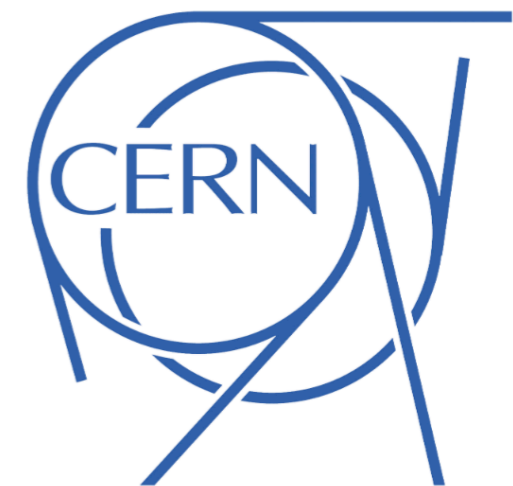


Boosting sensitivity in searches for heavy resonances with the ATLAS detector

CPAN 2023

Josu Cantero (UV/IFIC)

October 2nd, 2023

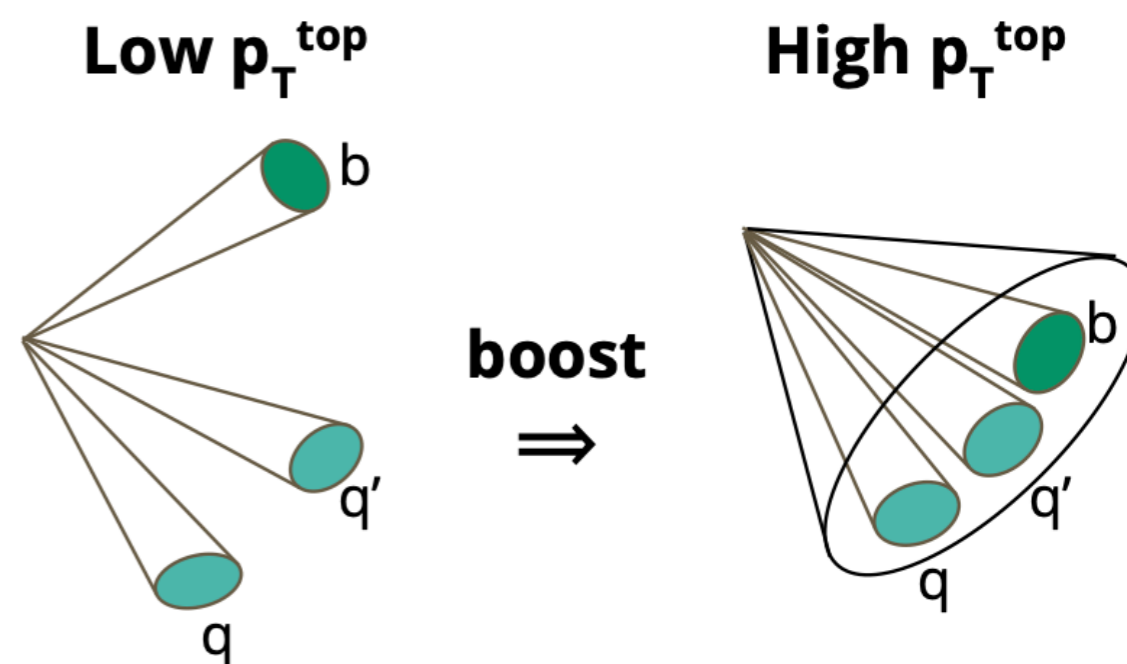
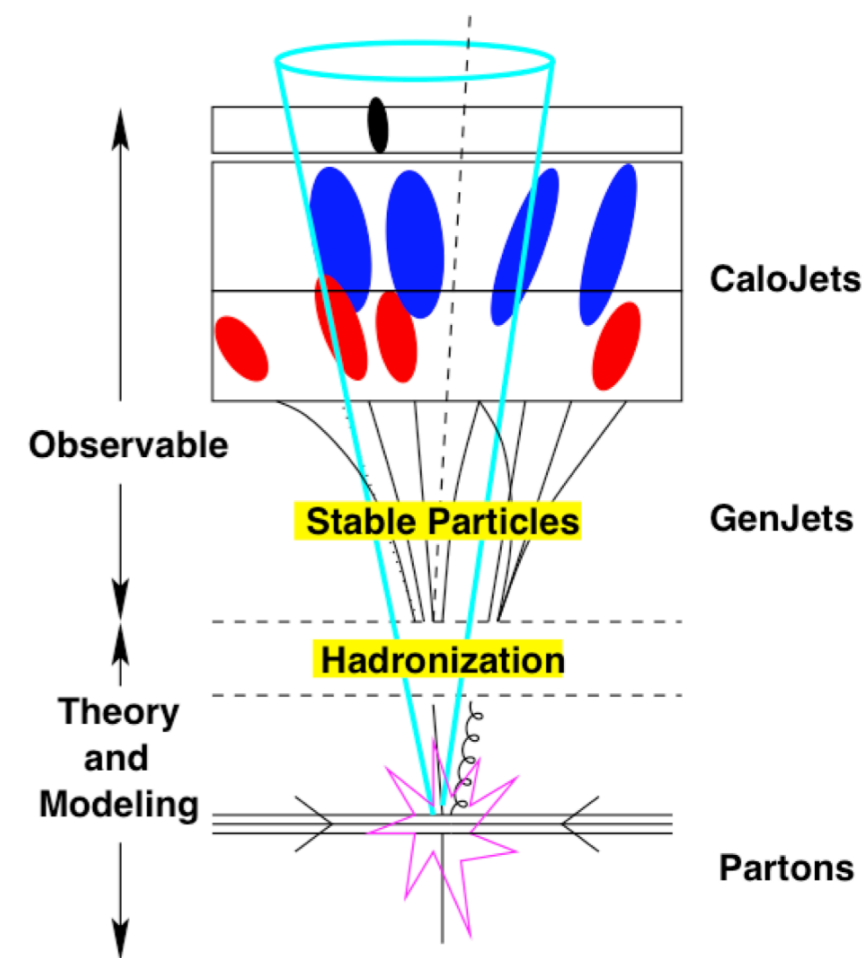


VNIVERSITAT
ID VALÈNCIA



Introduction

- What is a jet?
 - ▶ Collimated bunches of stable particles.
 - ▶ Jets are reconstructed using jet algorithms
 - ▶ Sequential Clustering algorithms: k_T , $\text{anti-}k_T$, CA.
 - ▶ These algorithms depend on one parameter: R (usually called jet radius)
- Jets are typically produced by light quarks/gluons created in the hard interaction or in hadronic decays of heavy resonances: $W, Z \rightarrow q\bar{q}$, $H \rightarrow b\bar{b}$, $t \rightarrow qq'b$.
 - ▶ If R sufficiently large, the decay products can be reconstructed in one jet: $R \sim 2m/p_T$.
 - ▶ $R = 1$ is used in ATLAS.
 - ▶ The so-called large- R jets.
- **Jet tagging** strategy tries to identify which particle originated the jet using the information contained in the jet.
 - ▶ This is typically quantified by means of **jet substructure variables**.



Jet tagging

- Jet tagging very useful in searches including massive exotic resonances decaying into SM massive bosons or t/\bar{t} quarks.
 - ▶ Extensively used nowadays in searches at LHC.
 - ▶ Higher \sqrt{s} increases the fraction of boosted objects.
 - ▶ Increase signal sensitivity searching for massive exotic resonances.
 - ▶ High- p_T SM resonances reconstructed as single large- R jets.
 - ▶ Jet tagging to discriminate between signal-like final states and background where jets are originated from light quarks and gluons.

Phys. Rev. Lett 100 (2008) 242001

Jet substructure as a new Higgs search channel at the LHC

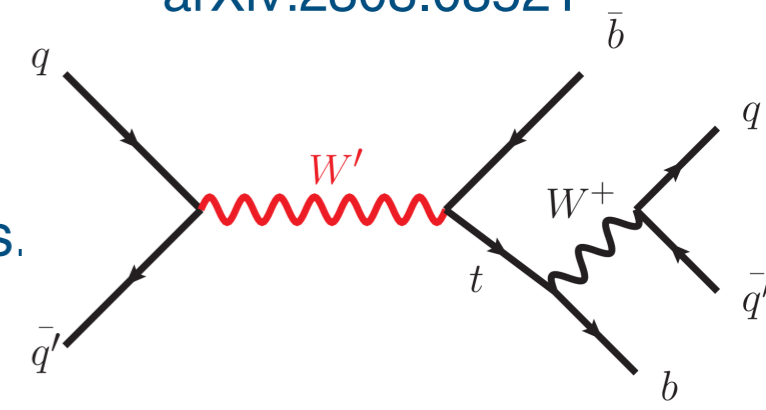
Jonathan M. Butterworth, Adam R. Davison
 Department of Physics & Astronomy, University College London.

Mathieu Rubin, Gavin P. Salam
 LPTHE; UPMC Univ. Paris 6; Univ. Denis Diderot; CNRS UMR 7589; Paris, France.

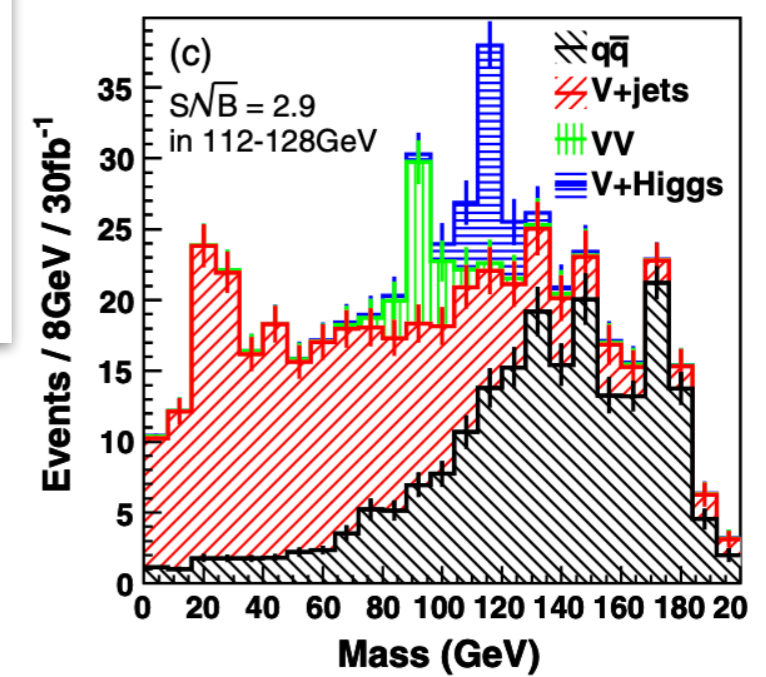
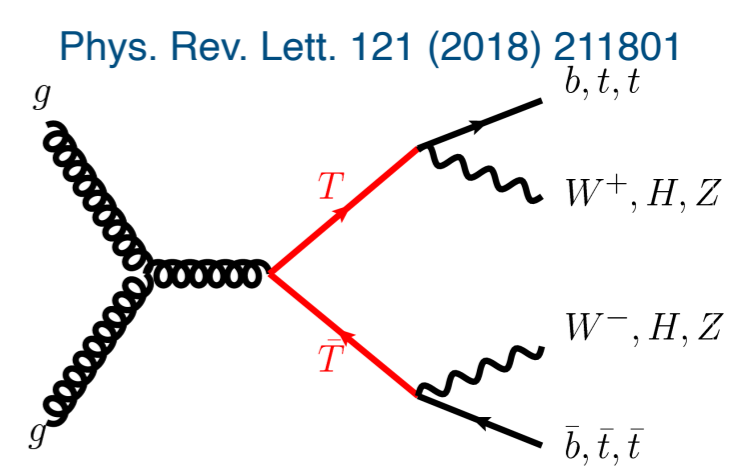
It is widely considered that, for Higgs boson searches at the Large Hadron Collider, WH and ZH production where the Higgs boson decays to $b\bar{b}$ are poor search channels due to large backgrounds. We show that at high transverse momenta, employing state-of-the-art jet reconstruction and decomposition techniques, these processes can be recovered as promising search channels for the standard model Higgs boson around 120 GeV in mass.

- Already proposed in 2008 to increase sensitivity of VH channel using $H \rightarrow b\bar{b}$ decays.

W' production
 arXiv:2308.08521



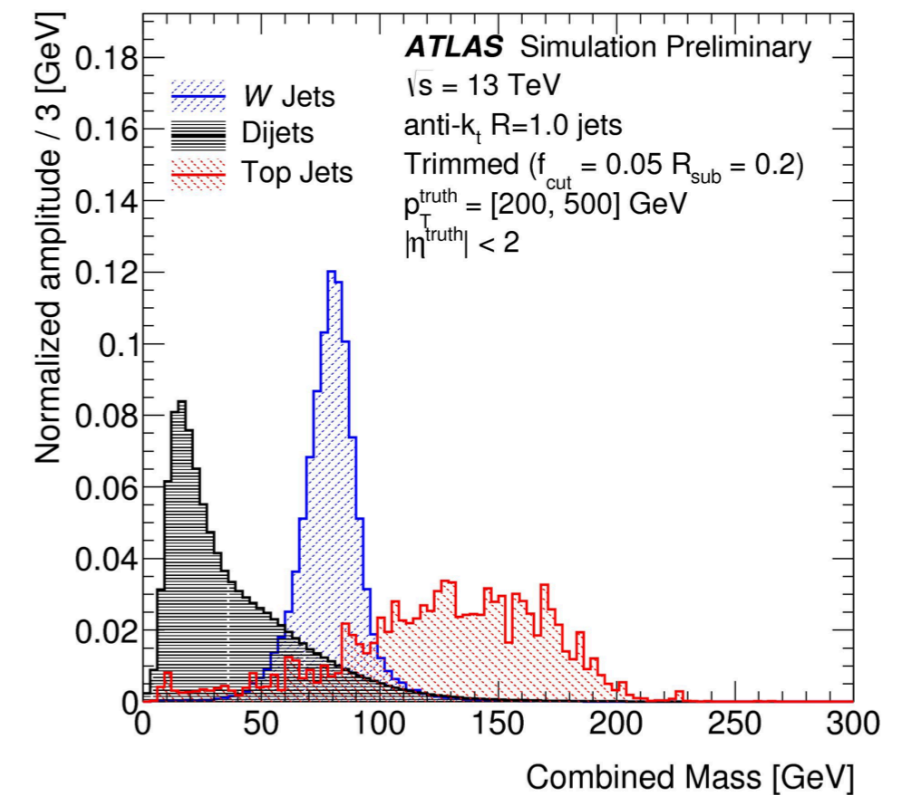
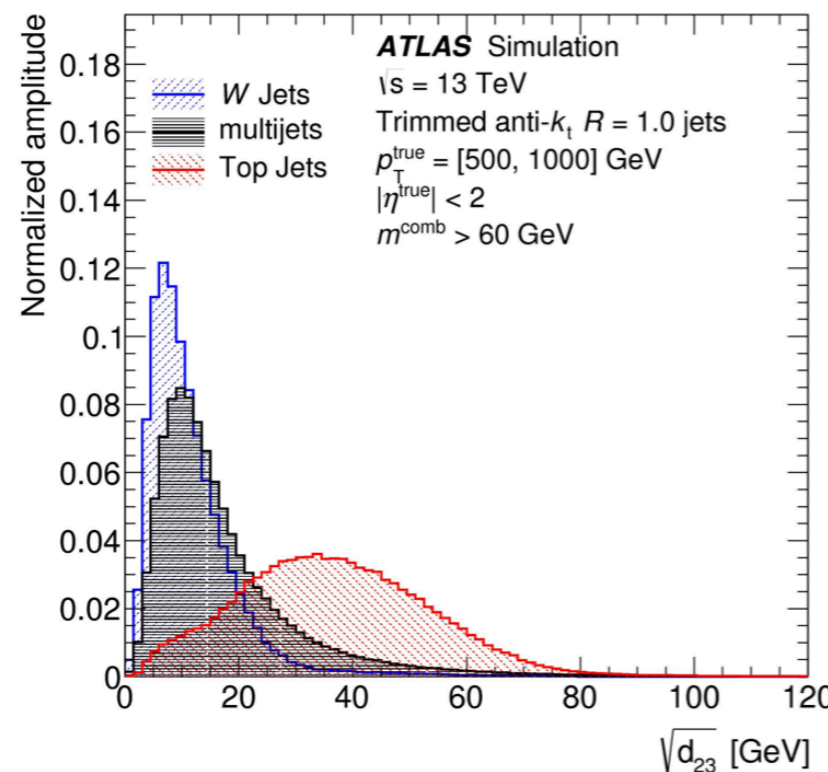
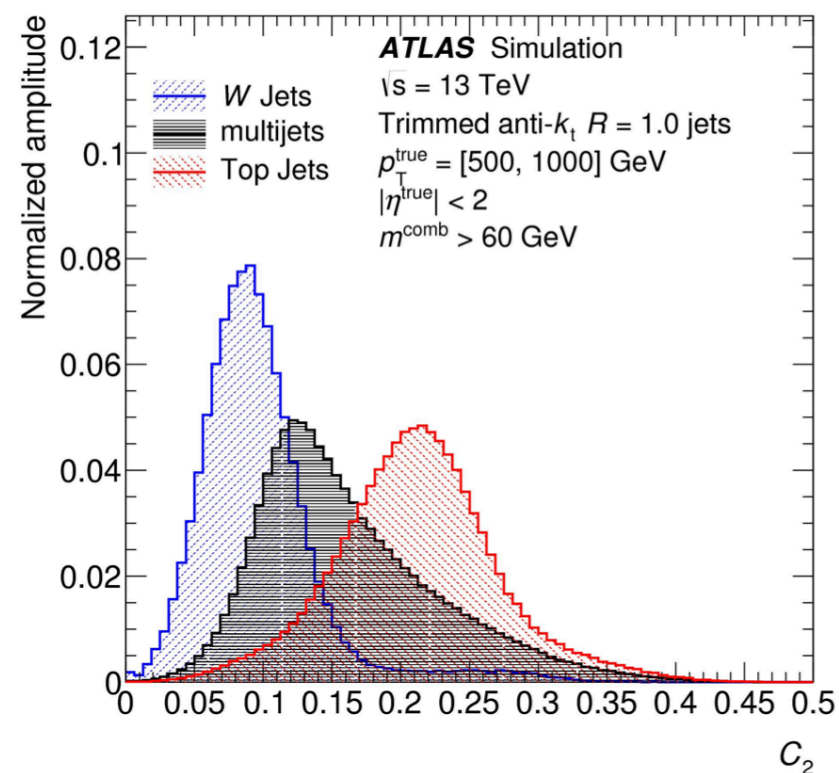
VLQ production
 Phys. Rev. Lett. 121 (2018) 211801



Jet substructure

- Jets originated from **boosted resonances** decay have **distinctive substructure** compared to jets originated from light quarks/gluons.
 - Sharp mass peaks/harder splitting scales.
 - Subjet multiplicity/b-tagging of large- R jet.
 - Color reconnection/Energy-Energy correlation variables.

Eur. Phys. J. C 79 (2019) 375



- Some jet substructure variables:

- Splitting scales** ($\sqrt{d_{12}}, \sqrt{d_{23}}$): scale of the $2 \rightarrow 1$ and $3 \rightarrow 2$ recombination step in a k_t clustering sequence.
- Jet mass**: distribution expected to peak at the resonance mass.
- Energy-Energy correlators**: weighted average of the angular moment of the energy deposits in a jet.

Dimensionless C_2^β and D_2^β quite useful to identify two-body structure.

EECs:

$$E_{CF0}(\beta) = 1,$$

$$E_{CF1}(\beta) = \sum_{i \in J} p_{T_i},$$

$$E_{CF2}(\beta) = \sum_{i < j \in J} p_{T_i} p_{T_j} (\Delta R_{ij})^\beta,$$

$$E_{CF3}(\beta) = \sum_{i < j < k \in J} p_{T_i} p_{T_j} p_{T_k} (\Delta R_{ij} \Delta R_{ik} \Delta R_{jk})^\beta$$

$$e_2^{(\beta)} = \frac{E_{CF2}(\beta)}{E_{CF1}(\beta)^2},$$

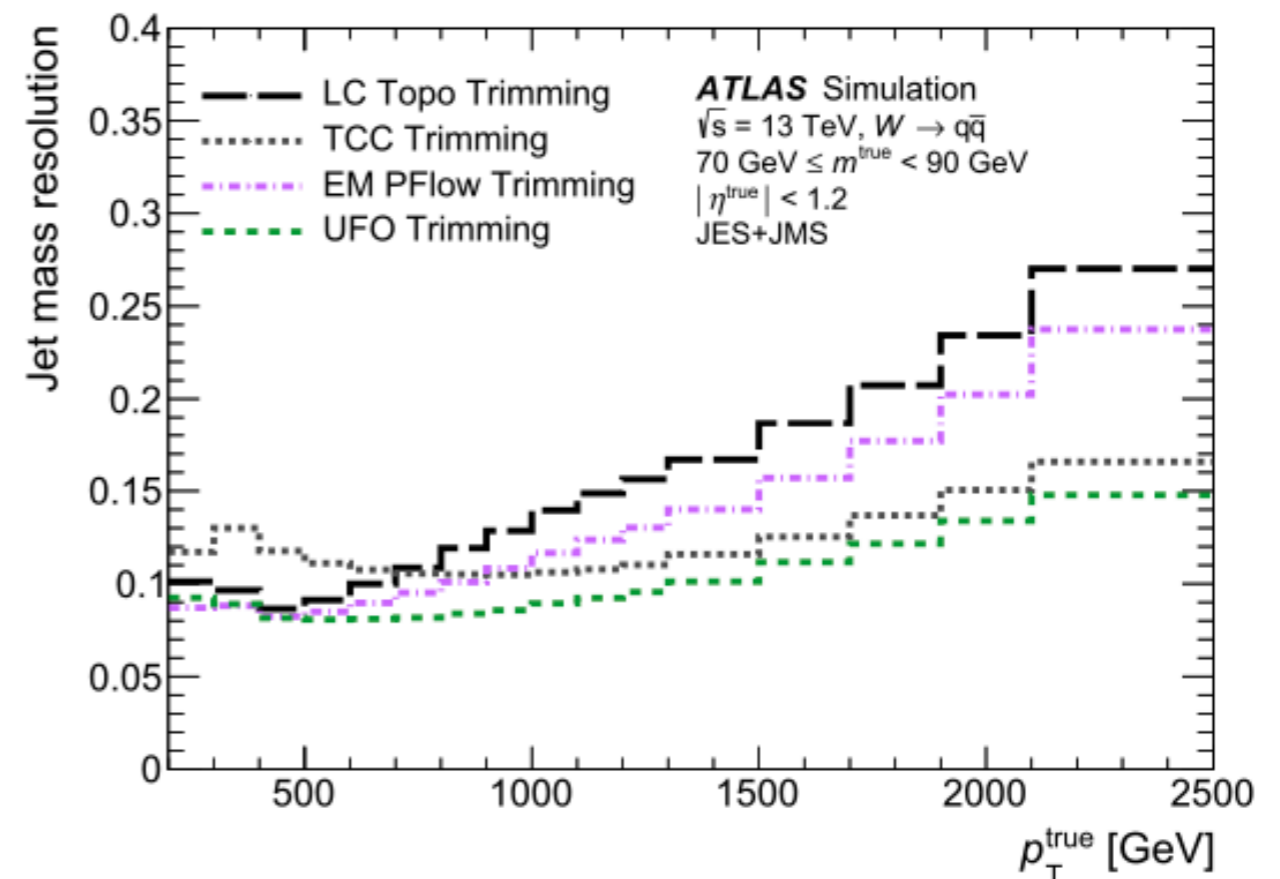
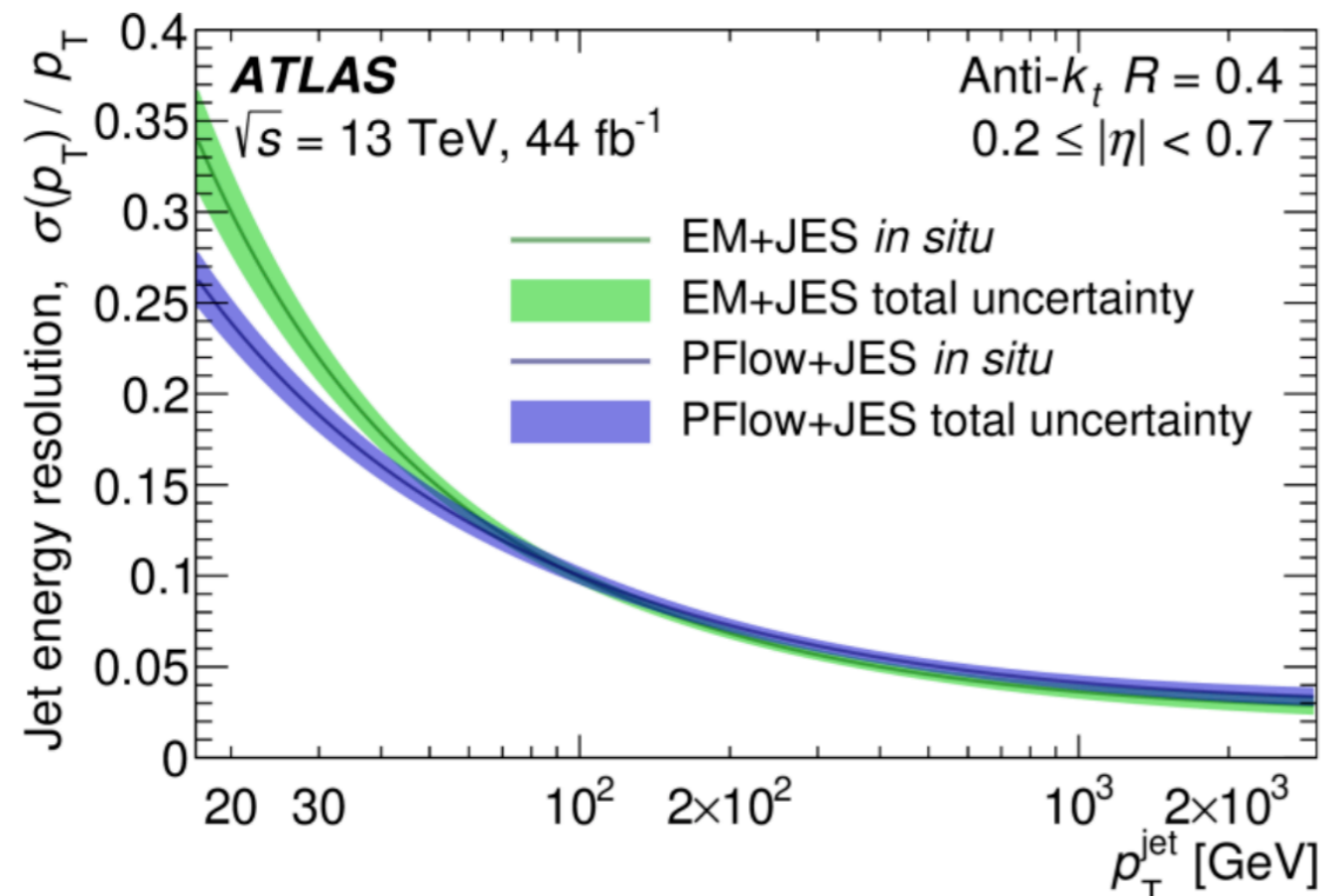
$$e_3^{(\beta)} = \frac{E_{CF3}(\beta)}{E_{CF1}(\beta)^3}.$$

$$C_2^{(\beta)} = \frac{e_3^{(\beta)}}{(e_2^{(\beta)})^2},$$

$$D_2^{(\beta)} = \frac{e_3^{(\beta)}}{(e_2^{(\beta)})^3}.$$

Large- R jet collections

- From experimental point of view jet constituents can be defined in different ways:
 - LCTopo (Local Calibrated Topoclusters)**: 3D clusters of noise-suppressed calorimeter cells.
 - PFlow (Particle Flow Objects)**: combine tracking and calorimeter information.
 - Improve performance/pile-up stability for low p_T jets. [Eur. Phys. J. C 77 \(2017\) 466](#)
 - High p_T tracks excluded (worse energy resolution than clusters).
 - TCC (Track to Calo Cluster)**: tracking and calorimeter information combined in preference to tracking information (i.e. η , ϕ information coming purely from tracks). [ATL-PHYS-PUB-2017-015](#)
 - Very good at high p_T .
 - Use calorimeter energy scale and tracker spatial coordinates.
 - UFO (Unified Flow Objects)**: merge PFlow and TCC for better performance in a wider p_T range. [Eur. Phys. J. C 81 \(2021\) 334](#)

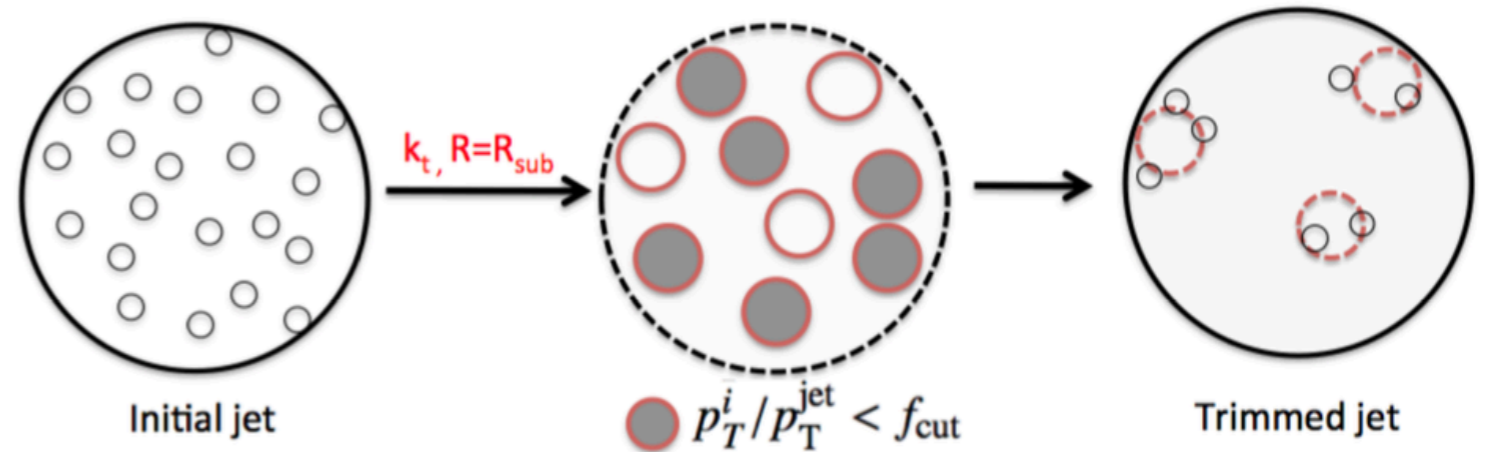


Large- R jet grooming

- Use to reduce sensitivity to pile-up and unrelated radiation.
 - ▶ Improve response ($O_{\text{reco}}/O_{\text{truth}}$) and pile-up stability of jet substructure variables.
- Several groomers in the market.
 - ▶ **Trimming and SoftDrop used in ATLAS.**

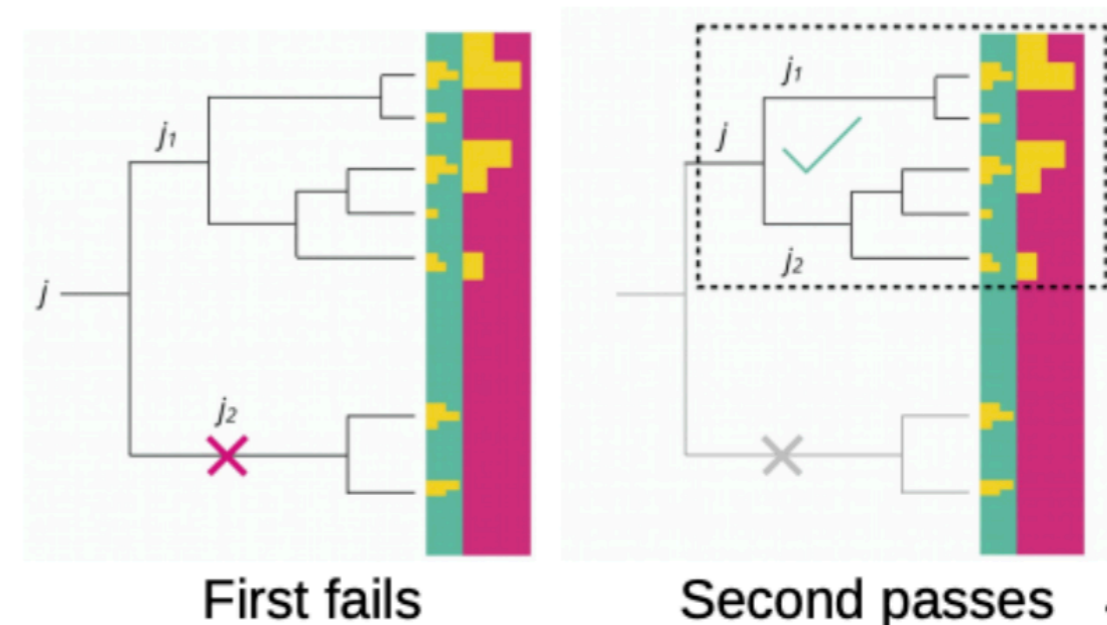
Trimming

- Remove low- p_T contributions at the subjet level.



Soft-Drop

- Define splittings using CA declustering.
- Apply SoftDrop:
 - ▶ Start from first jet splitting.
 - ▶ Reach passing condition for both splittings.



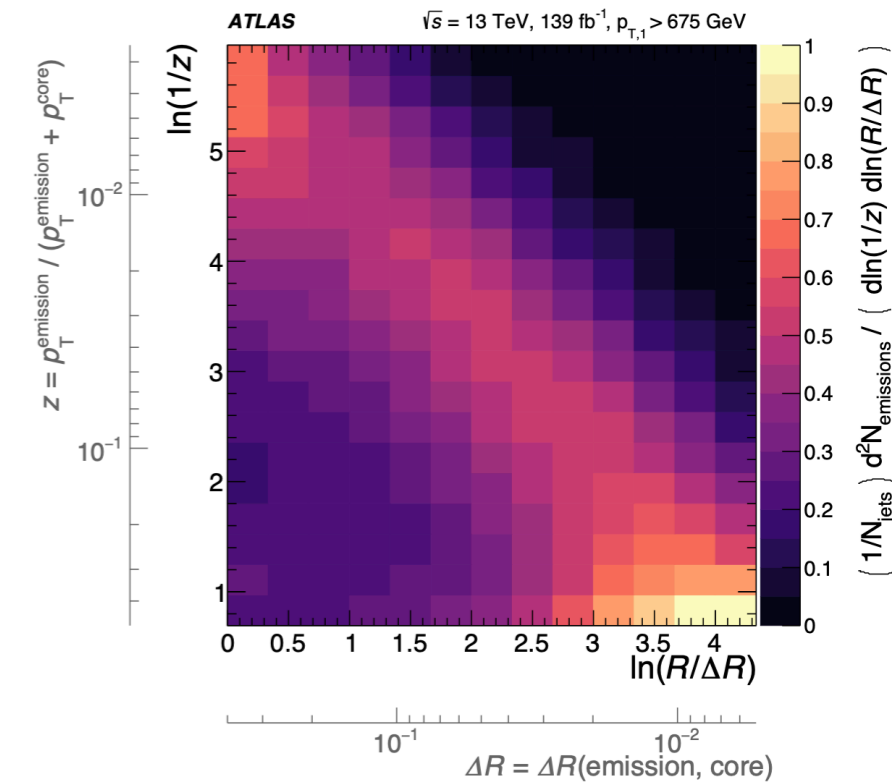
$$\frac{\min(p_{T1}, p_{T2})}{p_{T1} + p_{T2}} > z_{\text{cut}} \left(\frac{\Delta R_{12}}{R_0} \right)^\beta$$

➡ Keep both
➡ "Drop" constituent

Jet tagging techniques

- There are different ways to build a tagger.
 - **Moment-based taggers:**
 - ▶ Identify **jet substructure moments** with good separation power and apply cuts on them.
 - ▶ This can be done sequentially: **3-var W/Z tagger**.
 - ▶ On a NN score: **ANN W/Z tagger, DNN top tagger**.
 - **Declustering taggers:**
 - ▶ Attempt to reconstruct the shower history of the jet.
 - ▶ **Shower deconstruction:** quantify if jet declustering is signal or background like: **SD top tagger**.
 - ▶ **Lund plane based taggers:** use Lund-plane information to discriminate between signal and background jets.
 - ▶ NNs allow to do it quite efficiently i.e GNNs
 - **Constituent based taggers:**
 - ▶ Use jet constituent information to feed a NN to discriminate between signal and bkg jets.
 - ▶ Large number of parameters is typically needed.

[Phys. Rev. Lett. 124 \(2020\) 222002](#)



[arXiv:1902.09914](#)

	AUC	Acc	$1/\epsilon_B (\epsilon_S = 0.3)$			#Param
			single	mean	median	
CNN [16]	0.981	0.930	914±14	995±15	975±18	610k
ResNeXt [31]	0.984	0.936	1122±47	1270±28	1286±31	1.46M
TopoDNN [18]	0.972	0.916	295±5	382±5	378±8	59k
Multi-body N -subjettiness 6 [24]	0.979	0.922	792±18	798±12	808±13	57k
Multi-body N -subjettiness 8 [24]	0.981	0.929	867±15	918±20	926±18	58k
TreeNiN [43]	0.982	0.933	1025±11	1202±23	1188±24	34k
P-CNN	0.980	0.930	732±24	845±13	834±14	348k
ParticleNet [47]	0.985	0.938	1298±46	1412±45	1393±41	498k
LBN [19]	0.981	0.931	836±17	859±67	966±20	705k
LoLa [22]	0.980	0.929	722±17	768±11	765±11	127k
LDA [54]	0.955	0.892	151±0.4	151.5±0.5	151.7±0.4	184k
Energy Flow Polynomials [21]	0.980	0.932	384			1k
Energy Flow Network [23]	0.979	0.927	633±31	729±13	726±11	82k
Particle Flow Network [23]	0.982	0.932	891±18	1063±21	1052±29	82k
GoaT	0.985	0.939	1368±140		1549±208	35k

Jet tagging techniques

- Trimmed LCTopo Large- R jets taggers

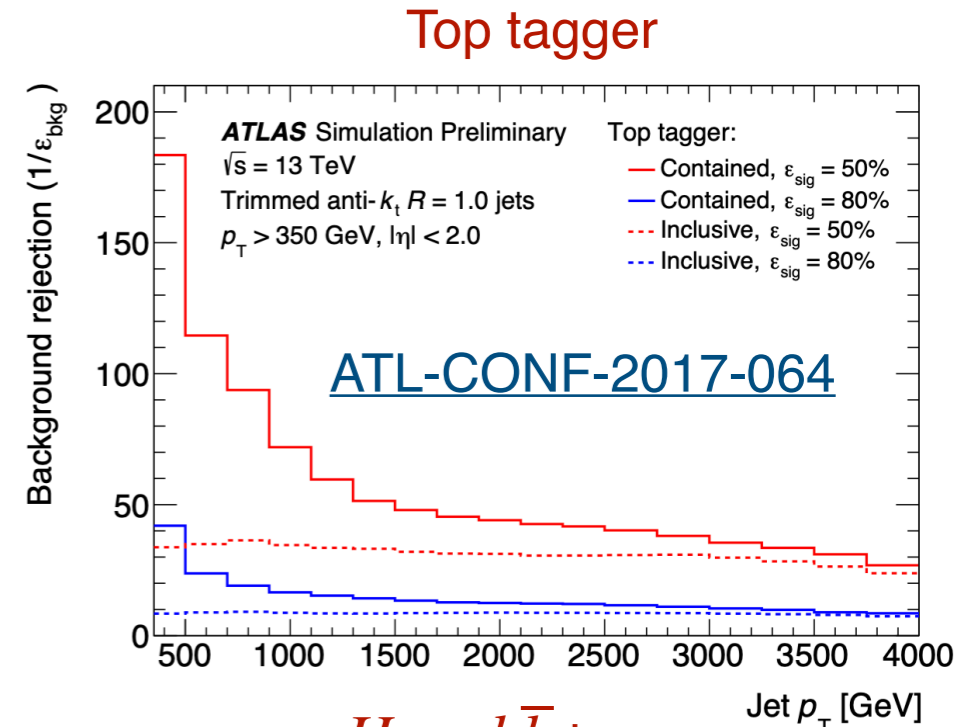
- 3-variable based W and Z taggers: D_2 , N_{trk} and $m_{\text{jet}}^{\text{comb}}$.
 - Also available for TCC Large- R jets.
- DNN top taggers: based on jet moments.
- Xbb tagger: b-tagging info from three leading VR-trackjets feeding a NN.

50% and 80% flat signal efficiency WPs defined

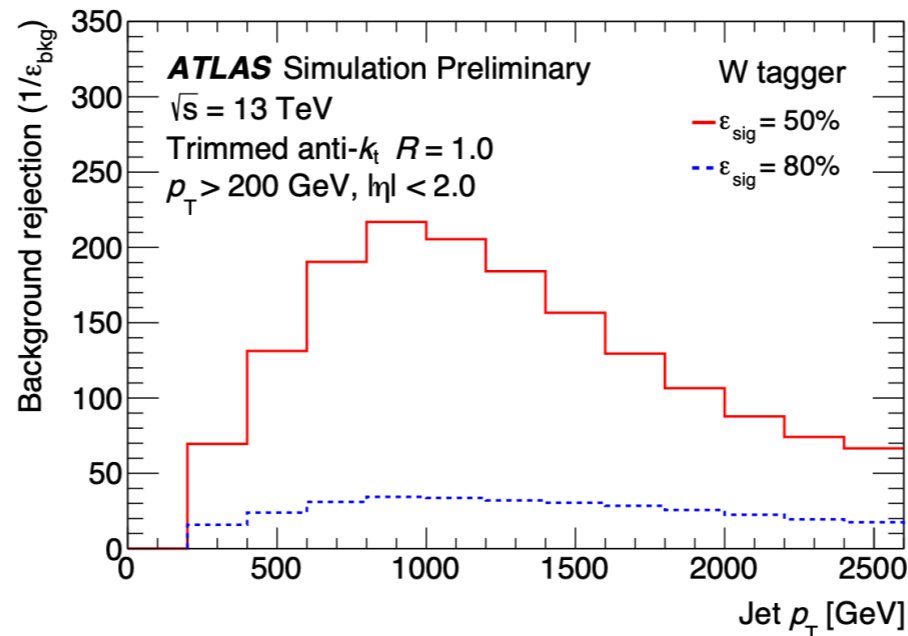
Inputs to DNN top tagger

$$D_{\text{Xbb}} = \ln \frac{p_{\text{Higgs}}}{f_{\text{top}} \cdot p_{\text{top}} + (1 - f_{\text{top}}) \cdot p_{\text{multijet}}}$$

Observable	Variable	Used for
Calibrated jet kinematics	p_T, m^{comb}	top, W
Energy correlation ratios	e_3, C_2, D_2	top, W
N -subjettiness	$\tau_1, \tau_2, \tau_{21}$	top, W
	τ_3, τ_{32}	top
Fox-Wolfram moment	R_2^{FW}	W
Splitting measures	z_{cut}	W
	$\sqrt{d_{12}}$	top, W
	$\sqrt{d_{23}}$	top
Planar flow	\mathcal{P}	W
Angularity	a_3	W
Aplanarity	A	W
KtDR	$KtDR$	W
Qw	Q_w	top

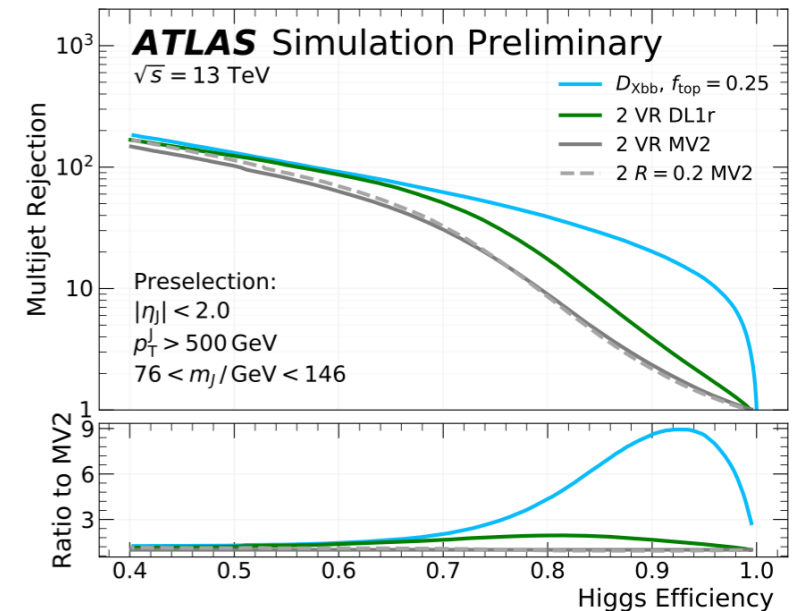


W tagger



$H \rightarrow b\bar{b}$ tagger

ATL-PHYS-PUB-2020-019



Jet tagging techniques

- Differences between data and MC on jet substructure lead to differences on jet tagging performance
 - Taggers need to be calibrated!
 - Calibration is done by means of Scale Factors (SFs).

- The MC efficiency is corrected to data efficiency:

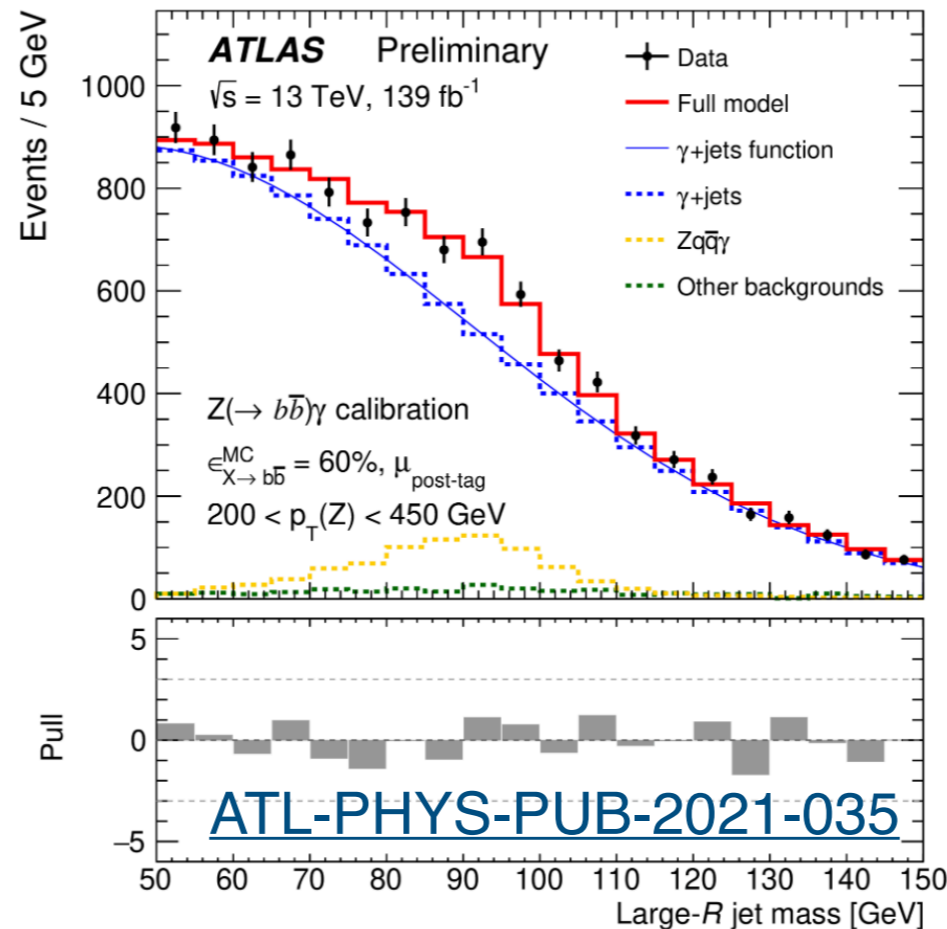
$$SF_{\text{eff}}(p_T) = \epsilon_{\text{data}} / \epsilon_{\text{MC}}$$

- Inefficiency SFs are needed to maintain unitarity:

$$SF_{\text{ineff}}(p_T) = (1 - SF_{\text{eff}} \cdot \epsilon_{\text{MC}}) / (1 - \epsilon_{\text{MC}})$$

- Semileptonic $t\bar{t}$ events to estimate the SFs for top and W taggers.

Xbb tagger 60% WP

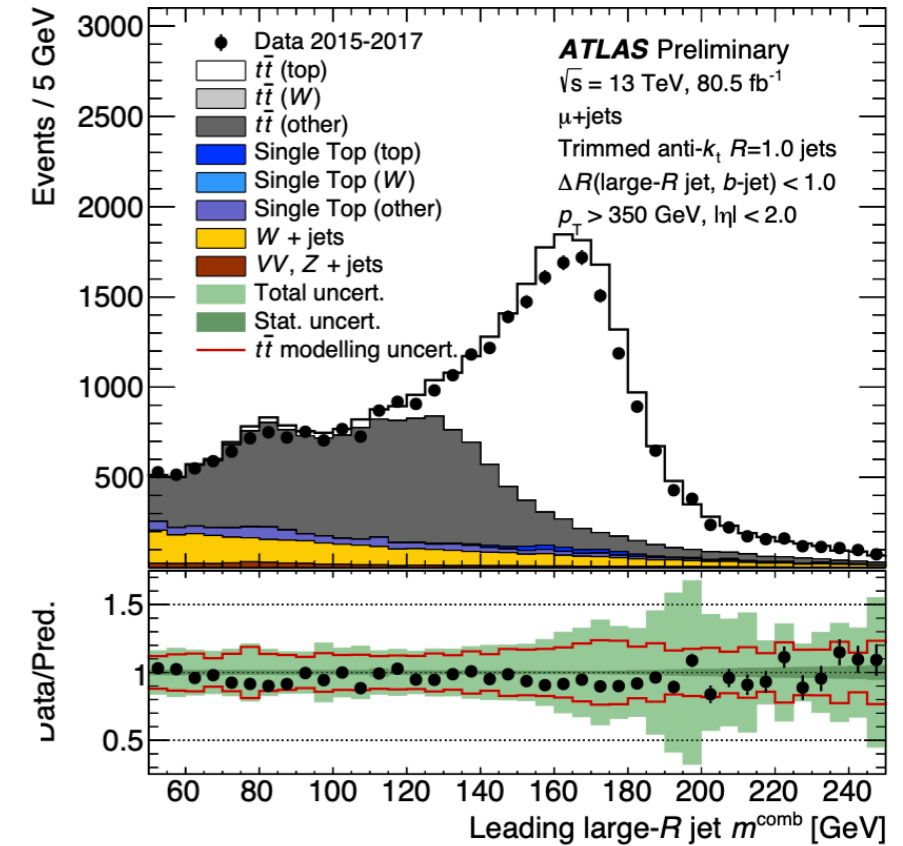


- $Z \rightarrow b\bar{b}$, $g \rightarrow b\bar{b}$ and semileptonic $t\bar{t}$ for Xbb tagger.

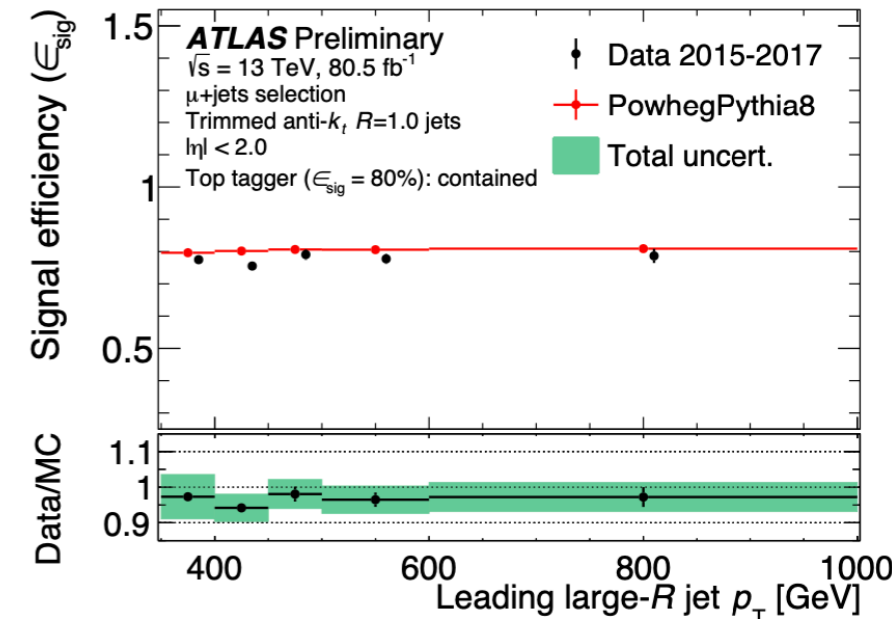
- Multijet and γ +jets events for background SFs in $200 \text{ GeV} \lesssim p_T \lesssim 3 \text{ TeV}$.

- MC based high- p_T extrapolation uncertainties

ATL-PHYS-PUB-2020-017



Top tagger 80% WP

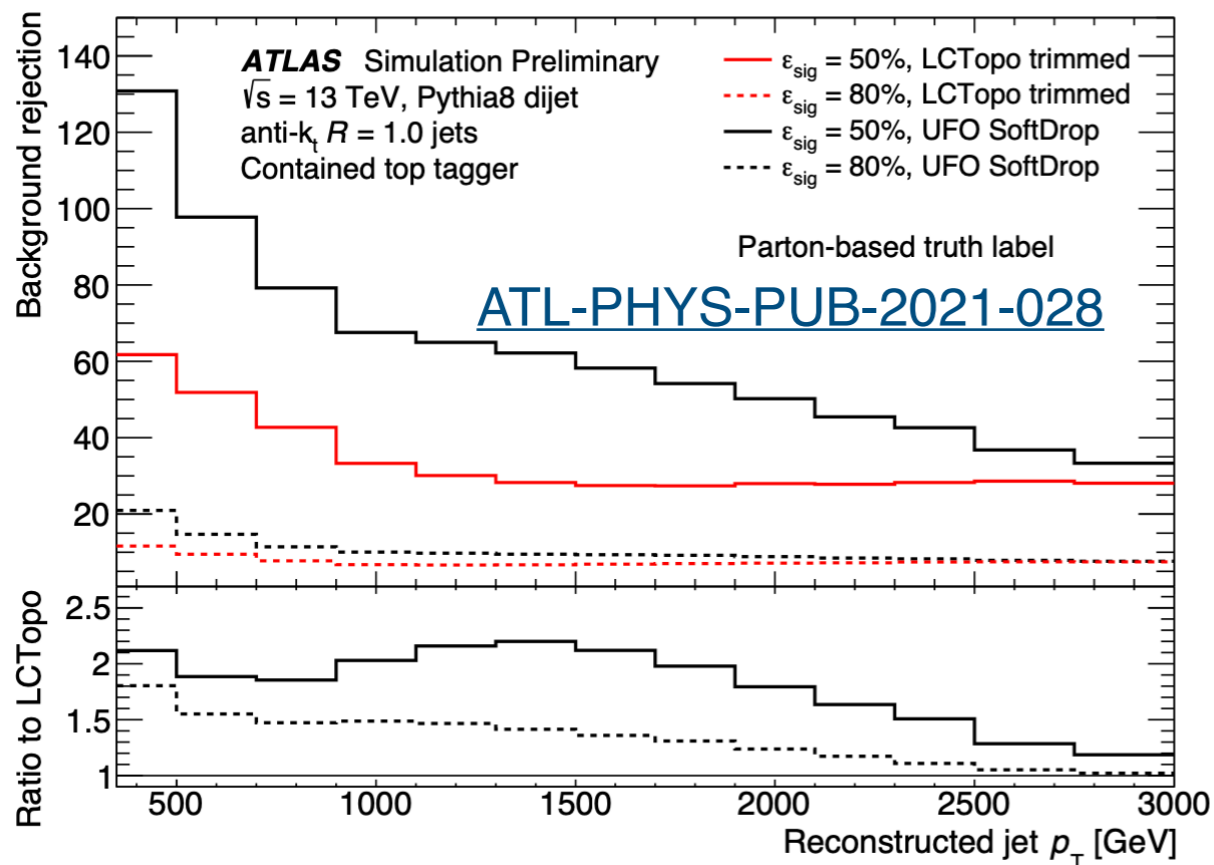


Jet tagging techniques

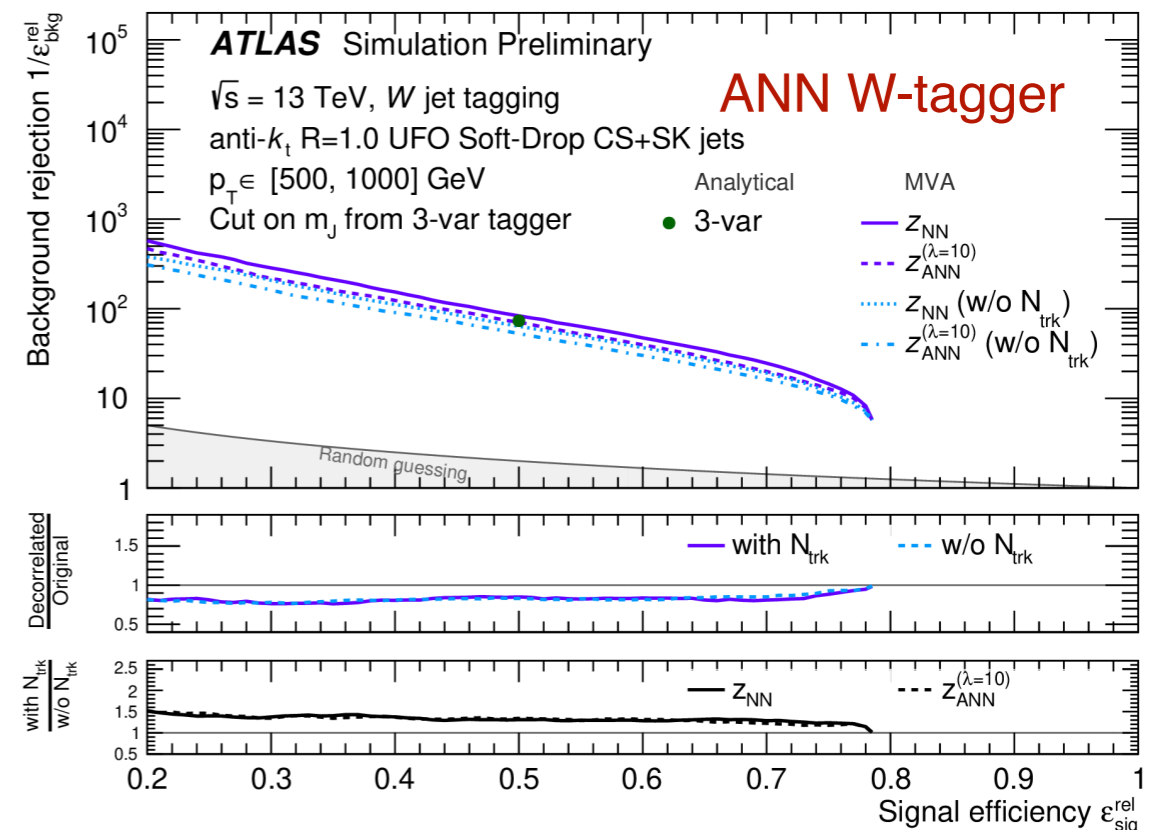
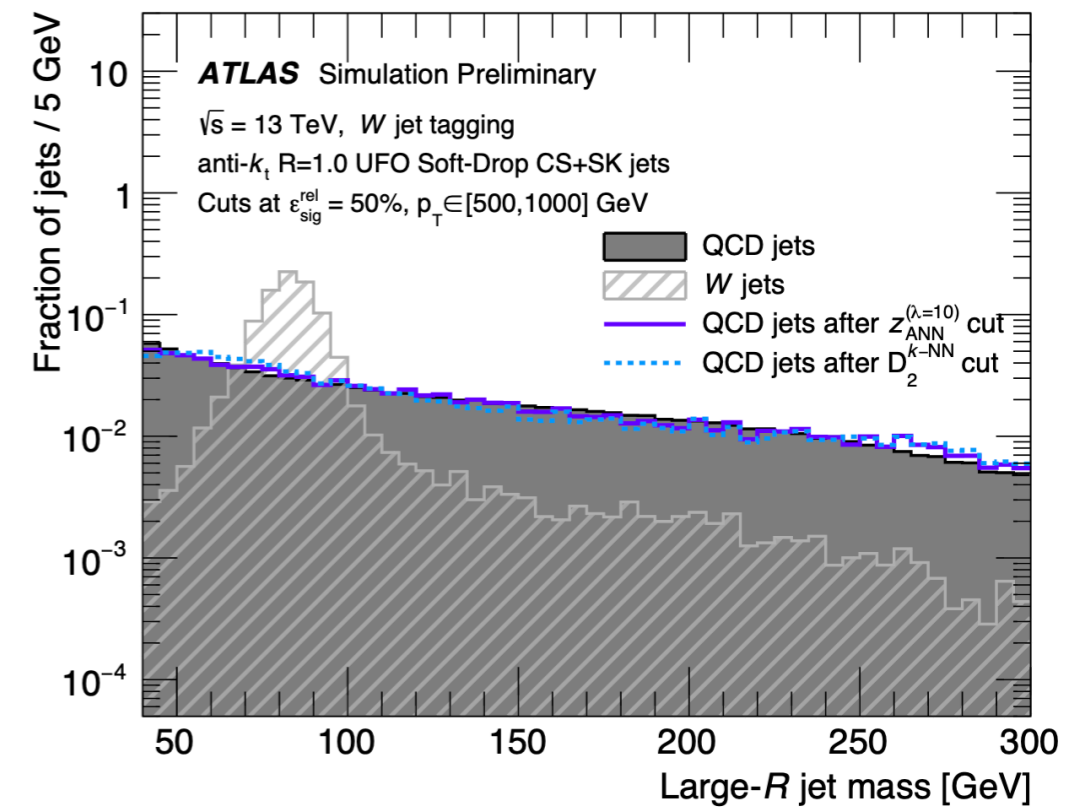
• SD CS+SK UFO Large- R jets.

- ▶ 3-variable based W and Z taggers: D_2 , N_{trk} and $m_{\text{jet}}^{\text{UFO}}$.
- ▶ ANN W/Z tagger:
 - ▶ decorrelating tagger score and $m_{\text{jet}}^{\text{UFO}}$.
 - ▶ Based on jet moments.
- ▶ Contained and inclusive DNN top taggers.
 - ▶ Optimisation of jet moments with respect to previous version.

Top tagger



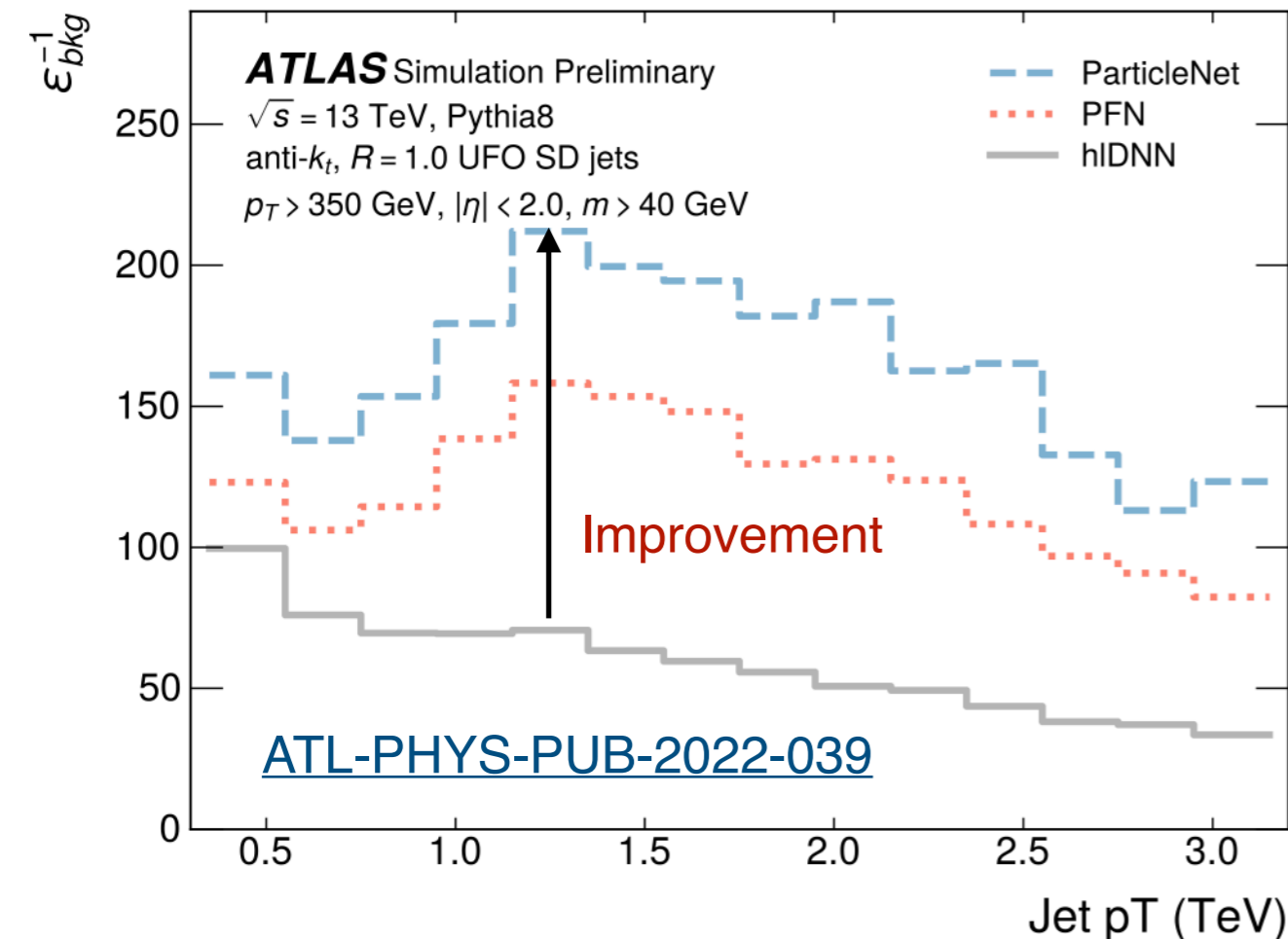
ANN W-tagger ATL-PHYS-PUB-2021-029



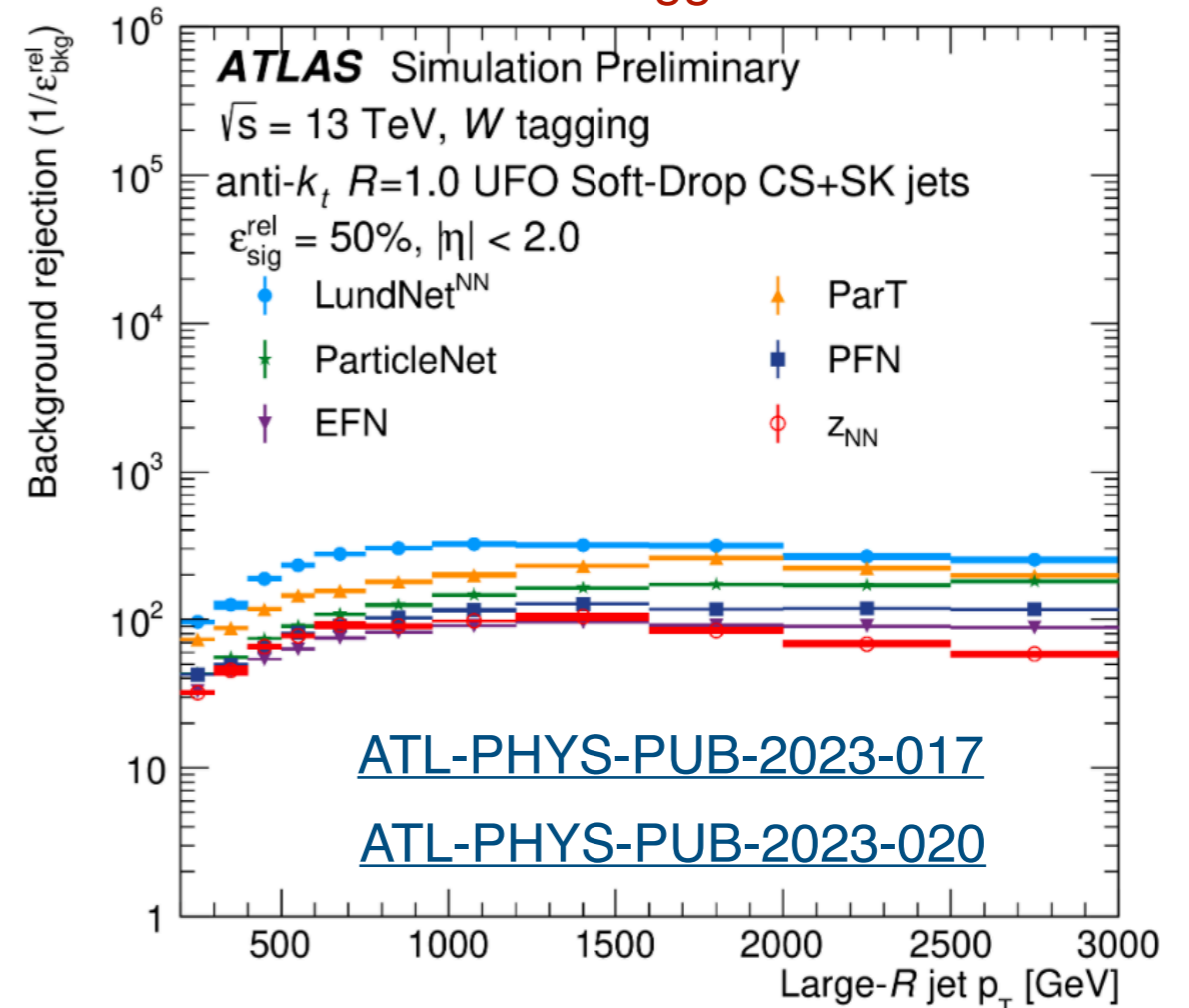
Jet tagging techniques

- Huge effort to derive taggers based on low level quantities.
 - ▶ All jet information is contained in the constituents.
 - ▶ Feed NNs with constituents to increase the performance?
 - Modern architectures help on this task: GNN, Transformers ...
 - Physics wise quantities replaced by powerful NNs?
 - Blackbox: difficult to understand which physics the NN is learning.

ParticleNet top tagger



W tagger

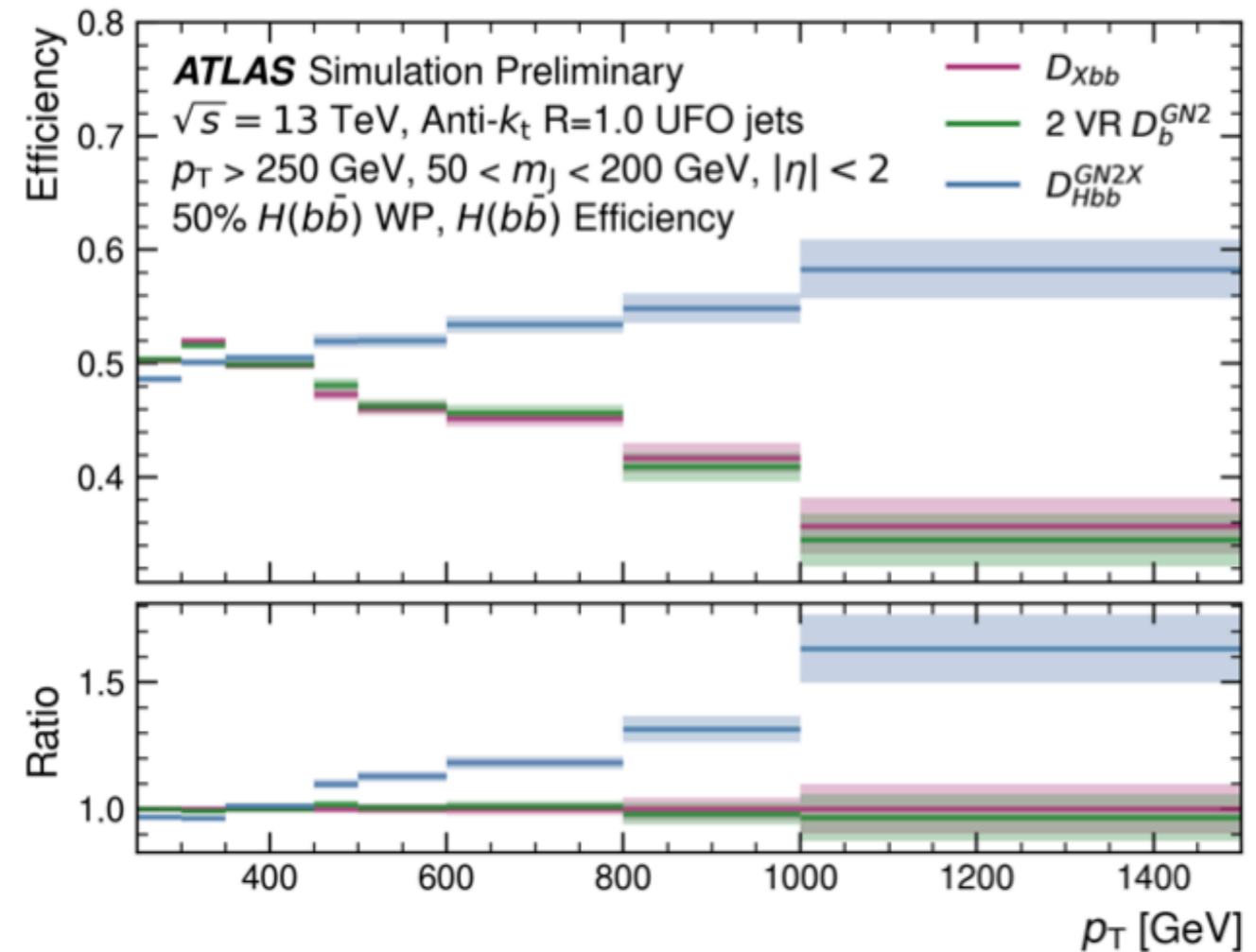
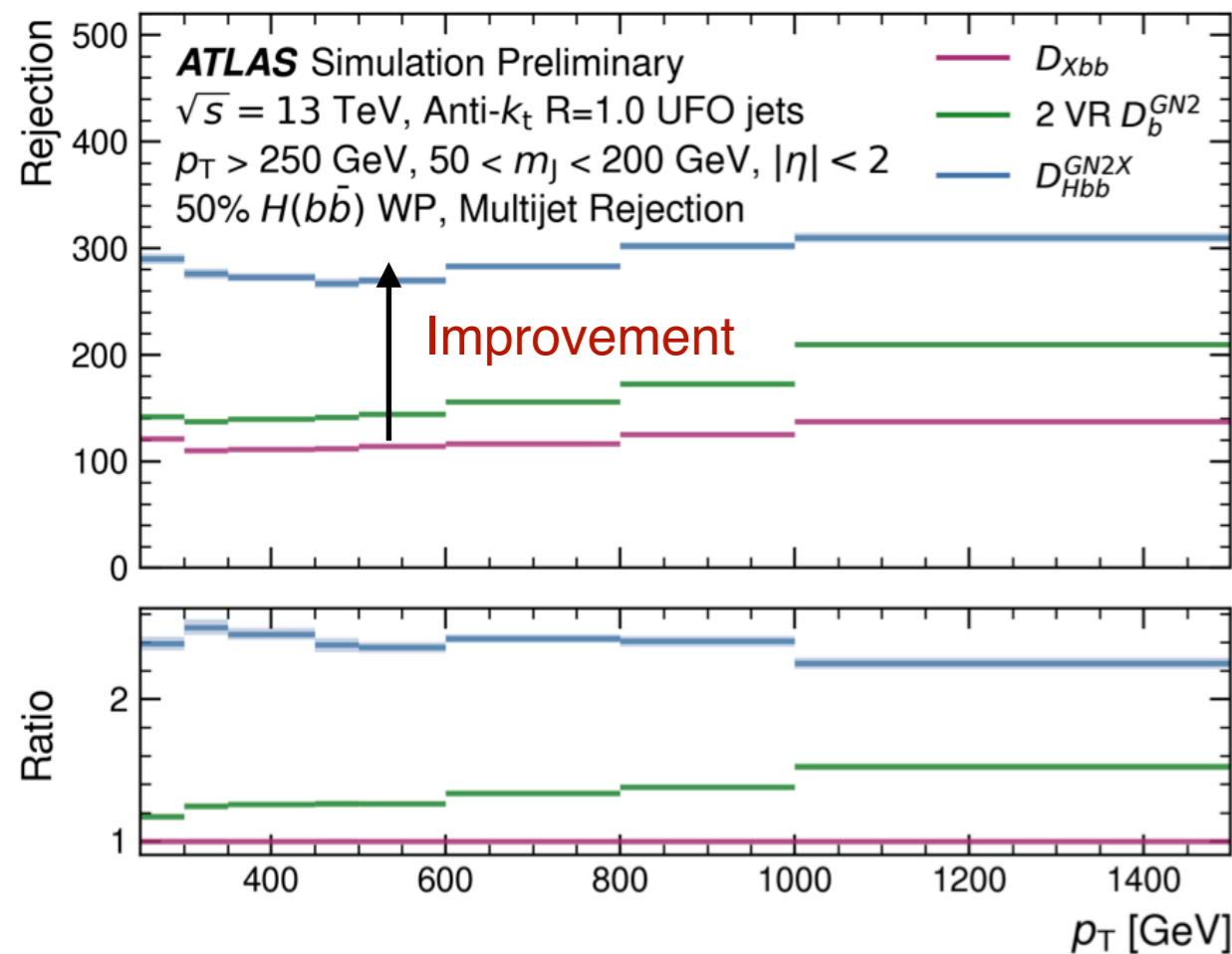


- GNNs including Lund-plane information best performance W tagger.
 - ▶ Physics motivated inputs → How the information is sorted still matters!

Jet tagging techniques

- Similarly, including all available tracking information beneficial for $H \rightarrow b\bar{b}$ tagger.
 - ▶ p_T , mass of the jet, impact parameters, number of hits in different ID layers of associated tracks
 - ▶ All these information is used to feed a Transformer architecture $\approx 1.5M$ of parameters.

[ATL-PHYS-PUB-2023-021](#)

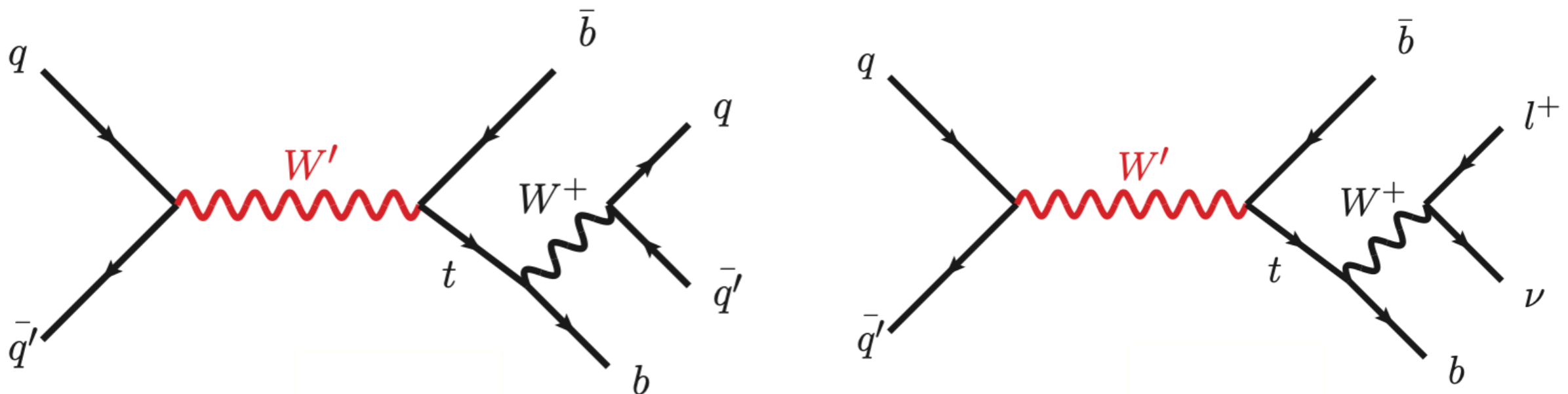


- A factor of 2.5 of improvement in multi jet rejection, having a 1.5 higher signal efficiency at $p_T > 1$ TeV!

Physics analyses

$W' \rightarrow tb$ search ([arXiv:2308.08521](https://arxiv.org/abs/2308.08521))

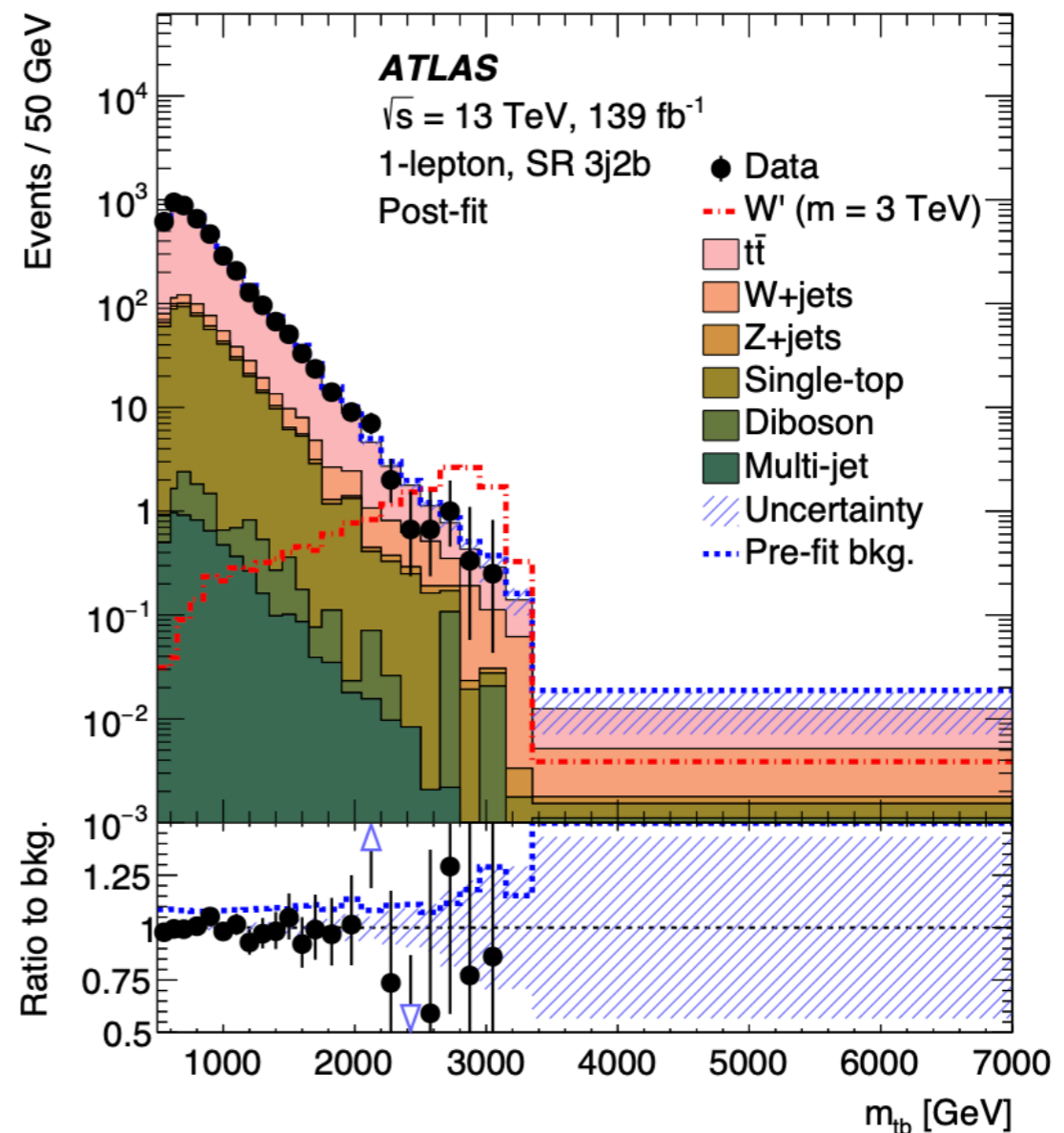
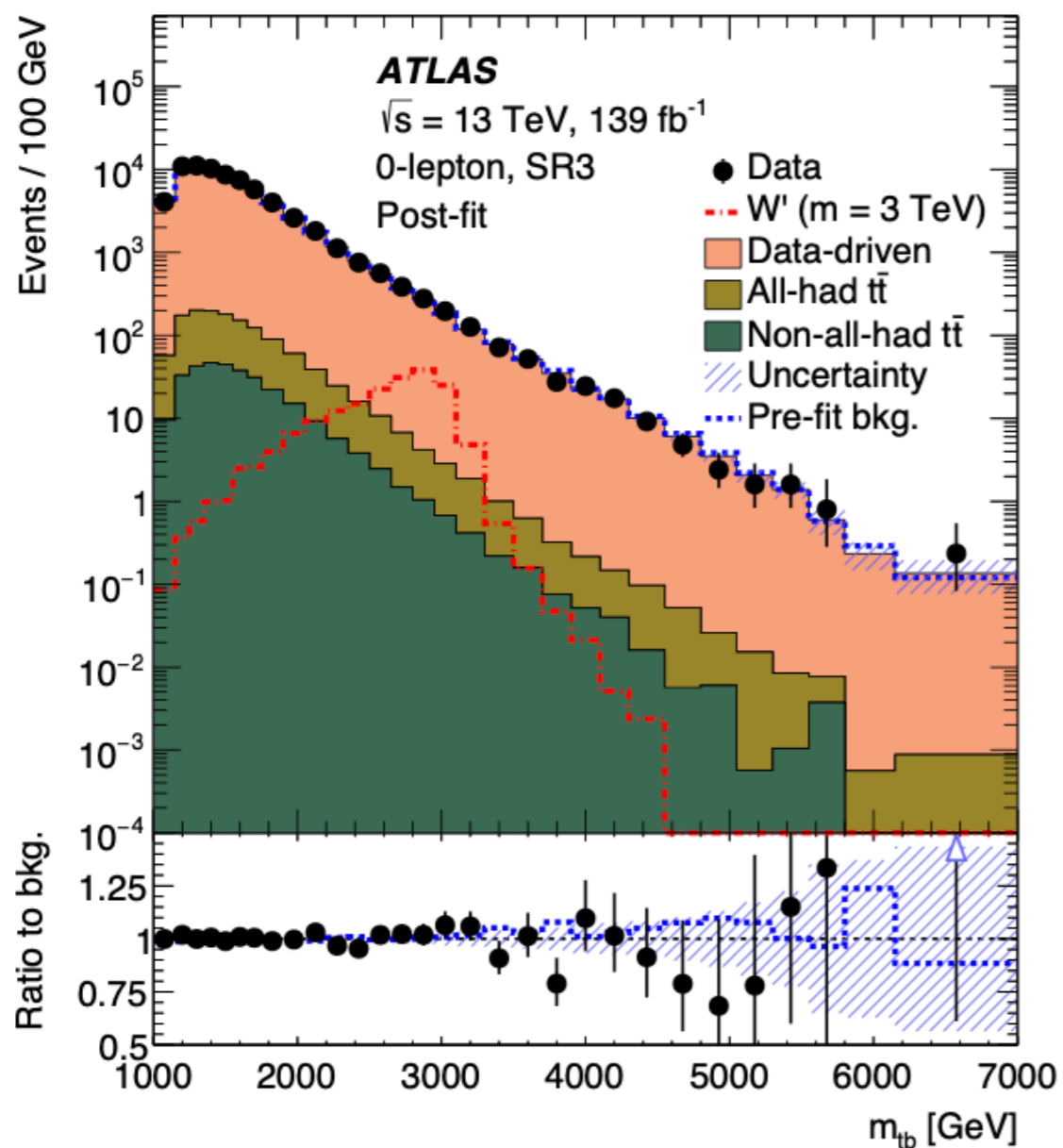
- Theories beyond the Standard Model (SM), involve enhanced symmetries that predict new gauge bosons, usually called W' or Z' bosons.
 - Different models/theories such as extra dimensions, strong dynamics predict new vector charged-current interactions, some with preferential couplings to quarks or third-generation particles.
 - Sequential Standard Model (SSM) to capture main phenomenology: $m_{W'}$, W_L/W_R , $\kappa = g'/g$
 - Focus on $W' \rightarrow tb$ decay channel.



- Hadronic and leptonic top decay channels considered in this search.
 - For large $m_{W'}$, boosted tops will be produced: jet tagging techniques to identify hadronically decaying top \rightarrow DNN top tagger!
- Different SRs are defined based on the number of final state b-jets (leptonic/hadronic channel) and DNN top tagger score (hadronic channel) to improve signal sensitivity.

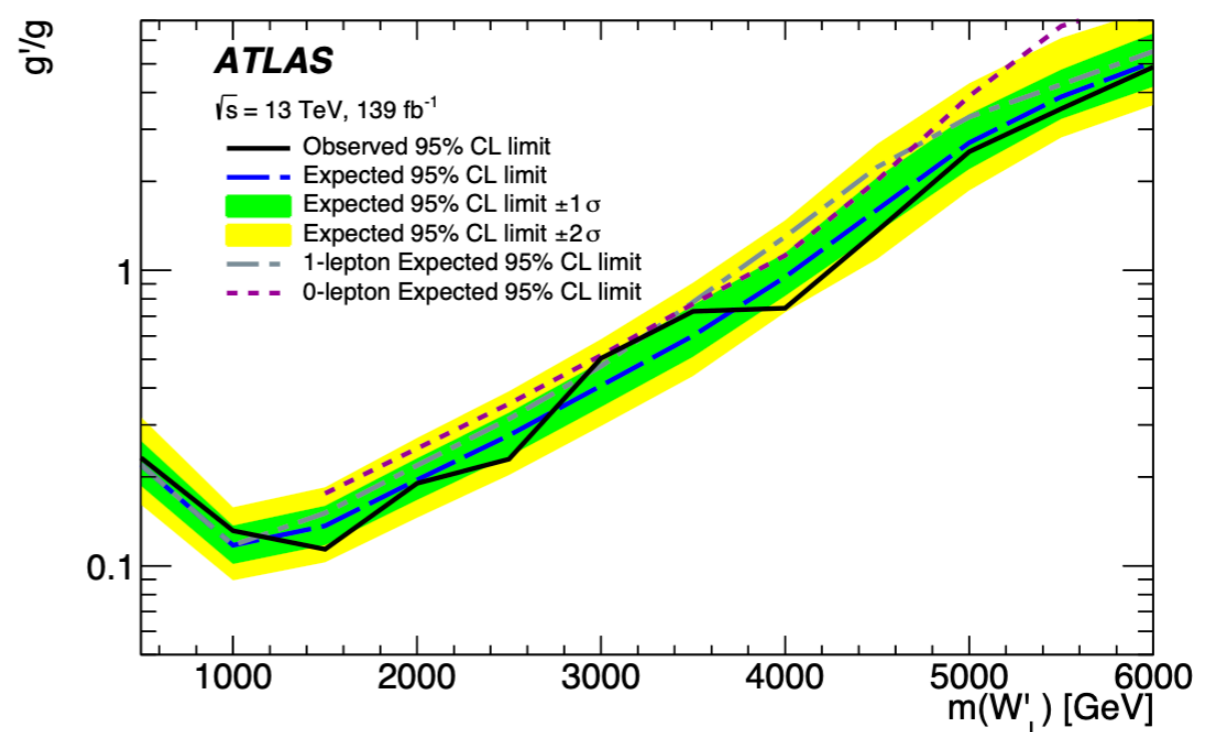
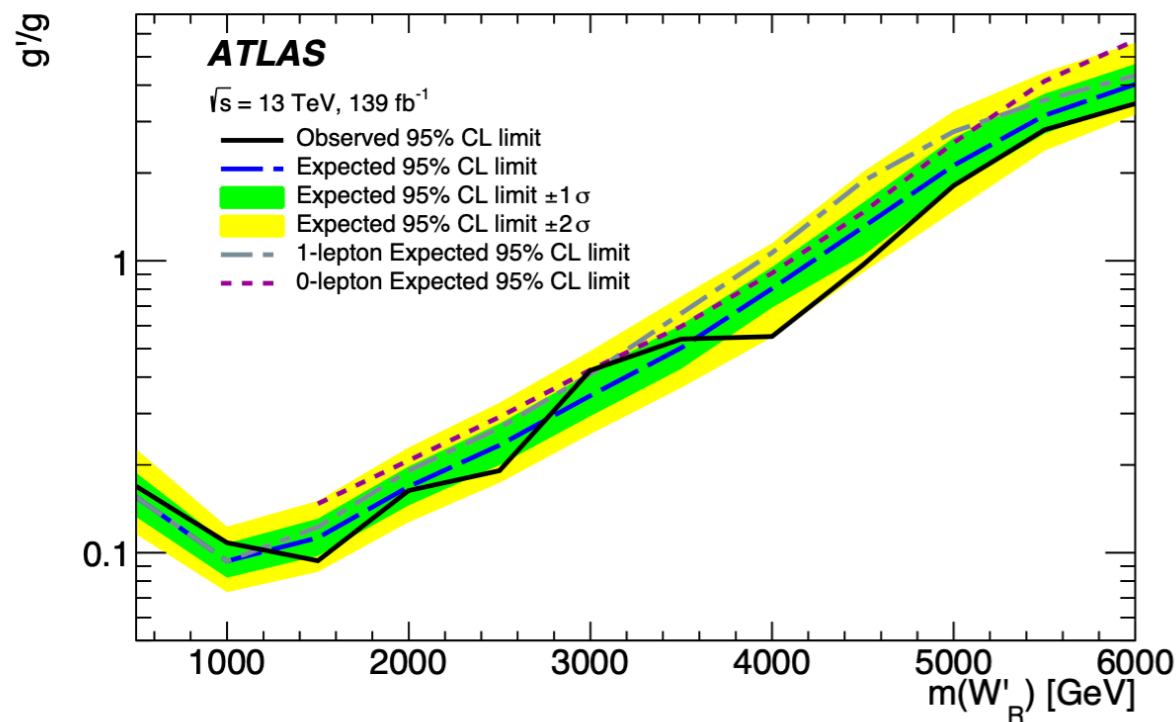
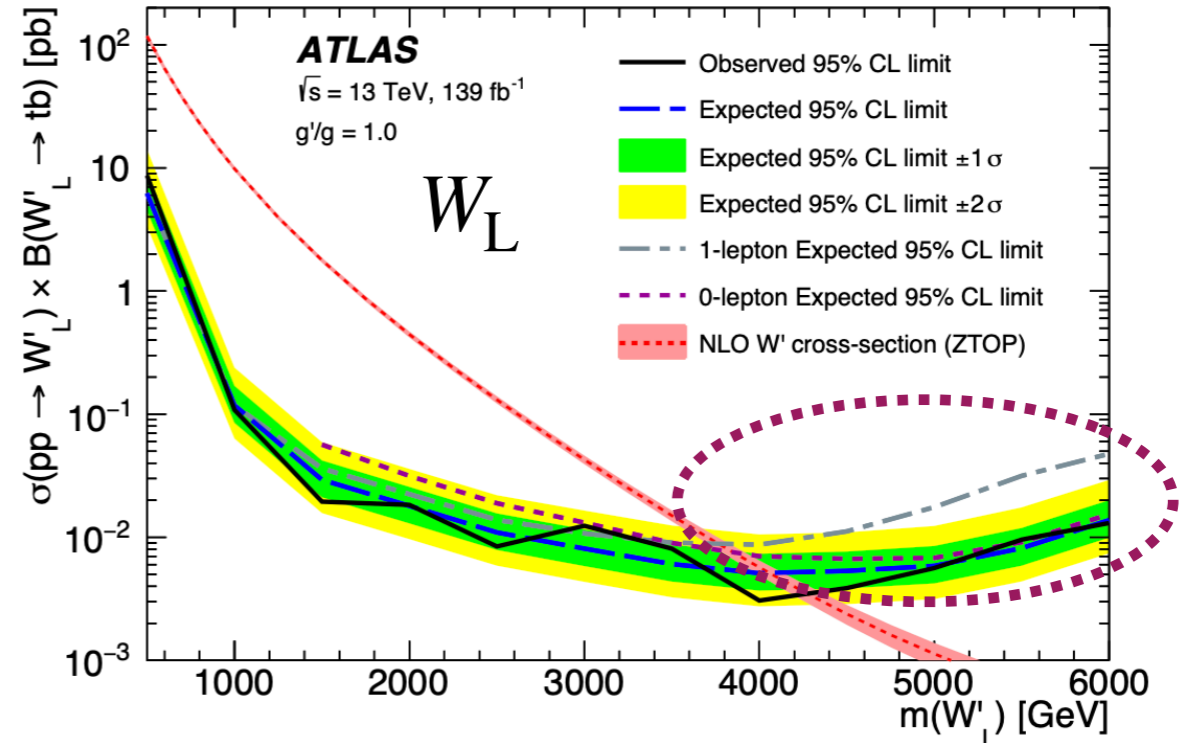
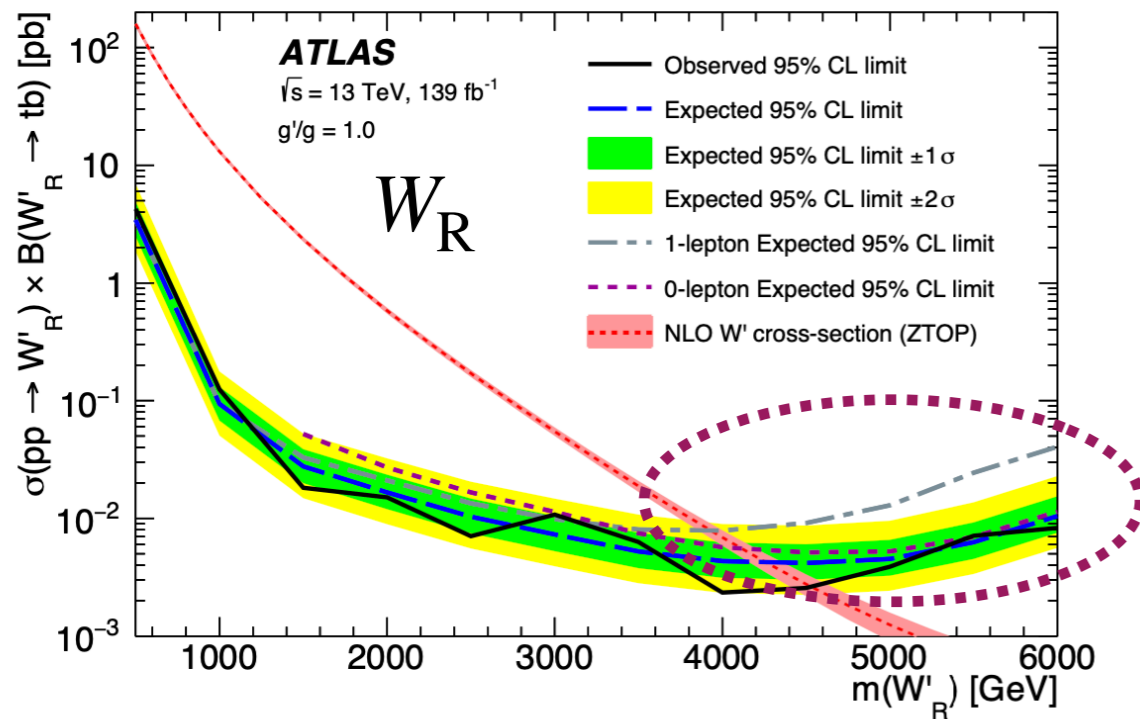
$W' \rightarrow tb$ search (arXiv:2308.08521)

- Main backgrounds:
 - Leptonic channel: $t\bar{t}$, W +jets, Z +jets, single top
 - ▶ Dedicated control regions (CRs) to get insight on background norm./shapes.
 - Hadronic channel: multijet
 - ▶ Data-driven estimation based on dedicated CRs.
- Profiled binned likelihood fit to SR and CRs to test signal hypothesis.



$W' \rightarrow tb$ search (arXiv:2308.08521)

- 2D limits as functions of κ and $m_{W'_L}/m_{W'_R}$ are derived:
 - For $\kappa = 1.0$, right-handed (left-handed) W' with masses below 4.6 (4.1) TeV are excluded.



High p_T^H measurement in VH final state (ATLAS-CONF-2023-067)

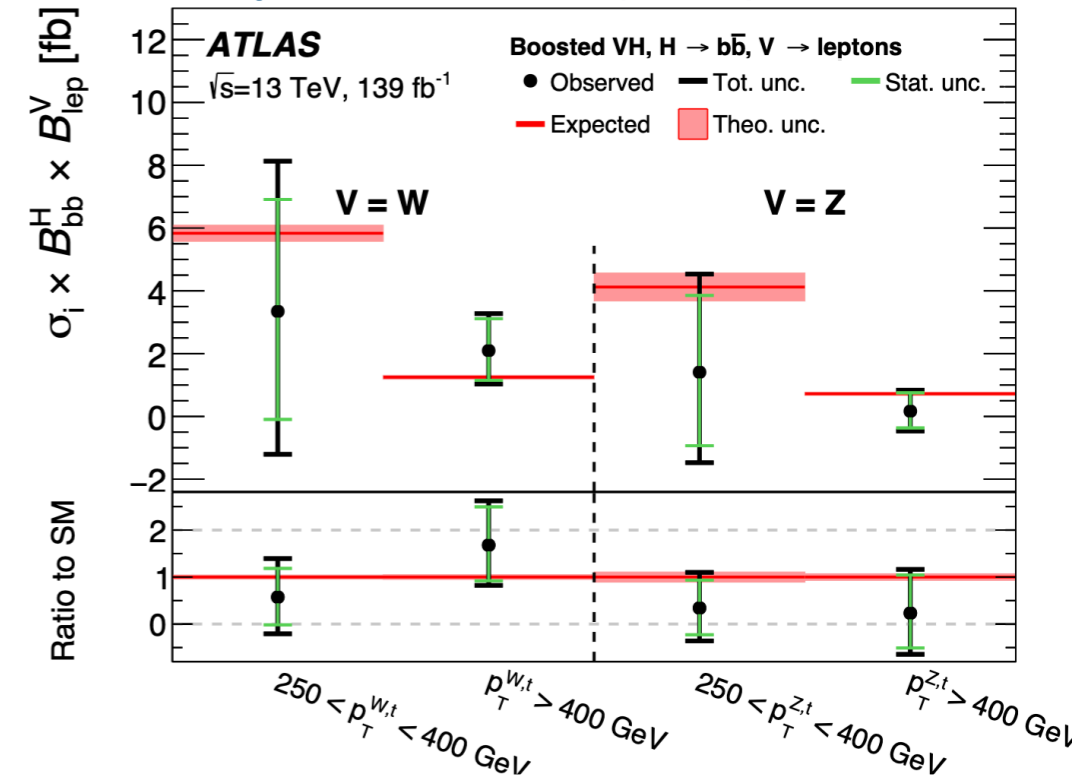
- The p_T^H distribution is measured in $pp \rightarrow V(q\bar{q})H(b\bar{b})$ processes.

- Larger W/Z branching fraction in hadronic channels allows to extend the VH measurement beyond $p_T > 400$ GeV.
- Main background coming from multi jet events estimated with a data-driven method.
- W/Z and Xbb taggers applied to improve signal sensitivity.

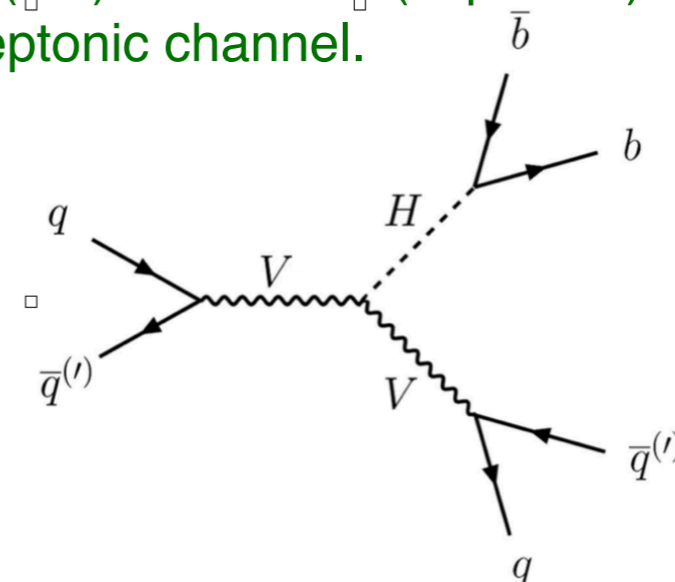
- Signal strength estimated from fits to the mass of the Higgs candidate.

WIZ leptonic channel

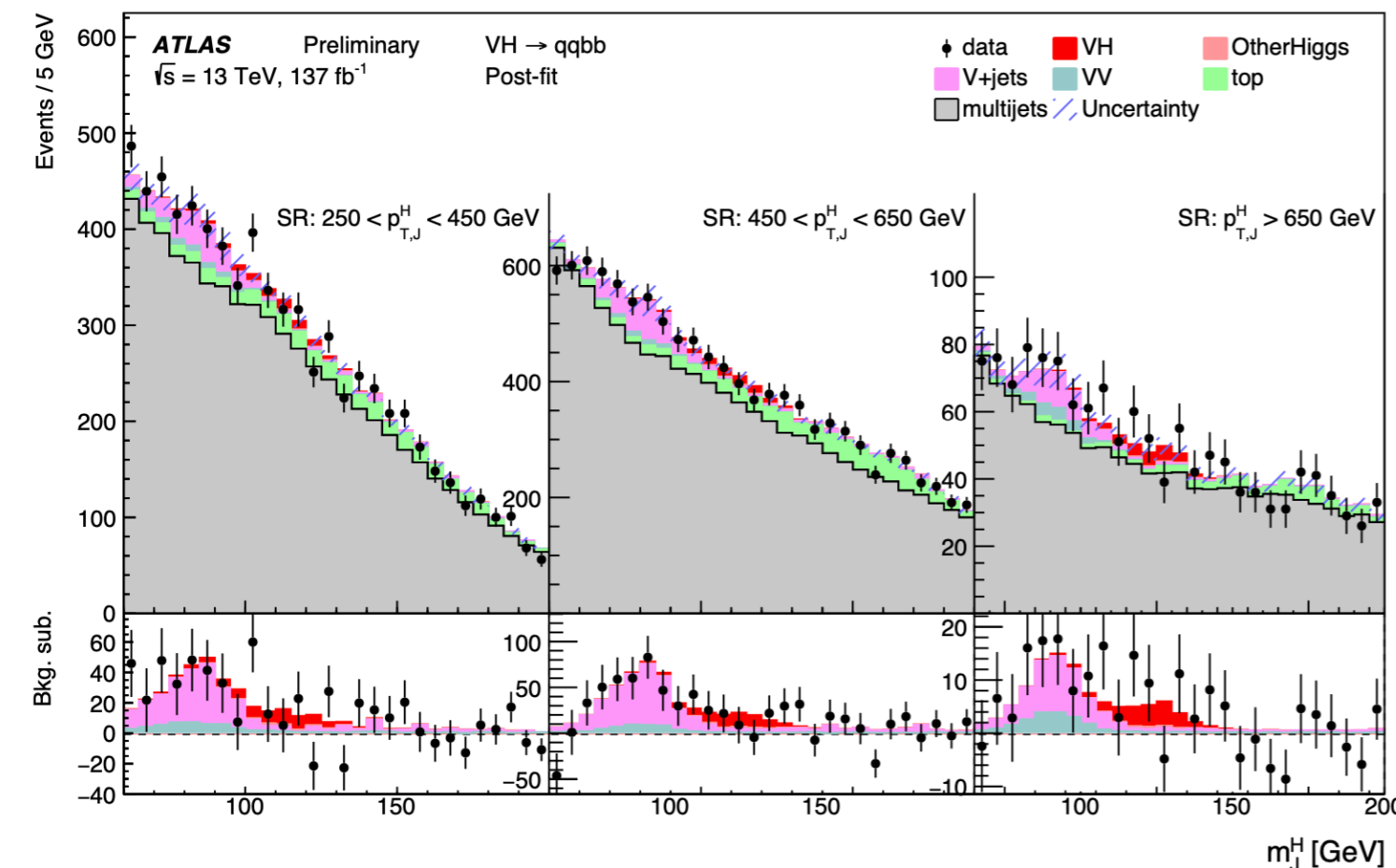
Phys. Lett. B 816 (2021) 136204



- 2.1 (2.7) σ observed (expected) for VH signal in leptonic channel.



- 1.7 (1.2) σ observed (expected) for VH signal in hadronic channel.



Conclusions

- Jet tagging is a powerful tool which helps to improve signal sensitivity.
 - ▶ It allows to include hadronic decay channels of boosted resonances.
 - Larger branching fractions than leptonic channels.
 - ▶ In the case of boosted tops, typically better sensitivity at very large p_T compared to leptonic decay channel.
 - ▶ Fully reconstruct the event with W s in the final state.
 - Not possible in $W \rightarrow l\nu_l$ decay channels.
 - ▶ Taggers need to be calibrated for an optimal use in physics analyses.
- On the experimental side, Large- R jet definition crucial to improve tagging performance.
 - Better grooming techniques, jet constituent definition ...
- Machine learning usual technique these days to improve tagging performance.
 - ▶ Better architectures able to extract crucial information from quite low level jet inputs.
 - ▶ Challenge for the calibration: modelling dependence, pile-up stability ...