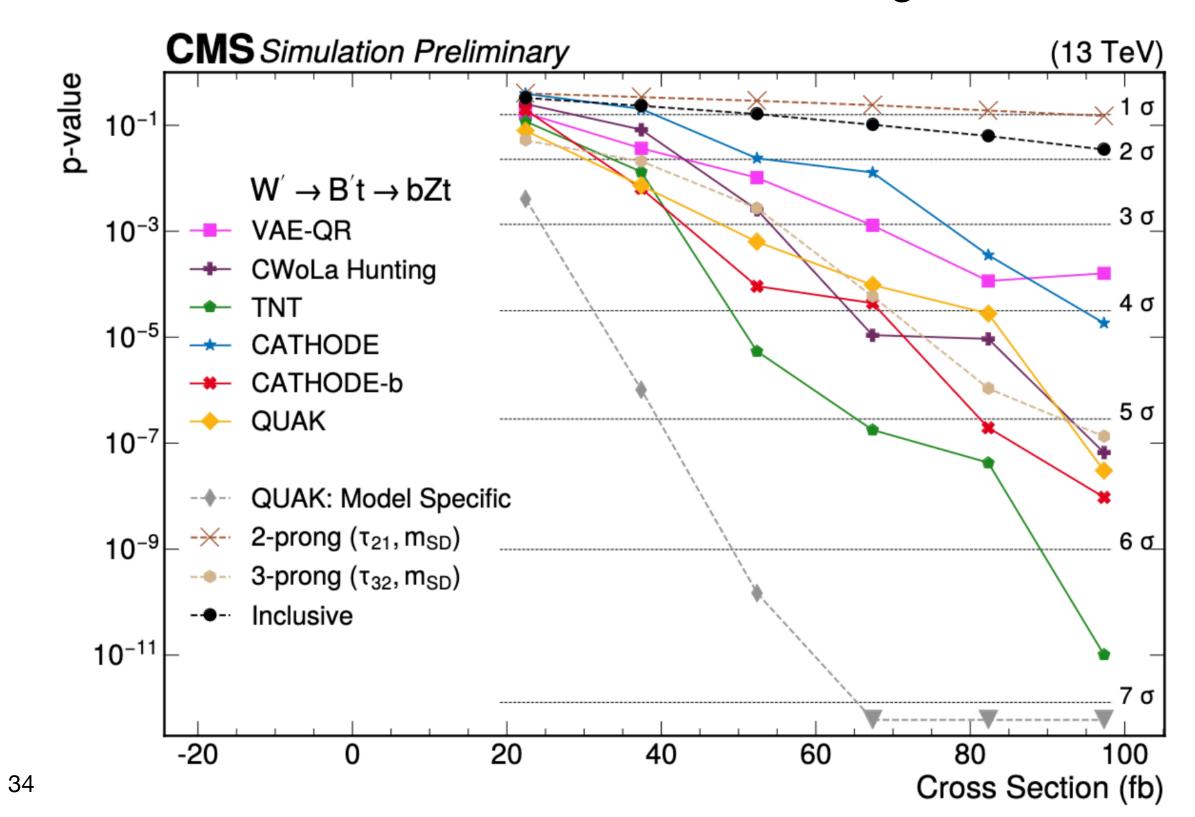
Correlations between cumulants reveal nonlinear mixings of moments, with deeper connections to QCD

- Many different model-agnostic search strategies on the market
- No single method outperforms the others
- Complementary sensitivity from different algorithms, and enables comparisons across a wide range of models

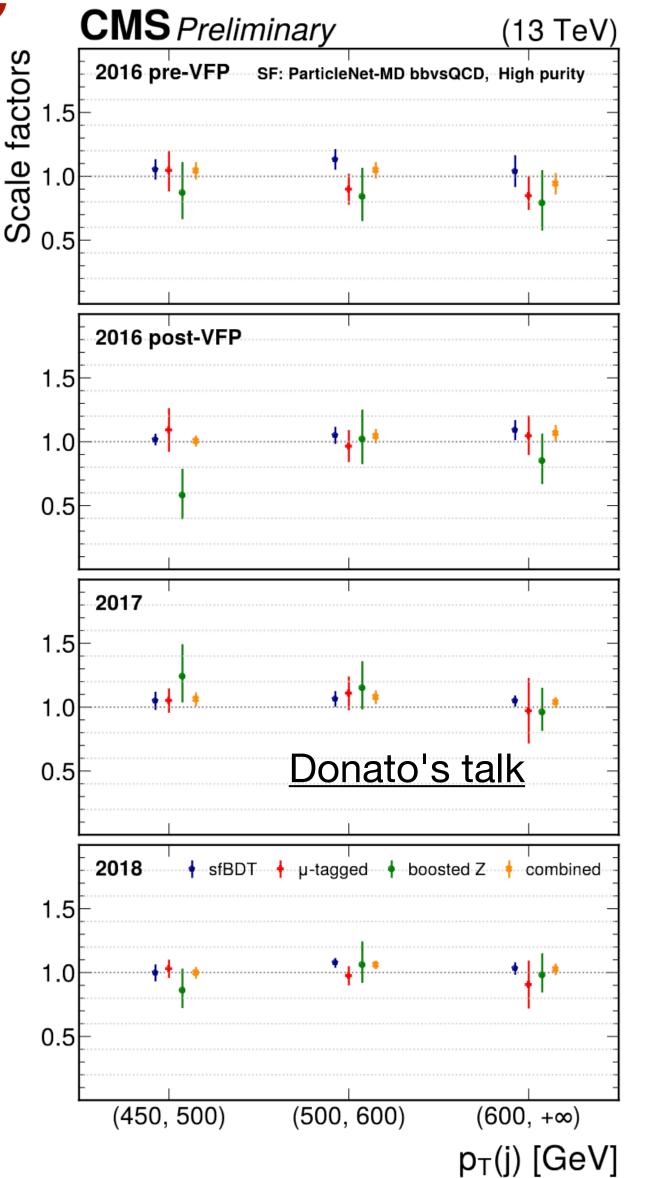


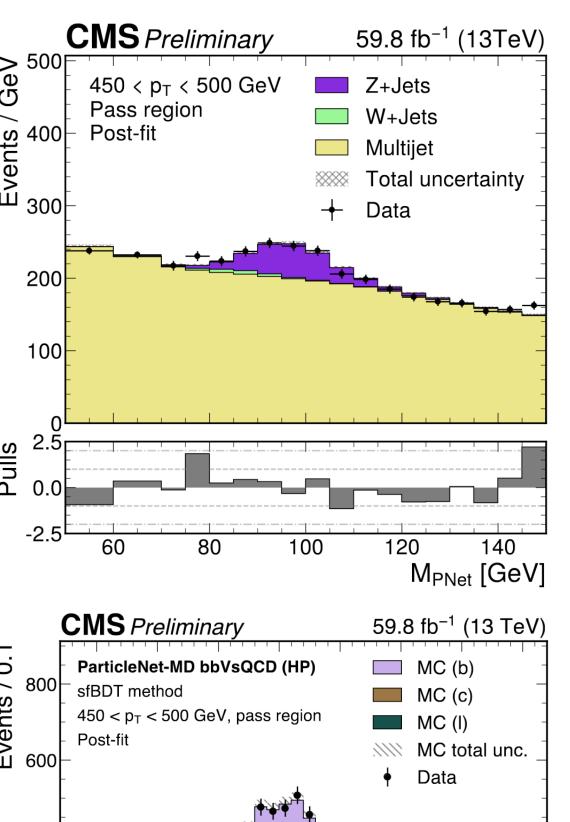


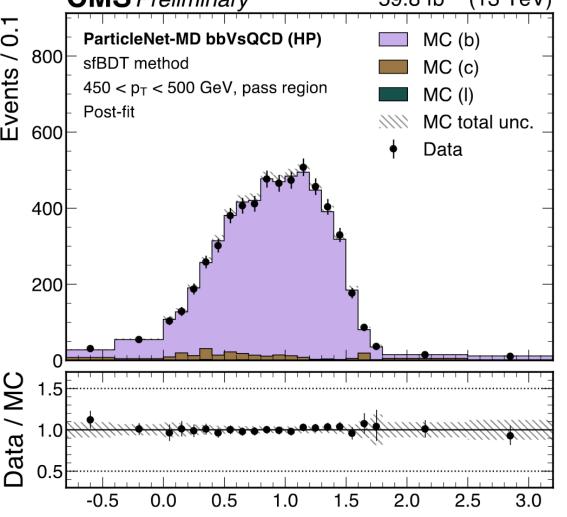
Even though these are model agnostic, can gain a lot through understanding their behavior across a wide range of models



- Need tagger calibrations
- Difficult to get for rare processes (like H->bb)
- Combining scale factors from multiple methods to get the tagger calibration
 - Using multiple topologies: g→bb and Z→bb
 - Good cross-check that scale factors are consistent
 - Will improve with higher stats

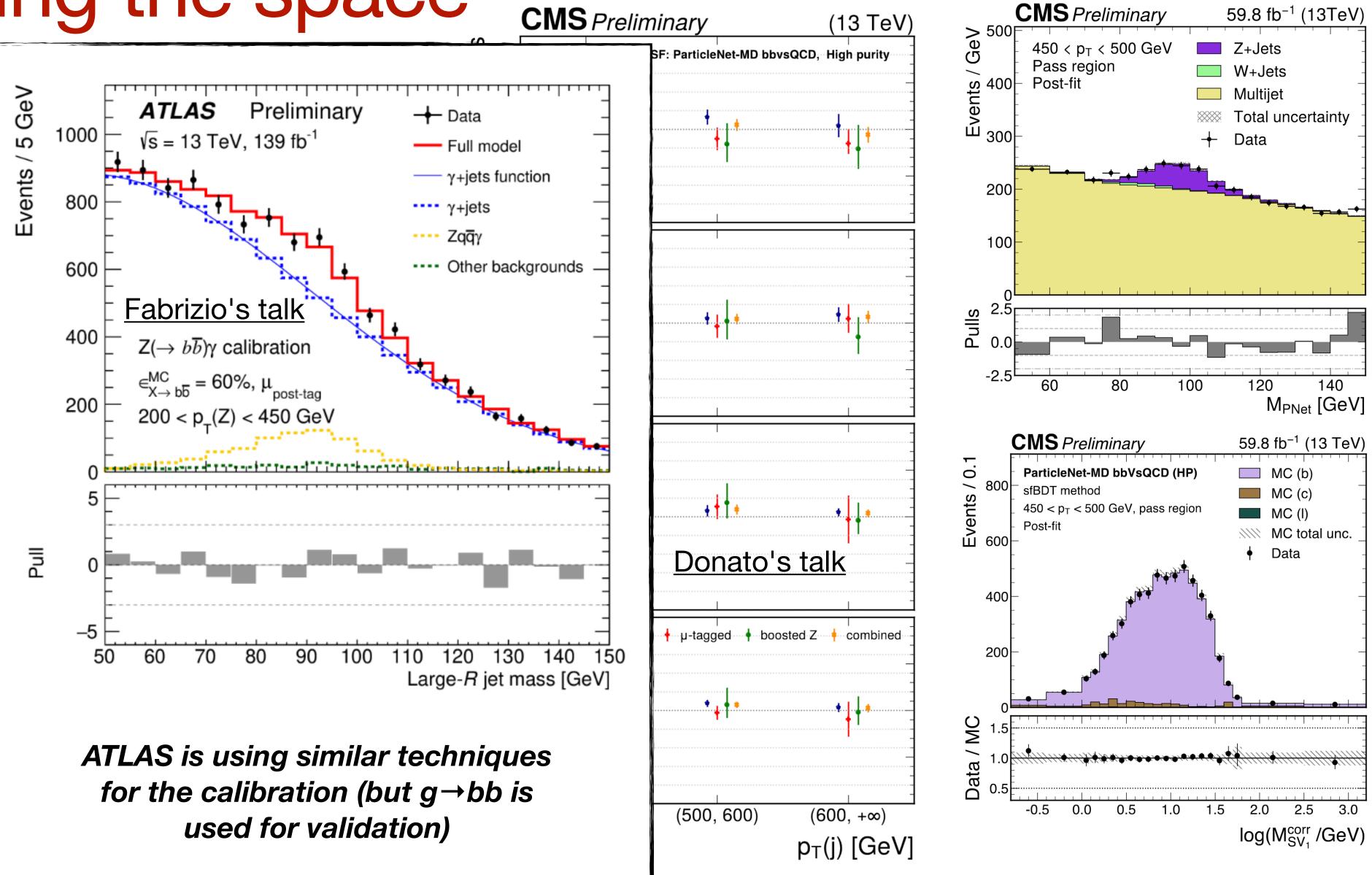




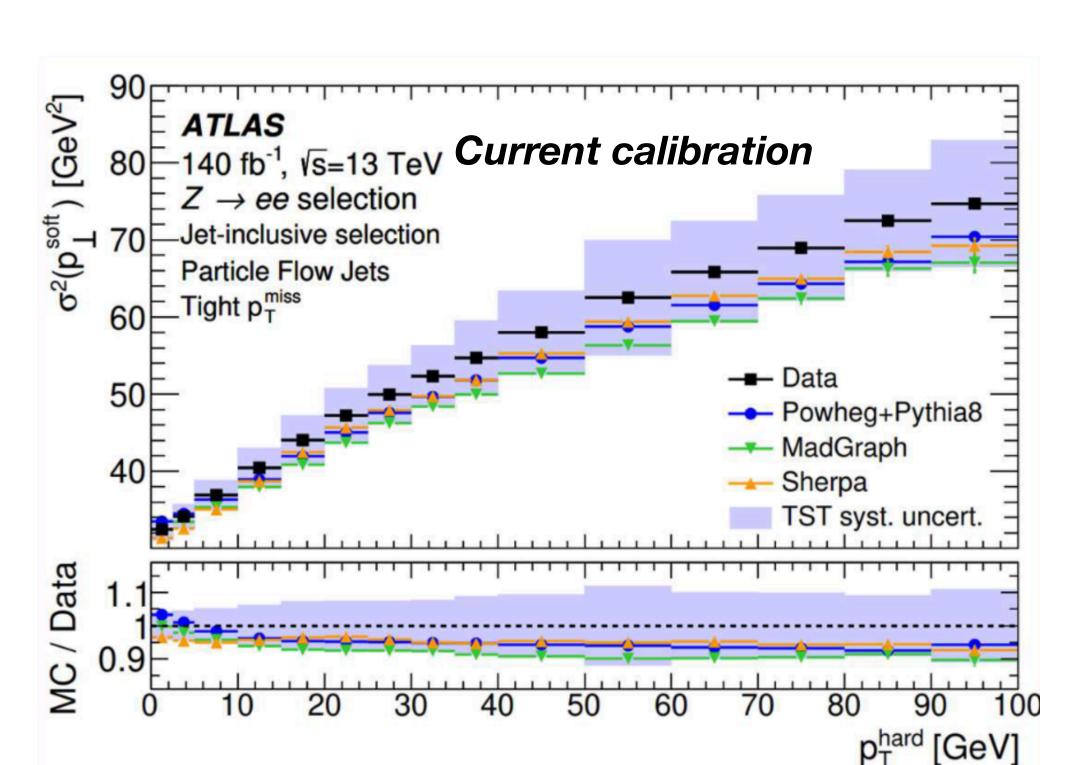


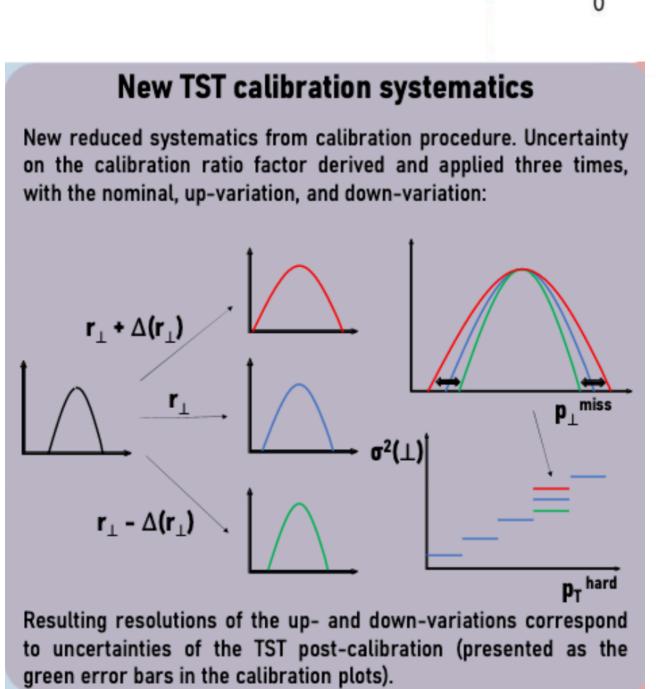
log(M_{SV1} /GeV)

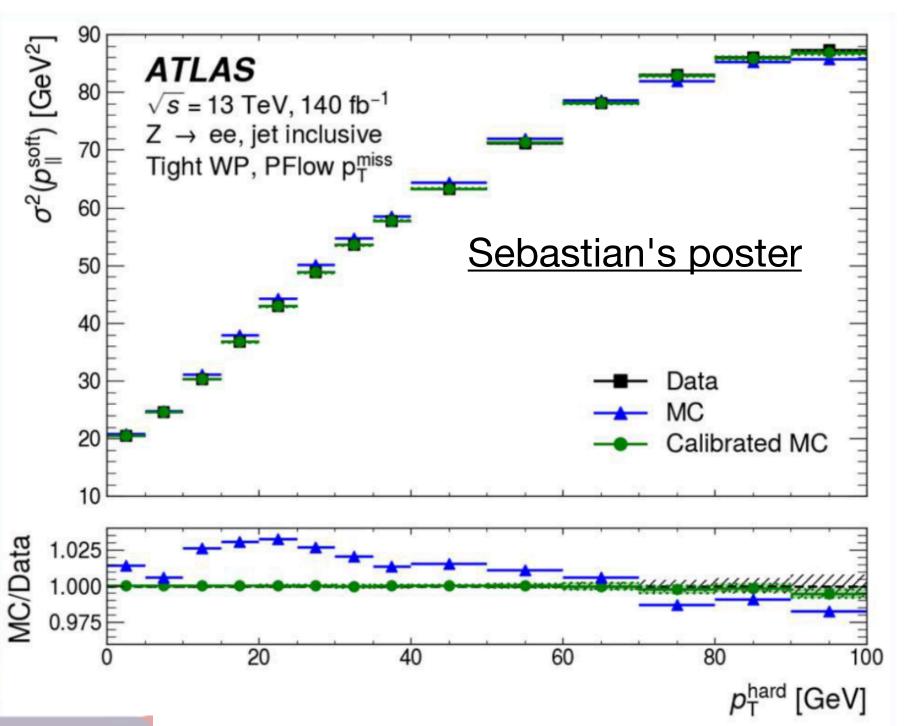
- Need tagger calik
- Difficult to get for H->bb)
- Combining scale methods to get the
 - Using multiple and Z→bb
 - Good cross-ch are consistent
 - Will improve will



- ATLAS is beginning to calibrate the MET
 - Calibration based on three components of the TST



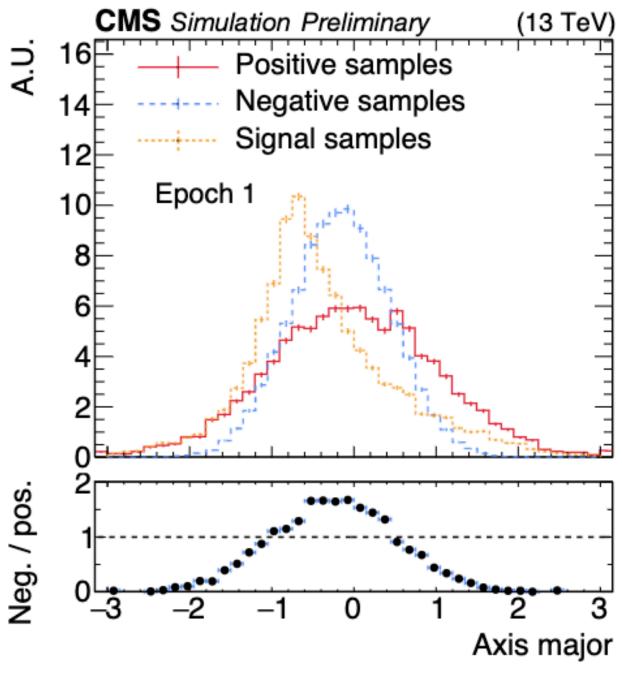


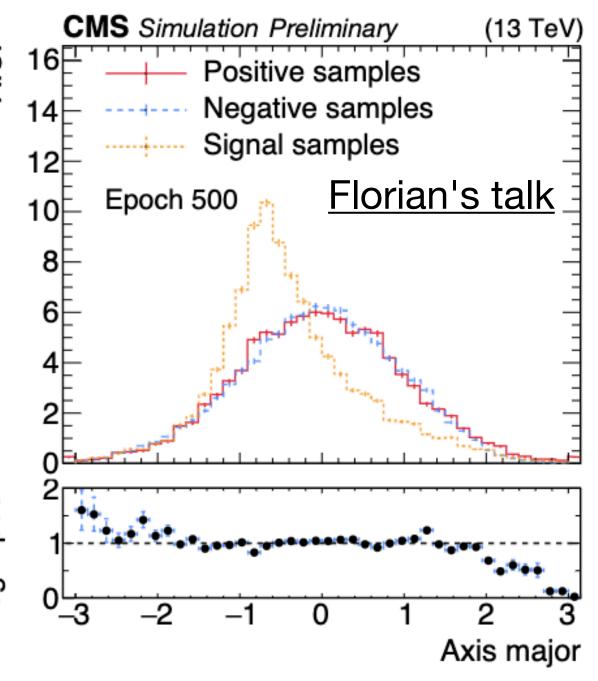


Potential to reduce the dominant systematic from MET!

- Trying to make sure we can control the behavior of the NNs that we train
 - If we know their limitations, can often account for this (with some cleverness)



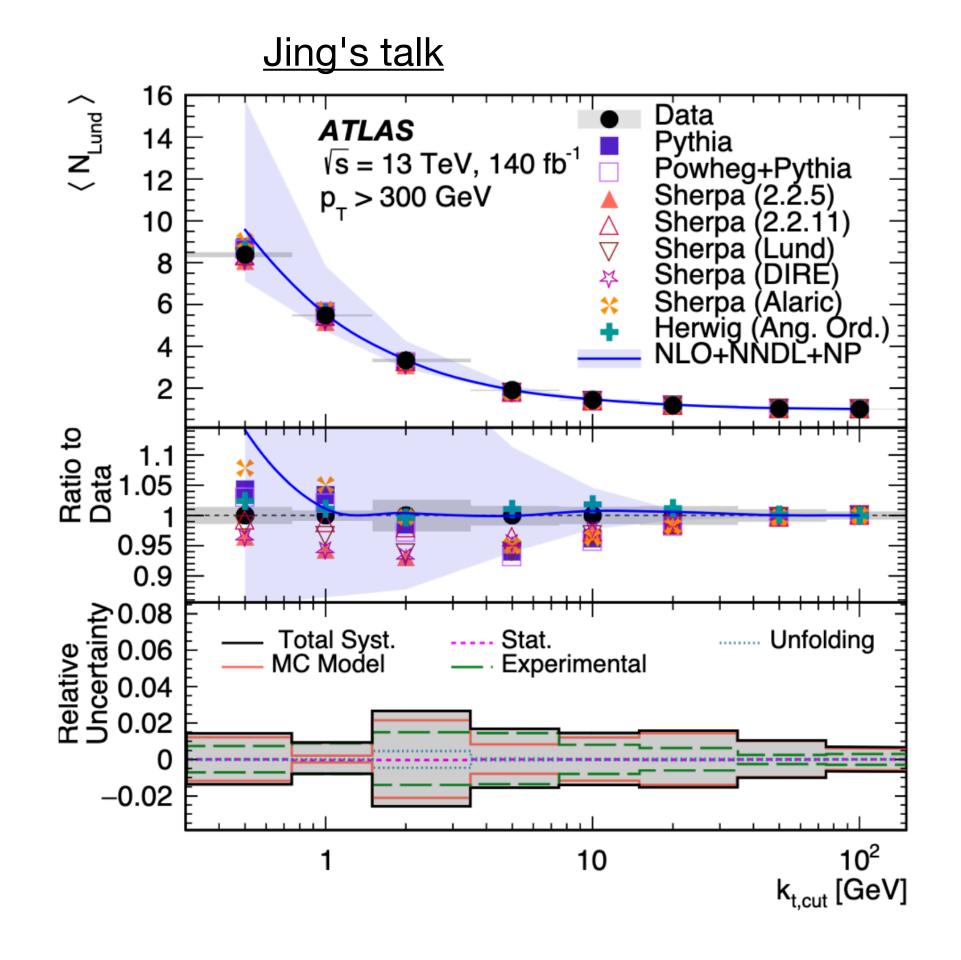


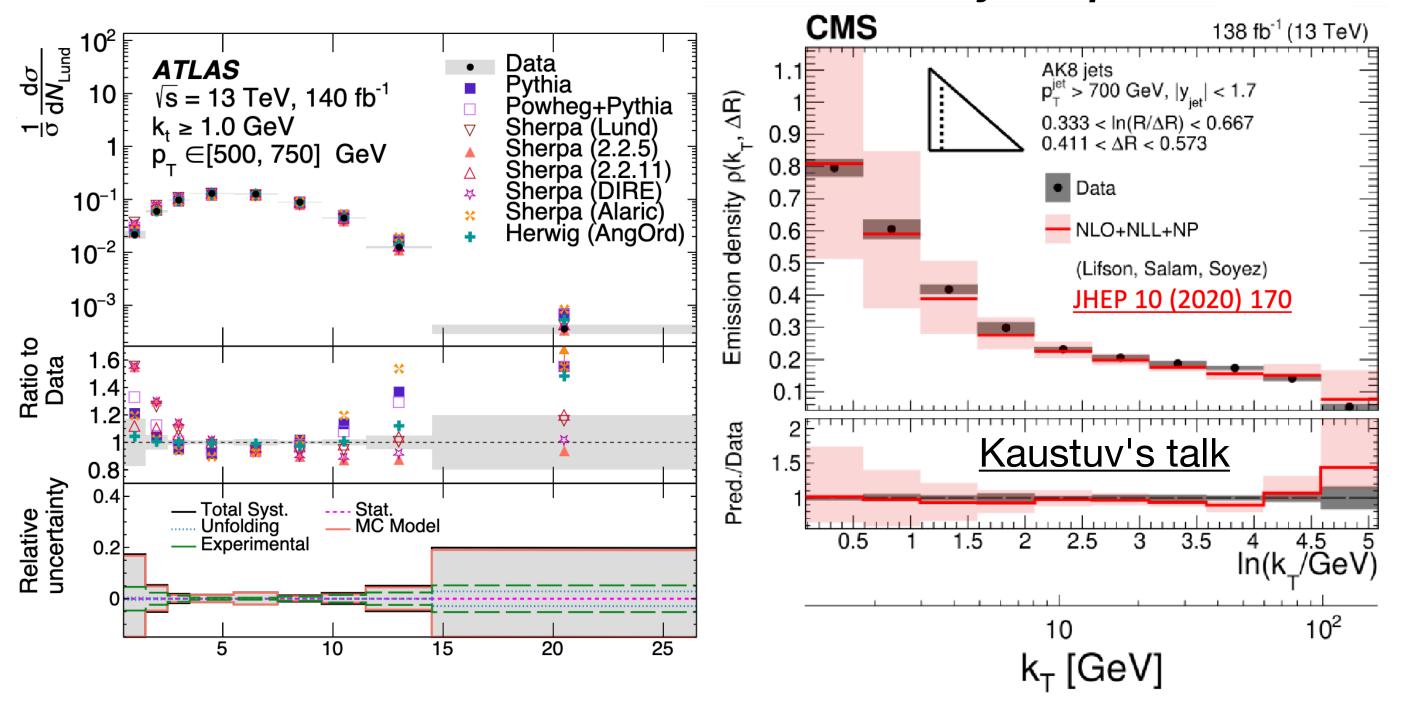


Able to compare to analytical predictions!

Understanding the space

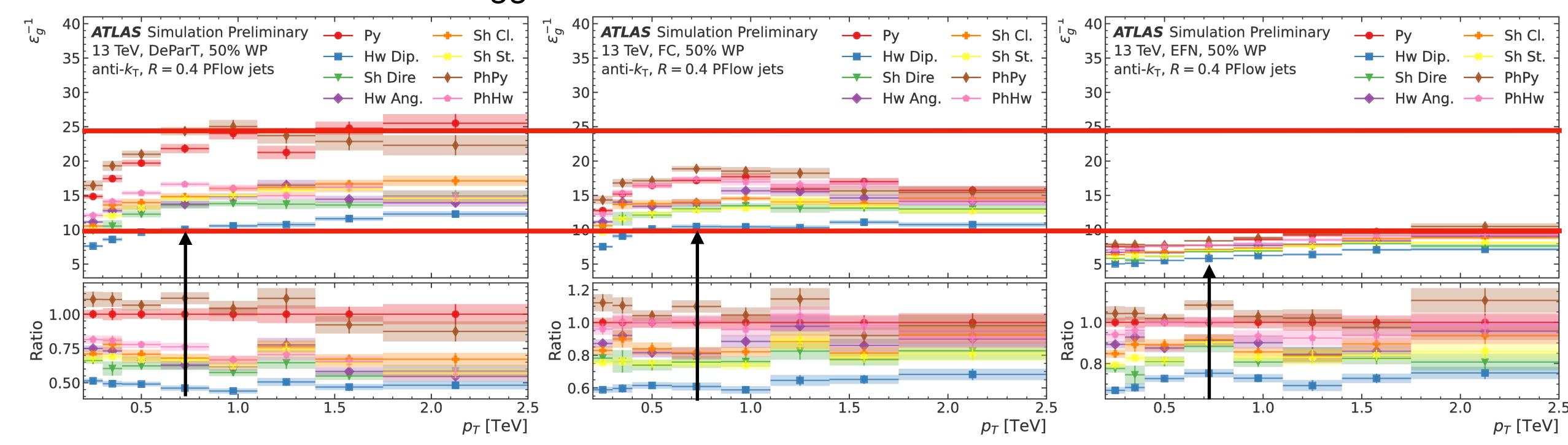
The Lund plane provides a window into a variety of QCD dynamics





- The Lund multiplicity is sensitive to NNDL effects, which are being included in higher accuracy MC predictions
 - Providing experimental tests of new theoretical calculations

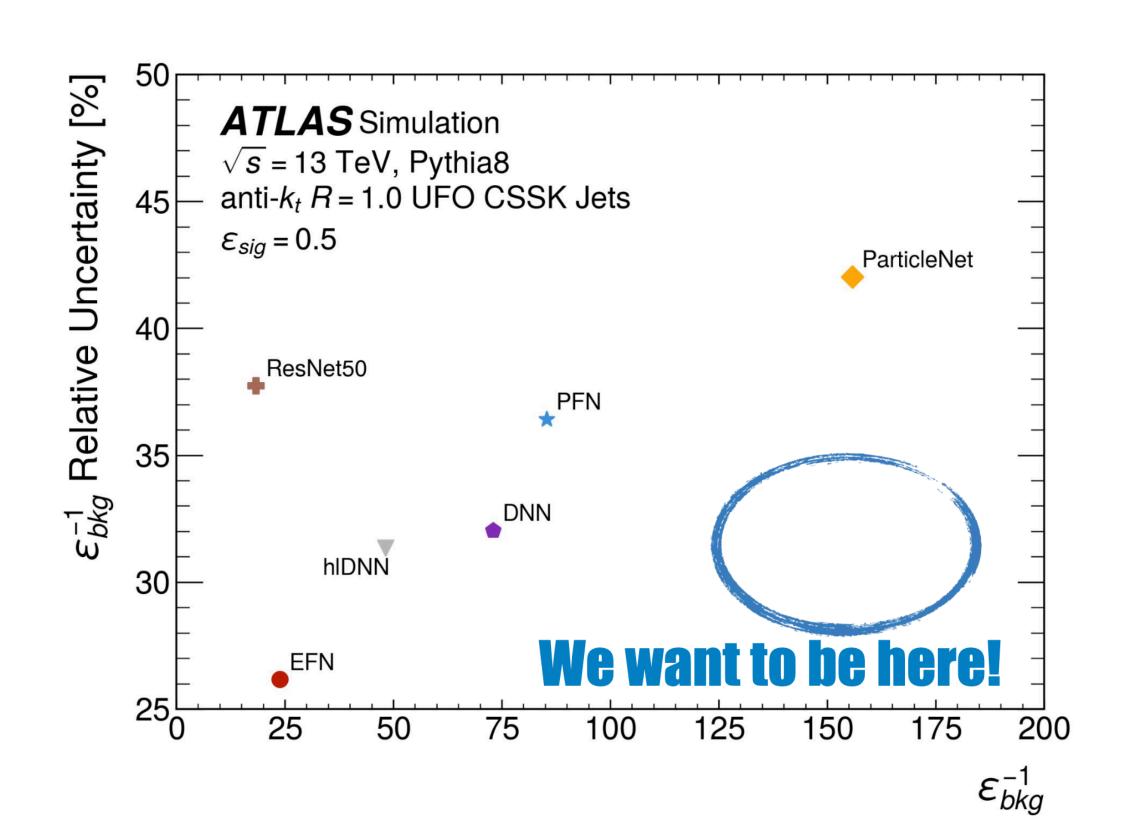
- Many discussions about the robustness of different ML-based taggers
- May be fine to sacrifice larger uncertainties to improve tagging performance
 - Understanding where these differences come from could help us reduce modeling uncertainties for future taggers

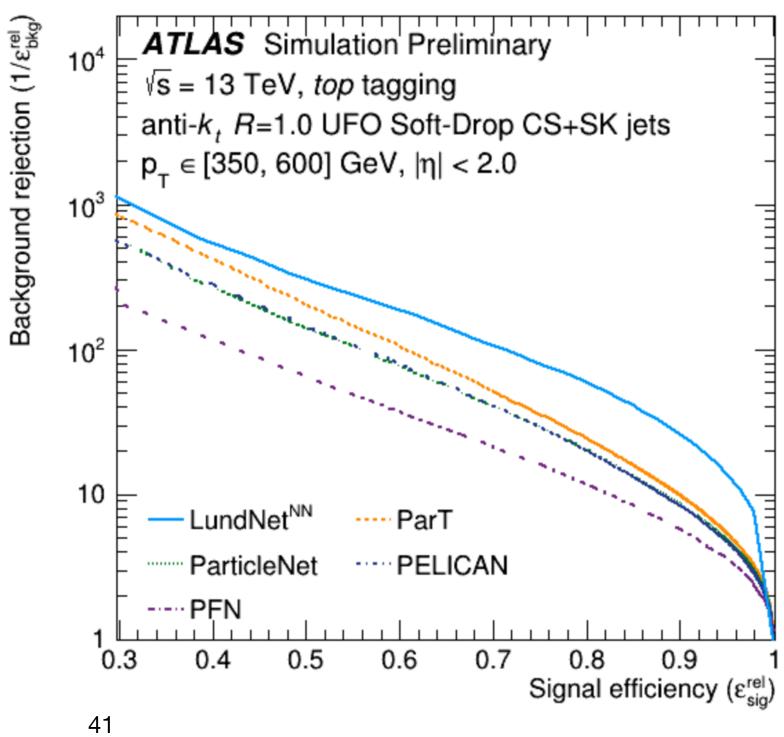


- Still working to understand which networks are robust against modeling effects (and why)
 - Can no longer use a single metric to quantify the performance of a tagger

<u>Kevin's poster</u>

Have an idea for controlling uncertainties? These datasets are public!









The Macro

Can we create algorithms and observables are broadly optimal?

The Micro

Can we expand to new applications by making targeted solutions?

BOOST

- 2010: These aren't your daddy's jets
- ► 2011: "First" data
- 2012: Kids in a candy store
- 2013: Bringing substructure into the mainstream
- 2014: if you ain't boostin' you ain't livin'
- 2015: What a difference five years makes
- 2016: I got 99 problems but my BOOST ain't one
- 2017: Deep thinking jets, they are among us
- 2018: DeepBOOST
- 2019: If you ain't boostin' in the morning, go back to bed!
- 2020: Jet vettin' without jet settin'
- 2021: Jet different
- 2022: we are all about that boost (no treble)
- 2023: Through BOOST, all things are possible (so jot that down)
- 2024: BOOST for all, and all for BOOST