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*b*-jet tagging: efficiency  
measurement and impact  
on  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis

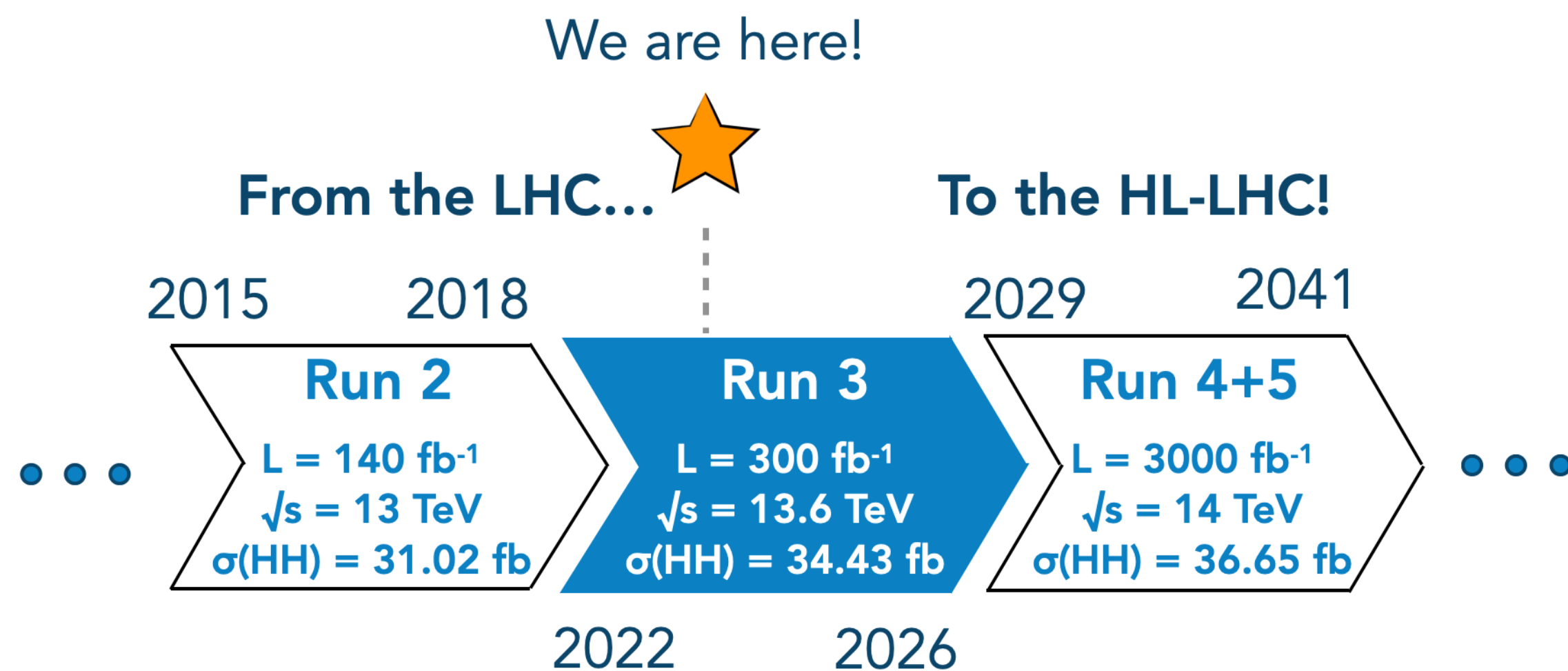
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UNIVERSITÀ  
DEGLI STUDI  
DI MILANO



# Outline: outlook for HH searches in Run 3



- Impressive sensitivity to HH production from the new batch of Legacy Run 2 HH analyses, that was finalized in the past months!

↳  $b\bar{b}\gamma\gamma$ ,  $b\bar{b}\tau\tau$ , and  $b\bar{b}b\bar{b}$  (main channels), +  $b\bar{b}ll$  and multi leptons + **combination!**

## HH combination

	36.1 fb <sup>-1</sup>	140 fb <sup>-1</sup>
Expected limit on $\mu_{HH}$	10	2.4

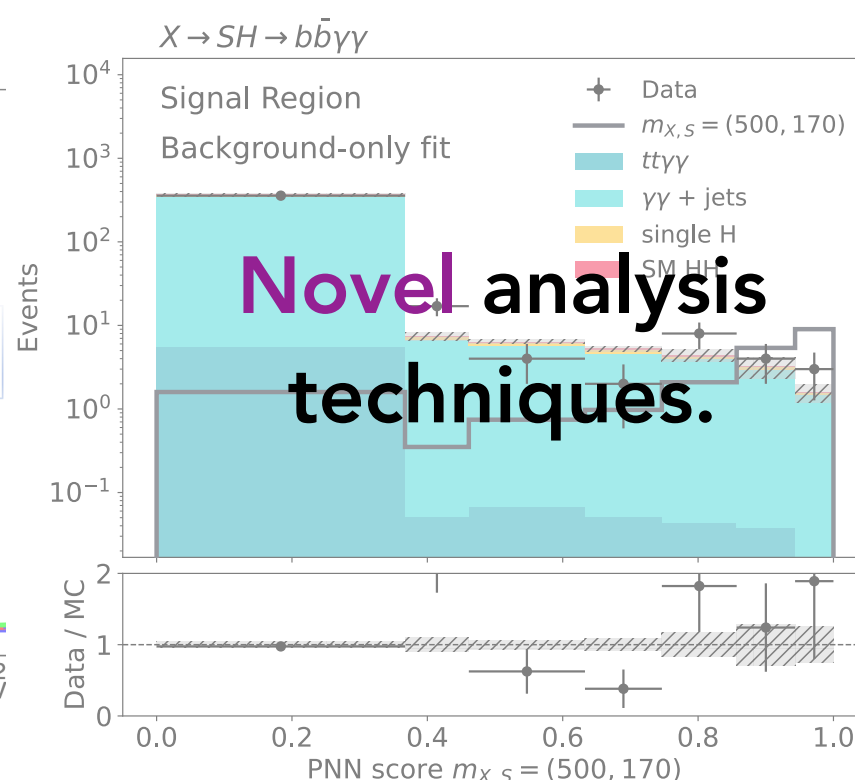
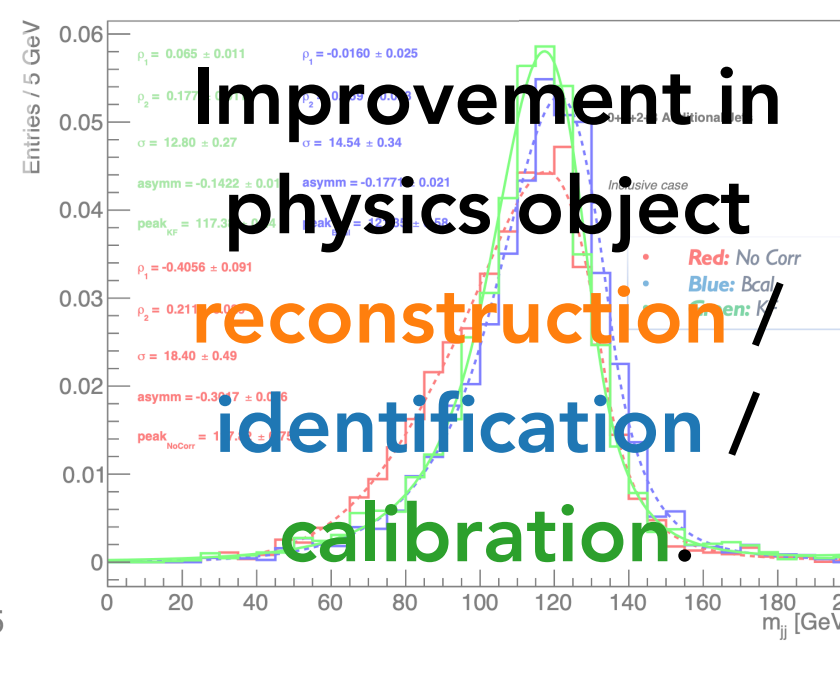
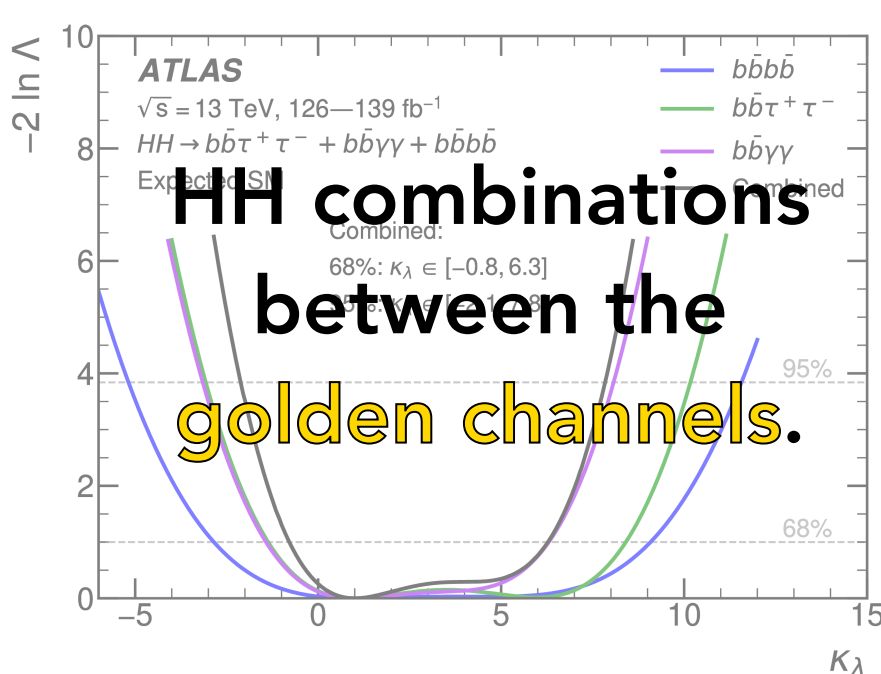
Improvement of a **factor > 4** w.r.t. partial Run 2!

Expected SM HH:

4000

~10<sup>4</sup>

~10<sup>5</sup>!



Tackling systematic uncertainties.

Source	Type	Upper limit	Observed Difference w.r.t. full syst.	Upper limit	Expected Difference w.r.t. full syst.
Factorization and renormalization scale	Normalization	3.81	-3.7%	4.80	-4.76%
PDF set and $\alpha_s$ value	Normalization	3.97	-0.08%	5.02	-0.08%
Parton showering model	Normalization	3.97	-0.13%	5.02	-0.13%
Heavy flavor content	Normalization	3.97	-0.14%	5.02	-0.14%
Flavor tagging	Normalization	3.97	+0.21%	5.02	-0.20%
$B_R(H \rightarrow \gamma\gamma, b\bar{b})$	Normalization	3.97	+0.21%	5.02	-0.20%
Spurious signal	Normalization	3.98	+0.45%	5.03	-0.02%

- Run 3 represents a unique opportunity for HH searches!

➔ High momentum and increasing interest from many teams / institutions!

- There are many aspects that we can (and should) explore to try and reach a first hint of observation of HH production already with the full Run 3 dataset.

➔ One of them is the improvement in object identification, especially  $b$ -tagging!

# Outline: state of the art $b$ -tagging in ATLAS and HH searches

- A new generation of GNN-based  $b$ -tagging algorithms was developed in ATLAS.

➡ **Boost** in performances of the new  $b$ -tagging algorithms w.r.t. the **older RNN-based generation** used in Run 2 (e.g. DL1r).

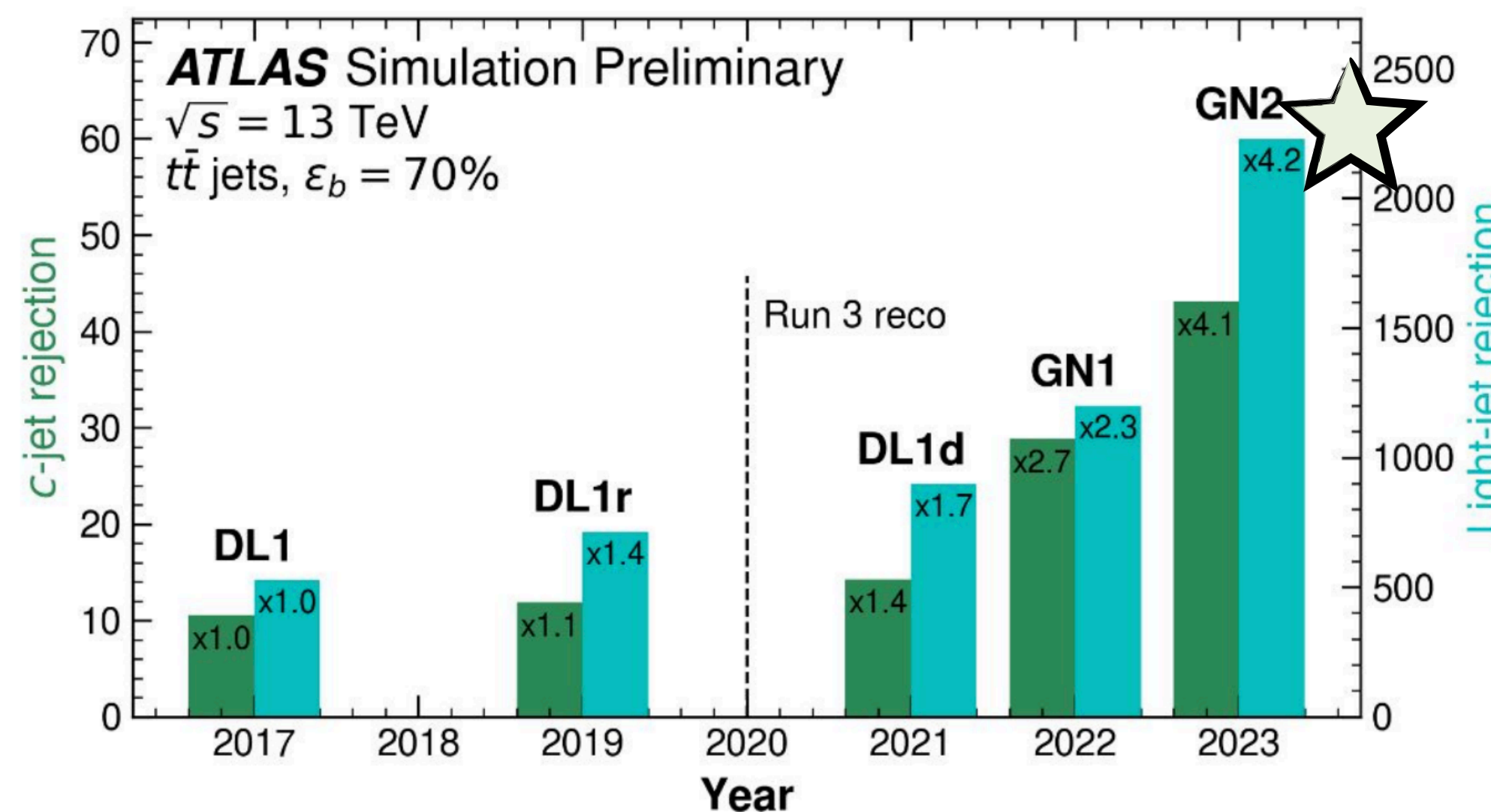
- Expected to **sharpen the sensitivity** of the new di-Higgs analyses!

➡ - The **most sensitive channels** (=  $b\bar{b}\gamma\gamma$ ,  $b\bar{b}\tau\tau$ , and  $b\bar{b}b\bar{b}$ ) have **at least 2  $b$ -jets** in the **final state**.

- Impressive performances in the  $HH \rightarrow 4b$  analysis by **CMS** after the **adoption** of their new **ParticleNet-based tagger already in Run 2!**

	35.9 fb <sup>-1</sup>	138 fb <sup>-1</sup>
Expected limit on $\mu_{HH}$	114	5.1

➡ **× 11 (!) improvement**, after factoring out the factor 2 improvement from the increase in luminosity.



**GN2** is the current **state of the art** in ATLAS!

- Before being able to employ the new GN2 tagger in ATLAS physics analysis, we need to calibrate these algorithms (i.e. measuring their efficiency on **real  $pp$  collision data**)!

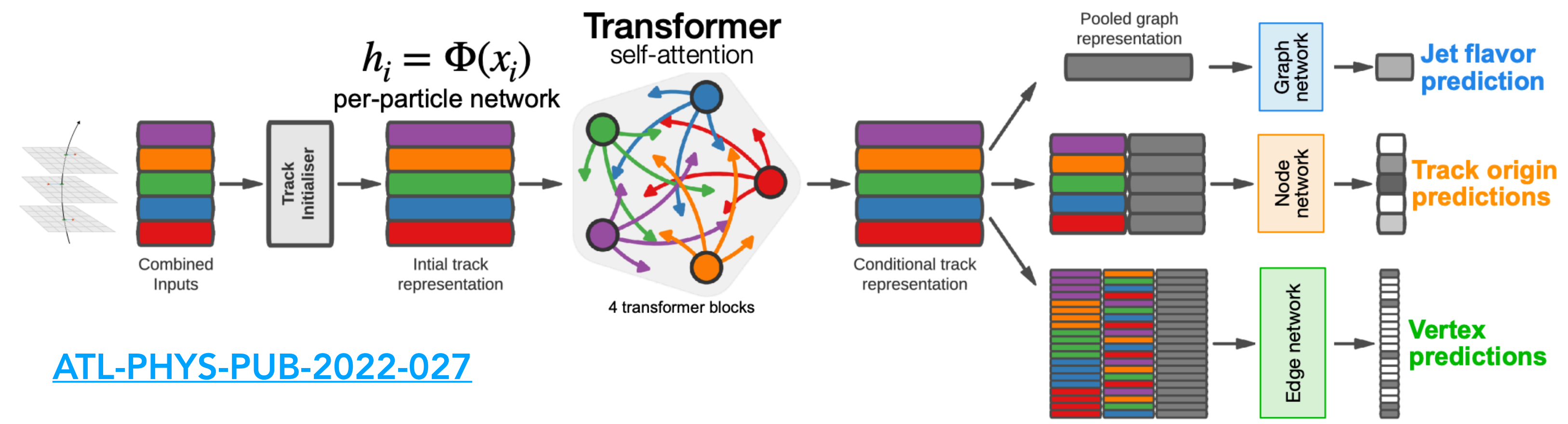
➡ - In this presentation I will try to give an overview of the new GN2 tagger and present the measurement of the  $b$ -tagging efficiency on data.

- Finally, I will try to estimate the impact of adopting the new GN2 tagger, using, as baseline, the Run 2 Legacy  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis.

**GN2: state of the art  $b$ -tagging in ATLAS**

# $b$ -tagging with GN2: inputs & training

- **GN2** sees a **jet** as collection of **tracks** + some **global information** (= such as its  $\eta$  and  $\phi$  coordinates)!
- Trained to **understand** the **internal structure** of the jets, thanks to relying on **low-level information** about **tracks**, **impact parameters**, and **hits**.



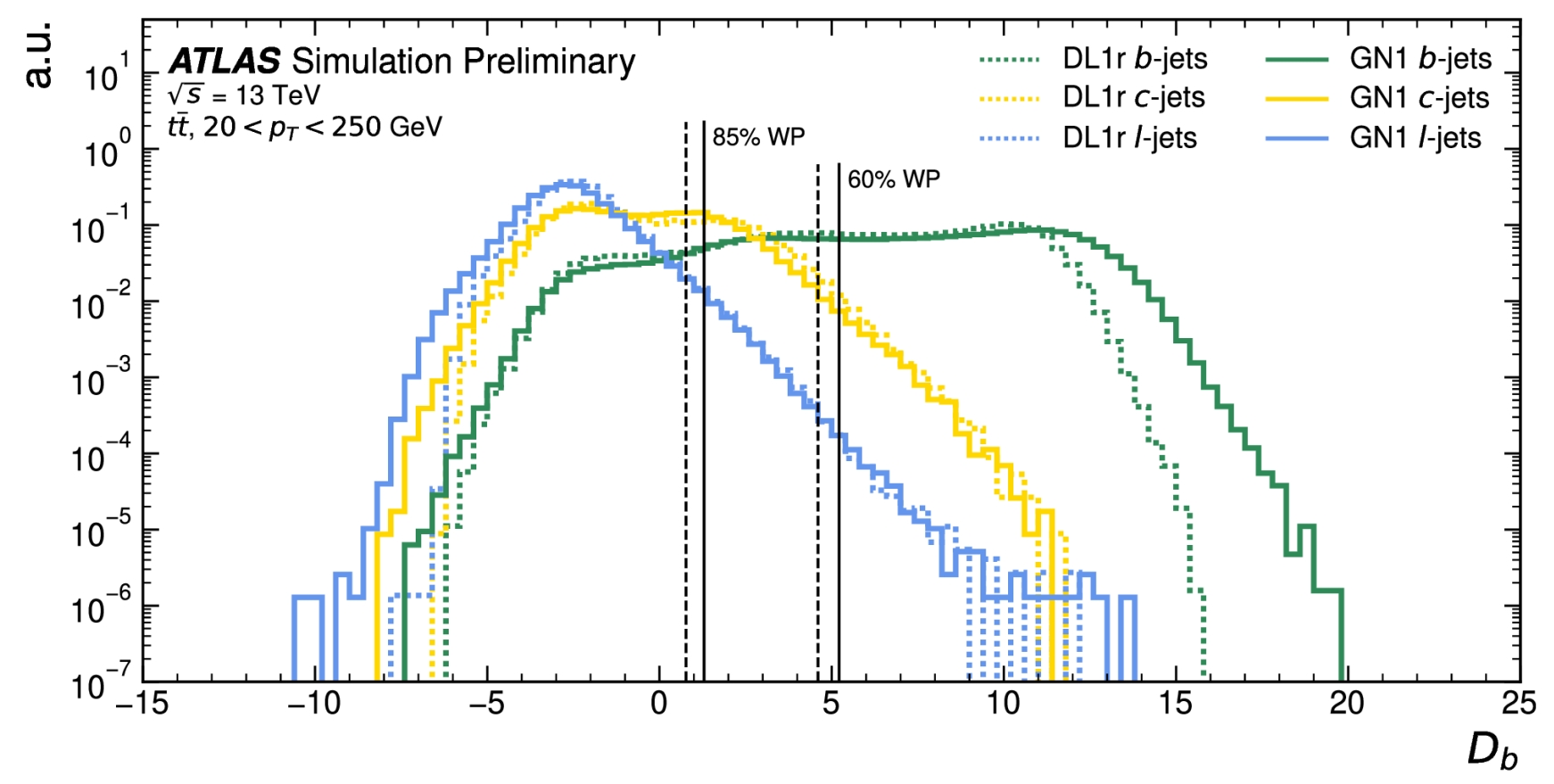
[ATL-PHYS-PUB-2022-027](#)

➔ Provides a prediction of the jet's flavour.

➔ 4 output classes:  $\begin{pmatrix} p_b \\ p_c \\ p_u \\ p_\tau \end{pmatrix}$  ★ **New!**

**Combined** together to define a **discriminant for  $b$ -tagging!**

➔ 
$$D_b = \log \frac{p_b}{f_c p_c + f_\tau p_\tau + (1 - f_c - f_\tau) p_u}$$

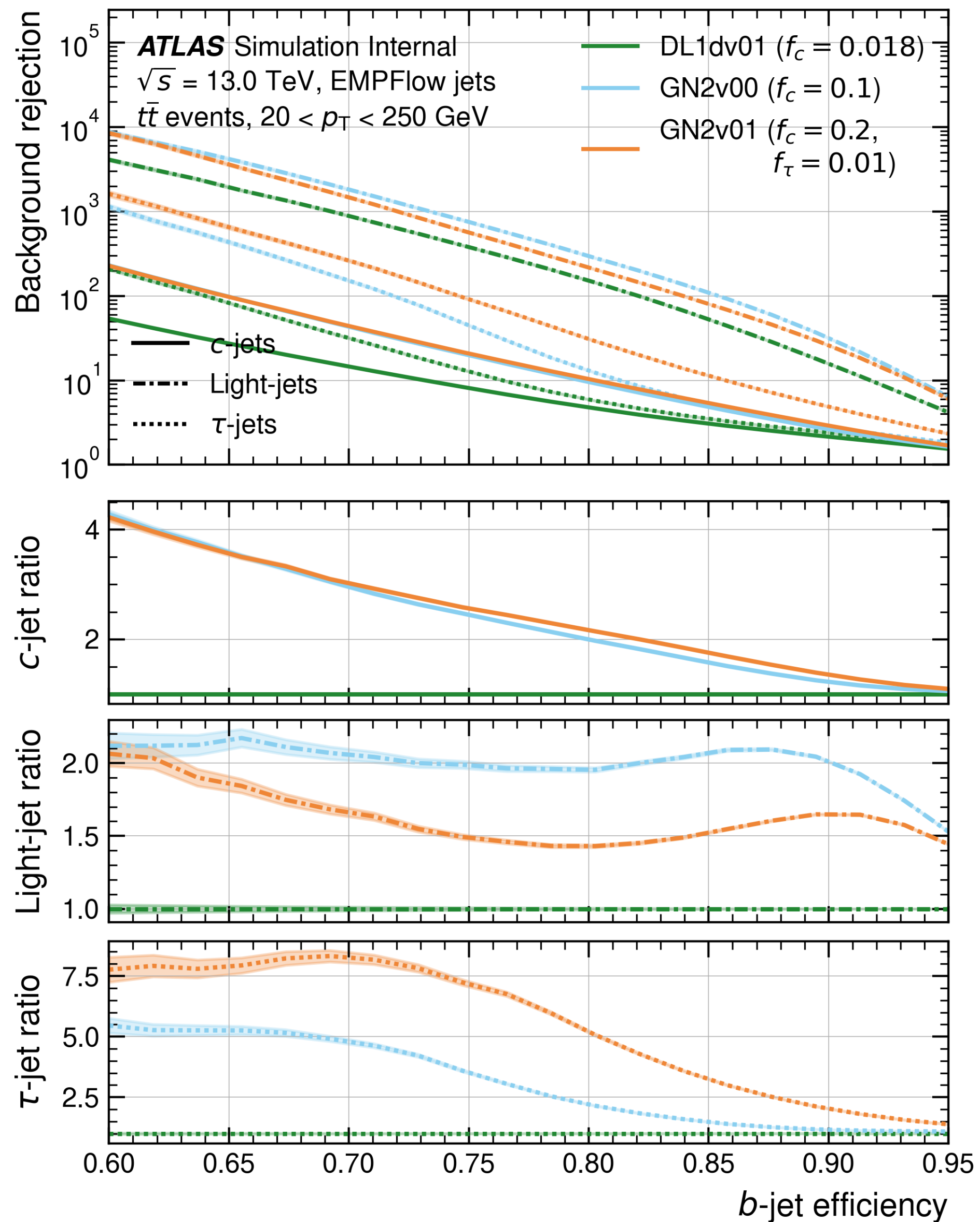


By cutting on the  $D_b$  discriminant we can define WPs characterized by a given  $b$ -tagging efficiency.

- ➔ - Measured inclusively on a  $t\bar{t}$  MC sample.
- ➔ - **5 WPs!** ➔ 90%, 85%, 77%, 70%, 65%.

- +
- Two **auxiliary tasks**:
1. **Prediction of the physics process** initiating the tracks.
  2. **Grouping the tracks into vertices!**
- ➔ Help convergence of the main task.

# $b$ -tagging with GN2: performances on MC



The performances of GN2 can be compared with those from the previous taggers by checking the **light** and **c-jet rejection**, given a determined  $b$ -jet efficiency.

➡ The new version of GN2 (= GN2v01) is also optimized to reject hadronic  $\tau$ s!

## Expected performances at a low $p_T$ regime

➡ Jets from  $t\bar{t}$  sample,  $20 \text{ GeV} < p_T < 250 \text{ GeV}$

- **Light jet rejection:** up to a factor  $\times 2$  improvement w.r.t. DL1d!
- **c-jet rejection:** up to a factor  $\times 4$  improvement w.r.t. DL1d!
- **$\tau$ -jet rejection:** up to a factor  $\times 7.5$  improvement w.r.t. DL1d!

# Calibration of the GN2-based $b$ -tagging efficiency

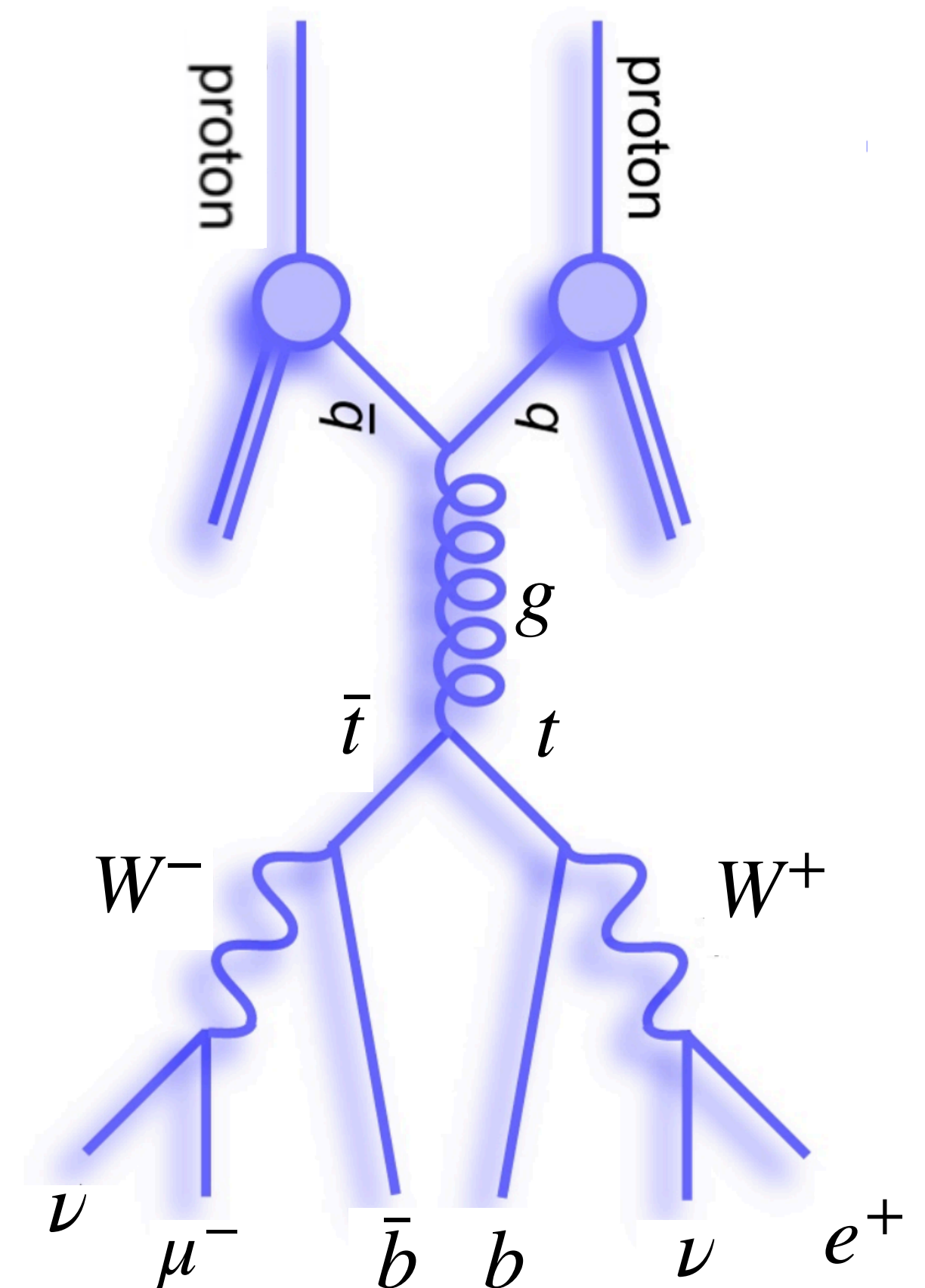
# $b$ -tagging efficiency measurement: outline

- Our **taggers (including GN2)** are optimized on **MC events**.
- However, MC samples may be affected by **imperfections** in the **modelling** of the physics process, or in describing the **detector response**.

➡ The **efficiency** has to be measured on **real  $pp$  collision data!**

➡ **How?**

1. Select a **pure sample of  $t\bar{t}$  events** in data, where **both top quarks decay leptonically**.
  - ➡ - Provides a sample **enriched** in true  **$b$ -jets**, thanks to the  $t \rightarrow Wb$  decay.
  - Can be **easily triggered upon** when considering **leptonic  $W$  decay**.
2. Identify **backgrounds** and estimate **systematic uncertainties**.
  - ➡ - Our leptonic  $t\bar{t}$  signal includes **two  $b$ -jets** and **two leptons** in the final state.
  - The **fake  $b$ -jet backgrounds** are estimated with the help of **CRs**, enriched in  $ll, bl, lb$  events.
3. Extract  **$b$ -tagging efficiency** (= selected  $b$ -jets / all  $b$ -jets) via a **maximum-likelihood fit** of the discriminant distribution **to data**, as a function of the leading and subleading jet  $p_T$  and  **$b$ -tag efficiency WP**.
  - ➡ **Scale factors** =  $b$ -tagging efficiency in data /  $b$ -tagging efficiency in MC.





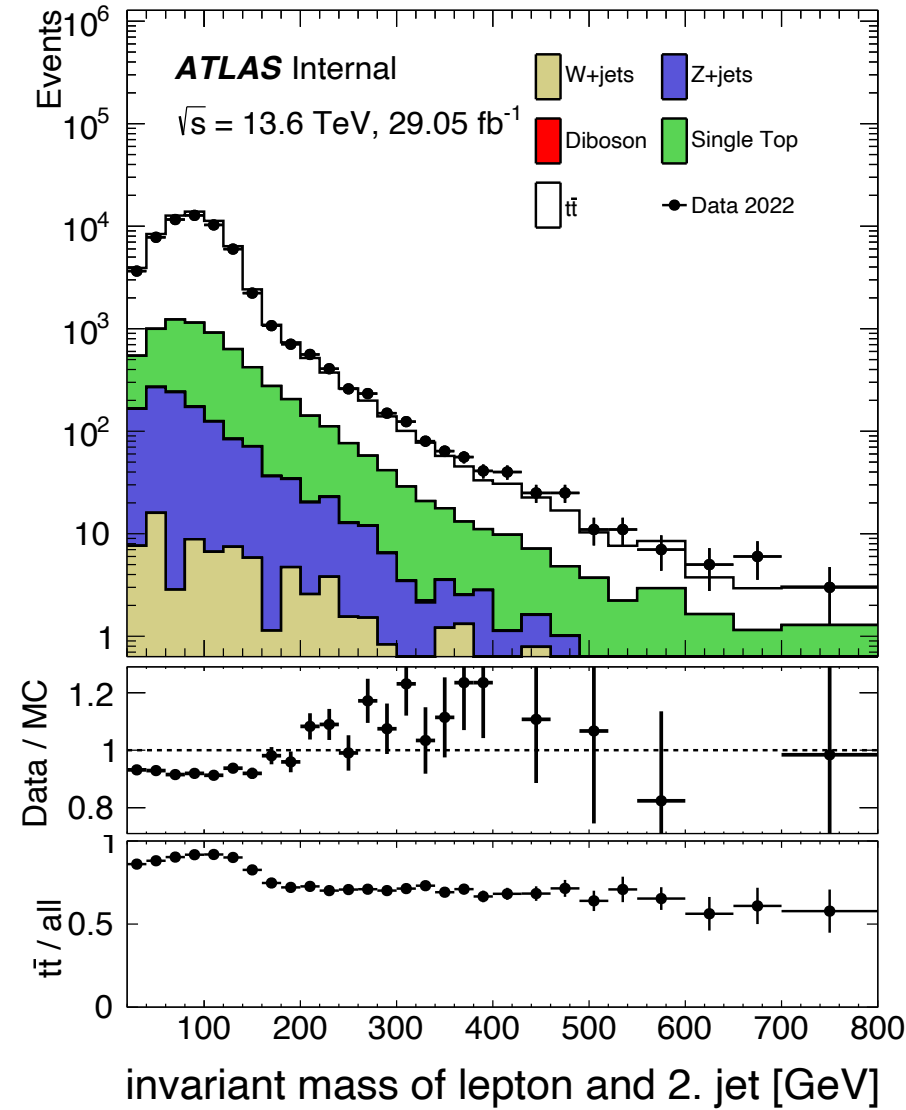
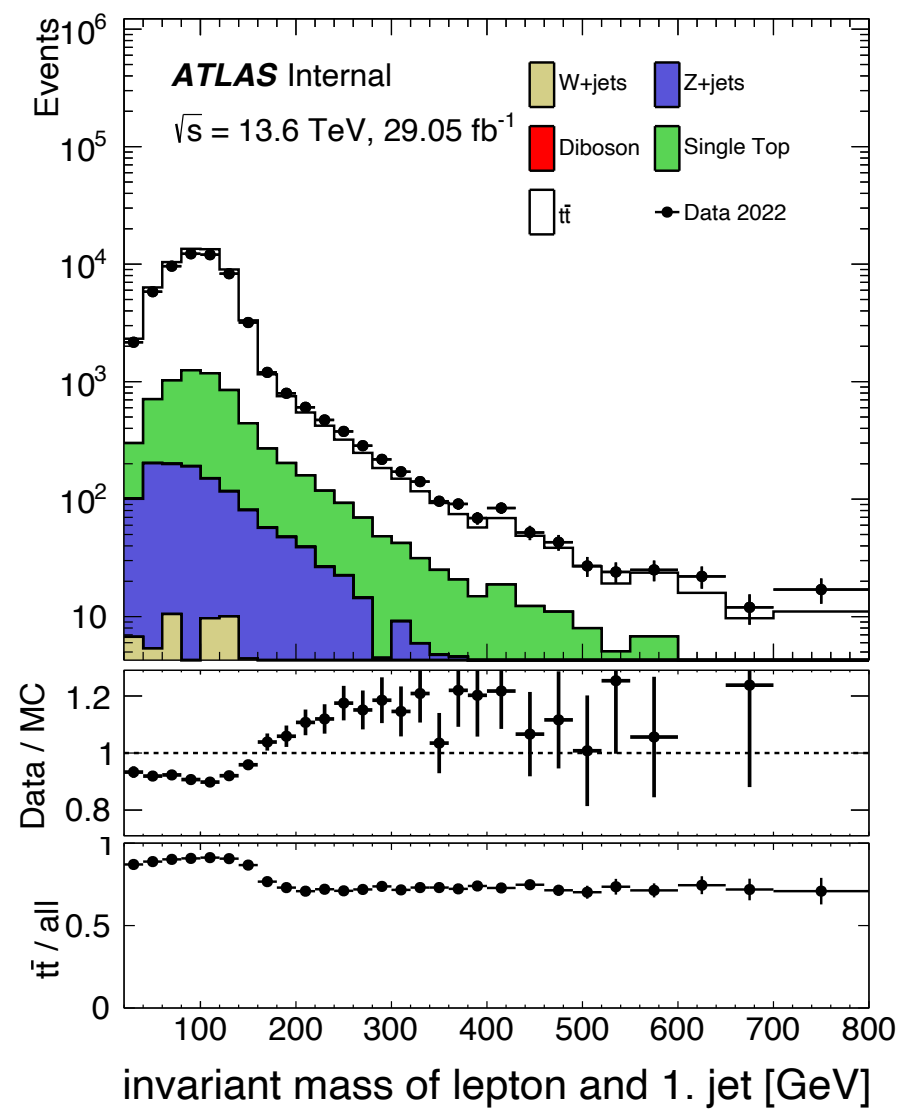
# $b$ -tagging efficiency measurement: event selection

## Event selection

- OR of **single-lepton** (electrons or muons) **triggers**.
- Exactly **two** reconstructed **opposite charge** and **different flavor leptons** ( $e$  and  $\mu$ ) with  $p_T > 27$  GeV.  $\rightarrow$  Reduces background from  $Z \rightarrow \ell\ell$  events.
- $m_{e\mu} > 50$  GeV.  $\rightarrow$  Reduces background from  $Z \rightarrow \tau^{lep}\tau^{lep}$  events.
- Exactly **two jets** with  $p_T > 20$  GeV.  $\rightarrow$  Reduces background from **single  $W$  production** and light jet production from ISR and FSR.

Main backgrounds: **W+jets**, **Z+jets**, **Single top production**, and **di-boson production**.

$\rightarrow$  Selected events are then split in **bins** of the **leading** and **subleading jet**  $p_T$ , and then further in a **SR** (enriched in events with two **true  $b$ -jets**) and in three **CRs** (enriched on  $bl$ ,  $lb$ ,  $ll$  events), based on the variables  $m_{j_1\ell}$  and  $m_{j_2\ell}$ .



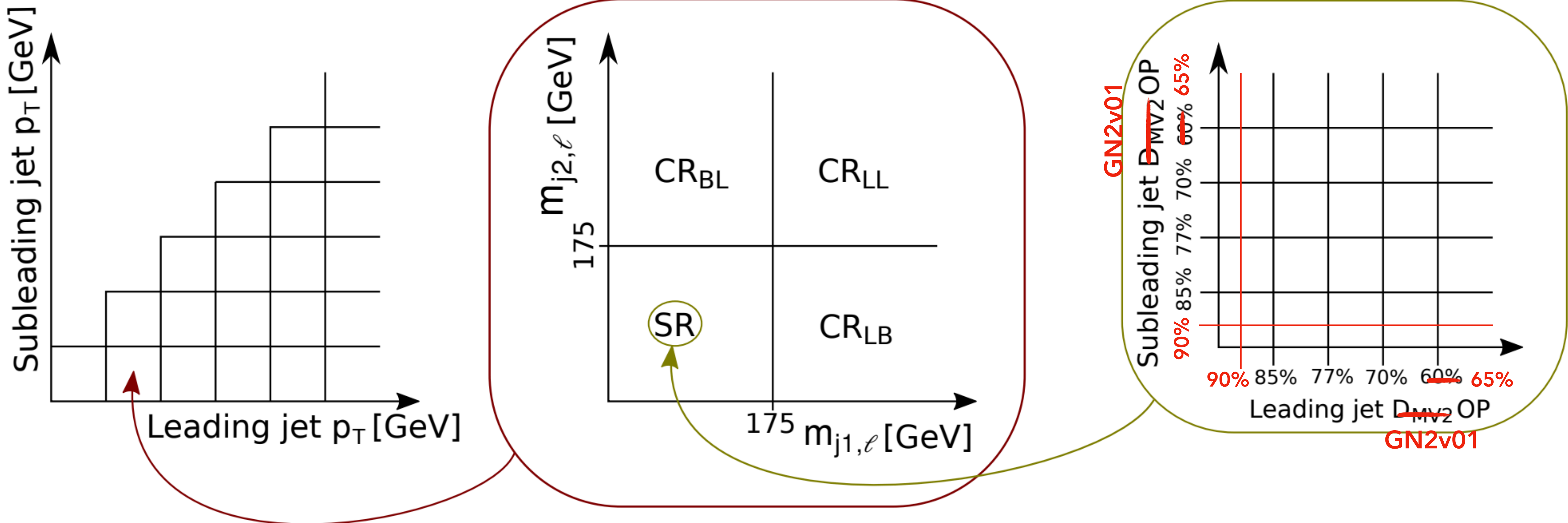
- The **two jets** are paired with the **two leptons**, in the configuration that minimizes  $m_{j_1\ell}^2 + m_{j_2\ell}^2$  (= penalizes configurations where  $m_{j_1\ell}$  and  $m_{j_2\ell}$  are asymmetric).  $\rightarrow$   $m_{j\ell}$  is a proxy of the **top mass!**  $\rightarrow$  For true  $b$ -jets,  $m_{j\ell} \leq m_t \approx 172.5$  GeV.
- In all regions,  $m_{j_1\ell} > 20$  GeV and  $m_{j_2\ell} > 20$  GeV.
- SR =  $m_{j_1\ell} < 175$  GeV and  $m_{j_2\ell} < 175$  GeV.
- CR<sub>BL</sub> =  $m_{j_1\ell} < 175$  GeV and  $m_{j_2\ell} \geq 175$  GeV.
- CR<sub>LB</sub> =  $m_{j_1\ell} \geq 175$  GeV and  $m_{j_2\ell} < 175$  GeV.
- CR<sub>LL</sub> =  $m_{j_1\ell} \geq 175$  GeV and  $m_{j_2\ell} \geq 175$  GeV.

# $b$ -tagging efficiency measurement: fitting strategy

The  $b$ -tagging efficiency is extracted from a **maximum likelihood fit**, performed simultaneously in the SR and in the three CRs.

	Run 2	Run 3
<b>Bins in <math>p_T(j1)</math> and <math>p_T(j2)</math></b>	[20, 30, 40, 60, 85, 110, 140, 175, 250, 400] GeV	[20, 40, 60, 140, 250, 400] GeV
<b>GN2v01 eff. WPs</b>	[90%, 85%, 77%, 70%, 65%]	[90%, 85%, 77%, 70%, 65%]
<b>Total number of bins</b>	$(9 \times 10) / 2 \times (3 + 1 \times 6 \times 6) = 1755$ total bins. $(9 \times 10) / 2 \times (6 \times 6) = 1620$ SR bins.	$(5 \times 6) / 2 \times (3 + 1 \times 6 \times 6) = 585$ total bins. $(5 \times 6) / 2 \times (6 \times 6) = 540$ SR bins.

= 4 regions, 5 bins  
in GN2 score, and  
10 or 6 bins in  $p_T$   
 $\times 2$  jets.



# $b$ -tagging efficiency measurement: fitting strategy

Extended binned likelihood:

$$\mathcal{L}(\nu(\theta)) = \prod_{i=1}^N \frac{\nu_i^{n_i}}{n_i!} e^{-\nu_i} \quad \longrightarrow$$

- $N$  = total number of bins.
- $\nu_i$  ( $n_i$ ) = observed (expected) number of events in the  $i$ -th bin.

- ➔ In the **CRs**, the expected number of events depend from the leading jet  $p_T$  bin  $T^m$  and the subleading jet  $p_T$  bin  $T^n$ , and receives contributions from each of the four flavor fractions  $bb$ ,  $bl$ ,  $lb$ , and  $ll$ .

$$\nu_{CR}(T^m, T^n) = c_{bb}^{m,n} \nu_{CR,bb}^{m,n} + c_{bl}^{m,n} \nu_{CR,bl}^{m,n} + c_{lb}^{m,n} \nu_{CR,lb}^{m,n} + c_{ll}^{m,n} \nu_{CR,ll}^{m,n}$$

- In the **SRs**, the expected number of events depend also from the  $b$ -tagger eff. WP  $O^k$  ( $O^p$ ) of the (sub) leading jet!

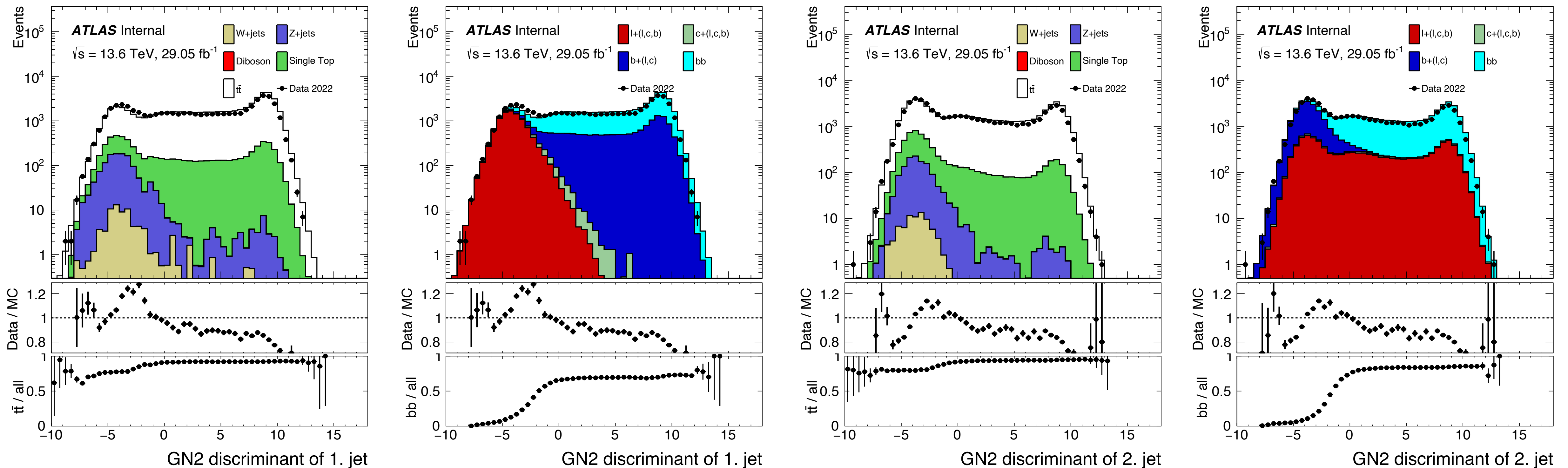
$$\begin{aligned} \nu_{SR}(T^m, T^n, O^k, O^p) = & c_{bb}^{m,n} \nu_{SR,bb}^{m,n} \cdot \mathcal{P}_b(O^k | T^m) \cdot \mathcal{P}_b(O^p | T^n) \\ & + c_{bl}^{m,n} \nu_{SR,bl}^{m,n} \cdot \mathcal{P}_b(O^k | T^m) \cdot \mathcal{P}_l(O^p | T^n) \\ & + c_{lb}^{m,n} \nu_{SR,lb}^{m,n} \cdot \mathcal{P}_l(O^k | T^m) \cdot \mathcal{P}_b(O^p | T^n) \\ & + c_{ll}^{m,n} \nu_{SR,ll}^{m,n} \cdot \mathcal{P}_l(O^k | T^m) \cdot \mathcal{P}_l(O^p | T^n) \end{aligned}$$

- $\nu_{CR(SR)ij}^{m,n}$  = constant param. ➔ Yields of events with true jet flavors  $i$  and  $j$  in each  $p_T$  bin (from MC).
- $c_{ij}^{m,n}$  = unconstrained NPs. ➔ Correction factor for the yields of the different flavor fractions in each  $p_T$  bin.
- $\mathcal{P}_b(O^k | T^m)$  = **POIs**. ➔ Probability that a true  $b$ -jet with  $p_T$  bin =  $T^m$  is selected by the  $O^k$  eff. WP.
- $\mathcal{P}_l(O^k | T^m)$  = constant param. ➔ Probability that a true light /  $c$ -jet with  $p_T$  bin =  $T^m$  is selected by the  $O^k$  eff. WP.

# $b$ -tagging efficiency measurement: data / MC agreement

- As a first step, we had a look at the **data / MC agreement**, for some interesting variables in our analysis (= leading and subleading jet's  $\eta$  and  $p_T$ , the  $m_{j\ell}$  variable, and the GN2 discriminant).
- For the MC, we tried to look at the different contributions from both the **physics process decomposition** (=  $t\bar{t}$ , single top, Z+jets, W+jets, and di-boson samples), and the **jet's flavor decomposition**.

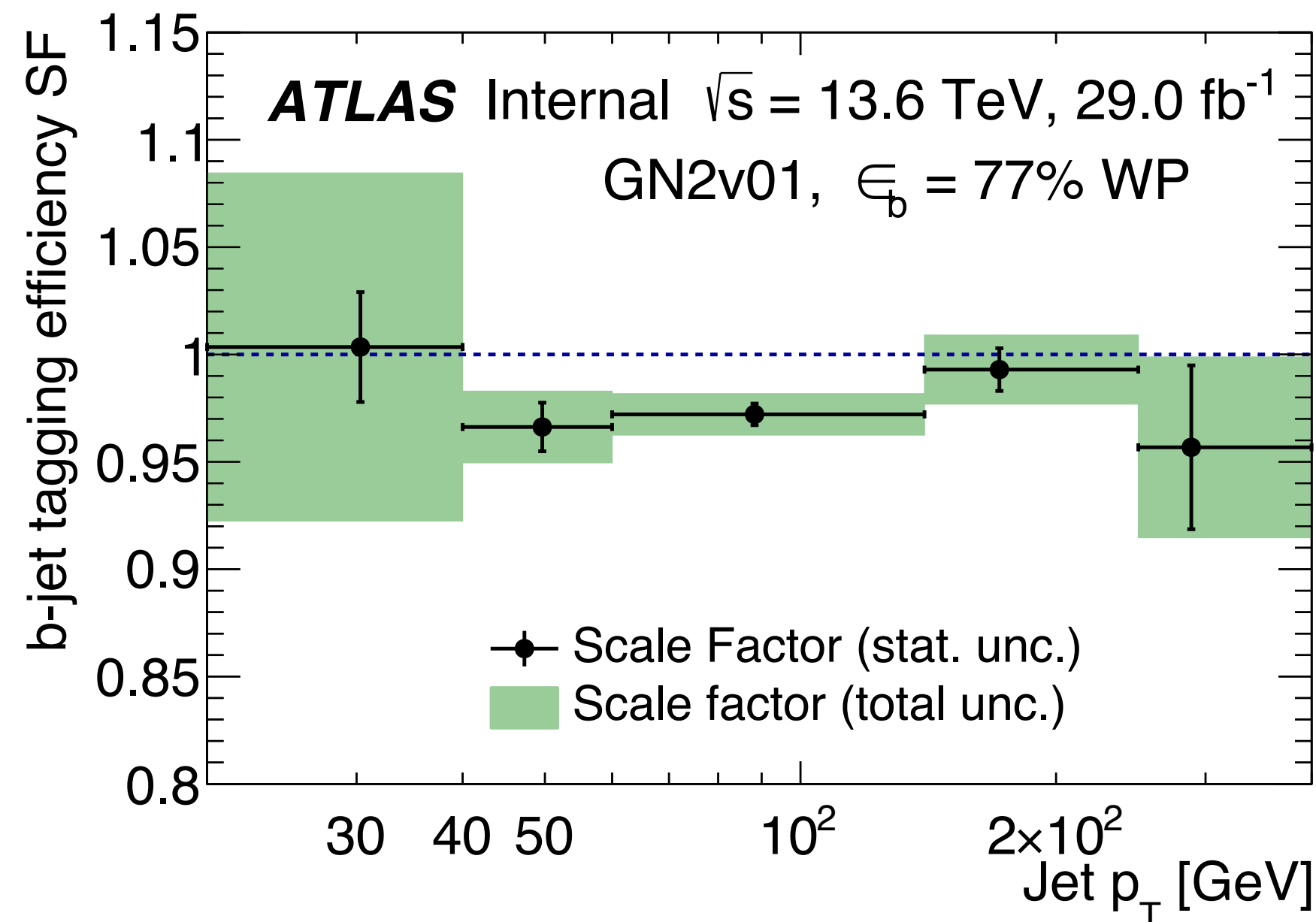
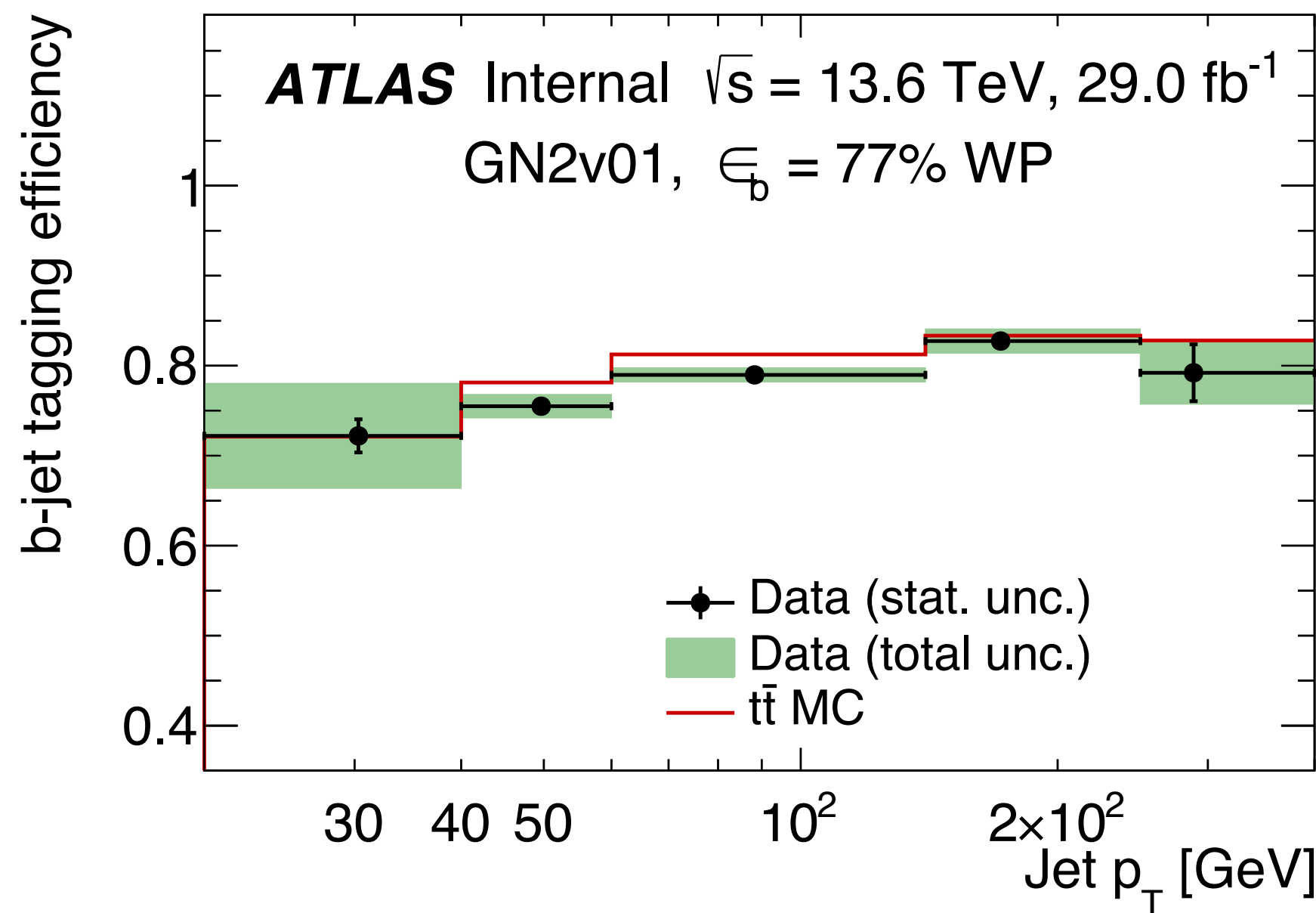
Other variables are shown in backup slides!



# $b$ -tagging efficiency measurement: Run 3 (2022 data)

- The shape of the GN2 discriminant, split between the contribution to each true jet flavor fraction for the leading and subleading jet ( $= bb, bl, lb, ll$ ), is the final observable in our fit!
- After extracting the histograms of the GN2 discriminant in the analysis regions, we are ready to look at the  $b$ -tagging efficiency, and to the Scale Factors.


$b$ -tagging efficiency and SFs as a function of the jet  $p_T$ , for the 77% eff. WP of GN2



# $b$ -tagging efficiency measurement: systematic uncertainties

The following systematic uncertainties (that may affect our results) are considered in the fit.

 Almost all included for the Run 3 results!

Theory	Modelling uncertainties	<ul style="list-style-type: none"> <li>Alternative PDF set correlated tt-bar and single top (difference between NNPDF30_nlo_as_0118 and PDF4LHC).</li> <li>Matrix element unc. for tt-bar (hdamp and pthard parameters).</li> <li>Parton shower for tt-bar (difference between Pythia8 and Herwig7).</li> <li>Modelling unc. for single top difference between DS and DR).</li> </ul>
	Scale variations, PDF variations, ISR & FSR	<ul style="list-style-type: none"> <li>Scale and PDF + <math>\alpha_s</math> for tt-bar and single top.</li> <li>ISR and FSR for tt-bar and single top.</li> </ul>
Exp.	Physics objects, triggers, pileup.	<ul style="list-style-type: none"> <li>Pile-up modelling;</li> <li>Electron and muon trigger efficiency;</li> <li>Electron energy scale and resolution;</li> <li>Jet energy scale and resolution;</li> <li>Jet vertex tagger efficiency; </li> <li>Muon energy calibration and identification;</li> <li>Missing energy reconstruction.</li> </ul> <p>For now excluded, since negative weights variations cause problems in the fit stability (= under investigation).</p>
	Custom	<ul style="list-style-type: none"> <li>Fake lepton correction.</li> <li><b>Light flavor jet mis-tagging efficiency.</b></li> </ul>

- We assigned a **preliminary 100%** uncertainty on the **light jet efficiency** (= used as input in our fit).

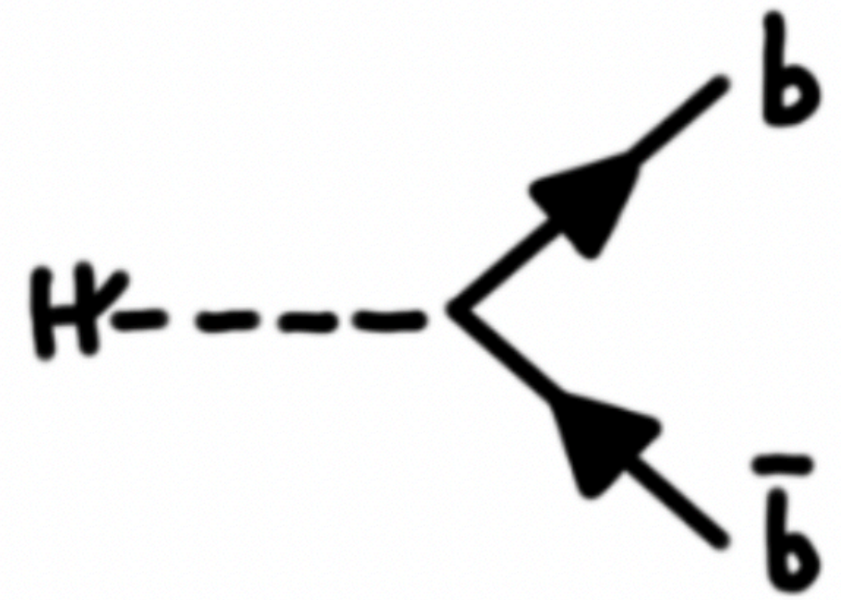
 **Strong impact** at low  $p_T$  in **high efficiency bins** (e.g. 90% WP).

- Will change**, once the **light jet mis-tagging efficiency measurement is ready!**

Estimating the impact of GN2-based  $b$ -tagging  
for  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis

# $b$ -tagging in the $HH \rightarrow b\bar{b}\gamma\gamma$ analysis

## $b$ -jet requirements for the $HH \rightarrow b\bar{b}\gamma\gamma$



- **Run 2**: exactly 2  $b$ -jets using the **77% efficiency WP** of **DL1r**.

- **Run 2 + partial Run 3**: exactly 2  $b$ -jets using the **77% efficiency WP** of **GN2**.

Our **signal** includes **two true  $b$ -jets!**

➔ Same **signal efficiency** but **better bkg. rejection** with new  **$b$ -tagging**.

➔ Improved **rejection of bkg. events** with **light and  $c$ -jets misidentified as  $b$ -jets**.



**Question:** how do we quantify the impact of **improved fake  $b$ -jet rejection** in the  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis?

1. Estimate the **contribution of bkg. events** with **true light** or  **$c$ -flavor jets** misidentified as  $b$ -jets.

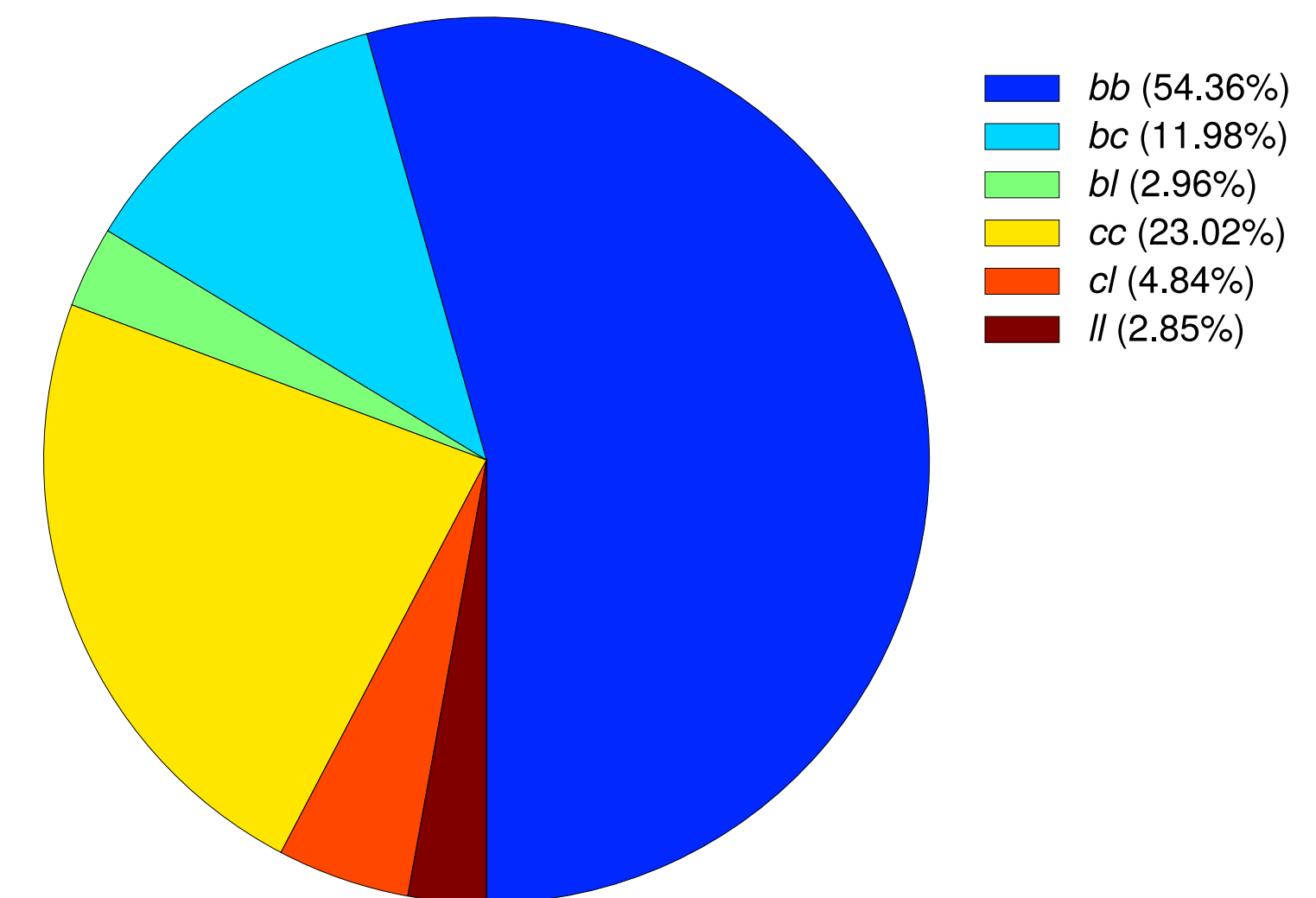
➔ Replacing the **DL1r 77% WP** with the **GN2 77% WP** will help to **suppress the fake  $b$ -jets backgrounds!**

2. Given the 77% signal efficiency, **rescale the contributions of each true jet flavors** with the ratio between the DL1r efficiency and the GN2 efficiency.

➔ Estimated using ROC curves of the two taggers!

3. Estimate the **new bkg. yields**, and evaluate the **impact on the upper limits on the HH signal strength**.

$\gamma\gamma$ +jets true jet flavor fractions.





# c- and light jet rejection with DL1r and GN2

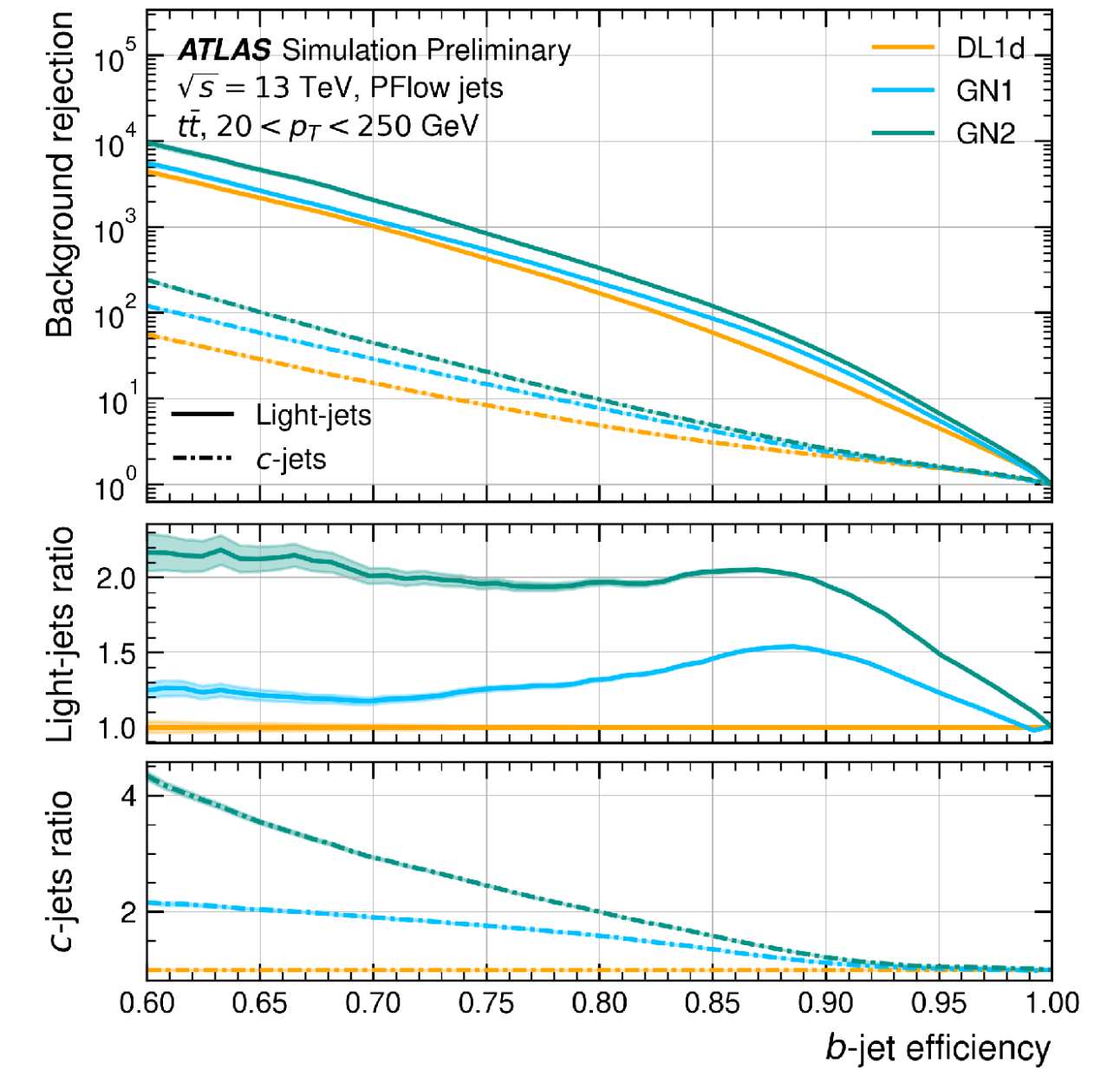
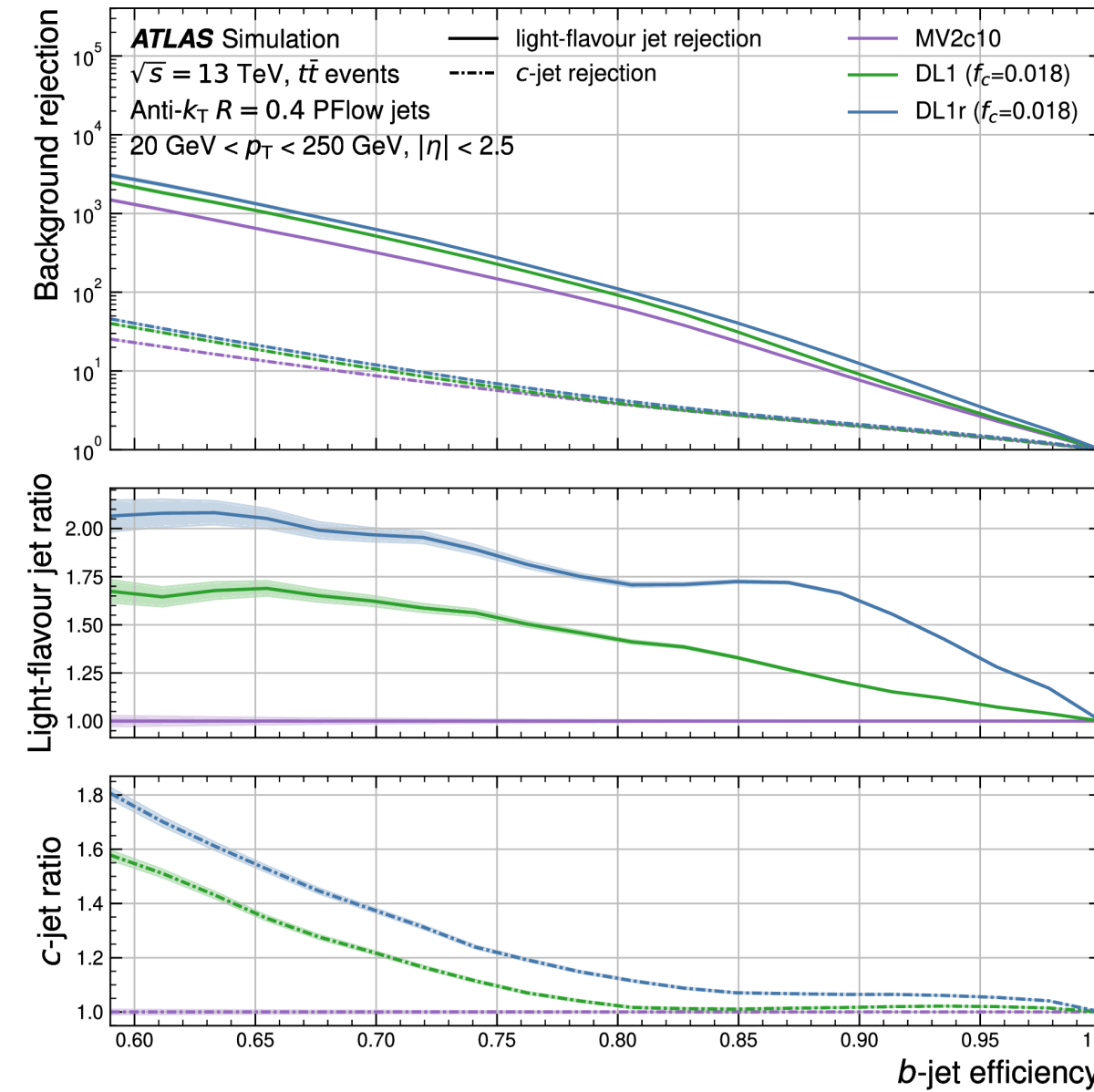
Tagger bkg. efficiency @ 77% b-jet eff. WP

	c-jet efficiency	Light jet efficiency
<b>DL1r</b>	1/6 = 16.6%	1/300 = 0.33%
<b>GN2</b>	1/12 = 8.3%	1/700 = 0.14%

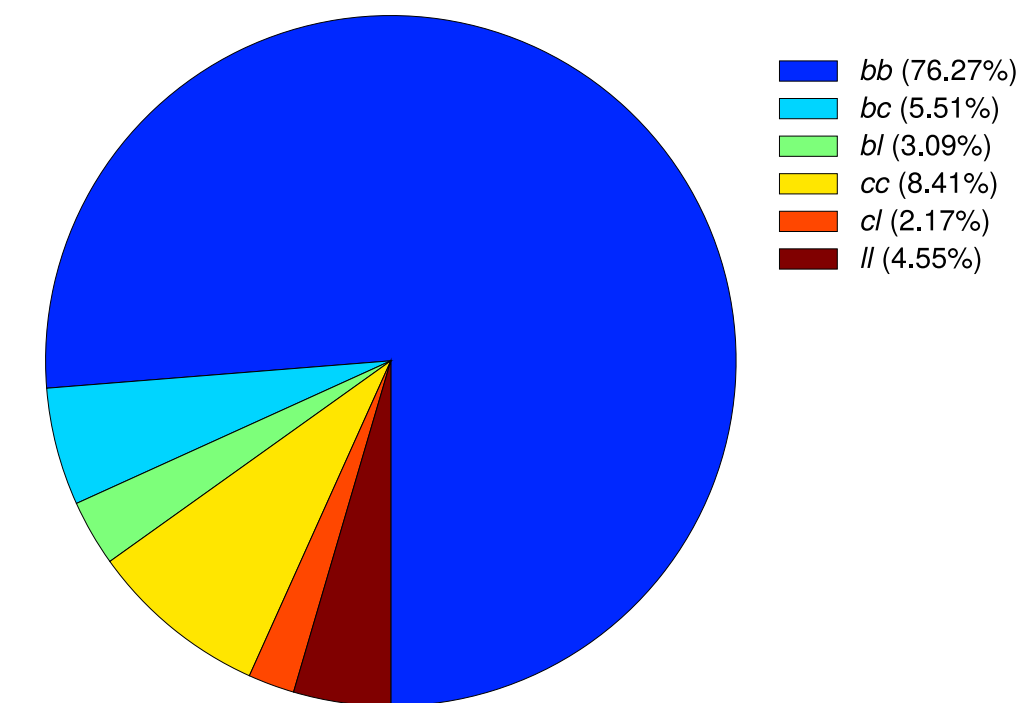
- Given the selected events with **two reco b-jets**, we can evaluate the ratio of the efficiency between DL1r and GN2 WPs for each true flavor fraction.
- When rescaling the true jet flavor contributions with the GN2 / DL1r efficiency ratios, both the true jet flavor composition of the selected (bkg.) events will change, and the overall yields in each analysis category!  $\rightarrow$  Will have an impact on the HH significance!

$\rightarrow$  The **True bb / All purities** for each background samples in each analysis categories are available in backup slides!

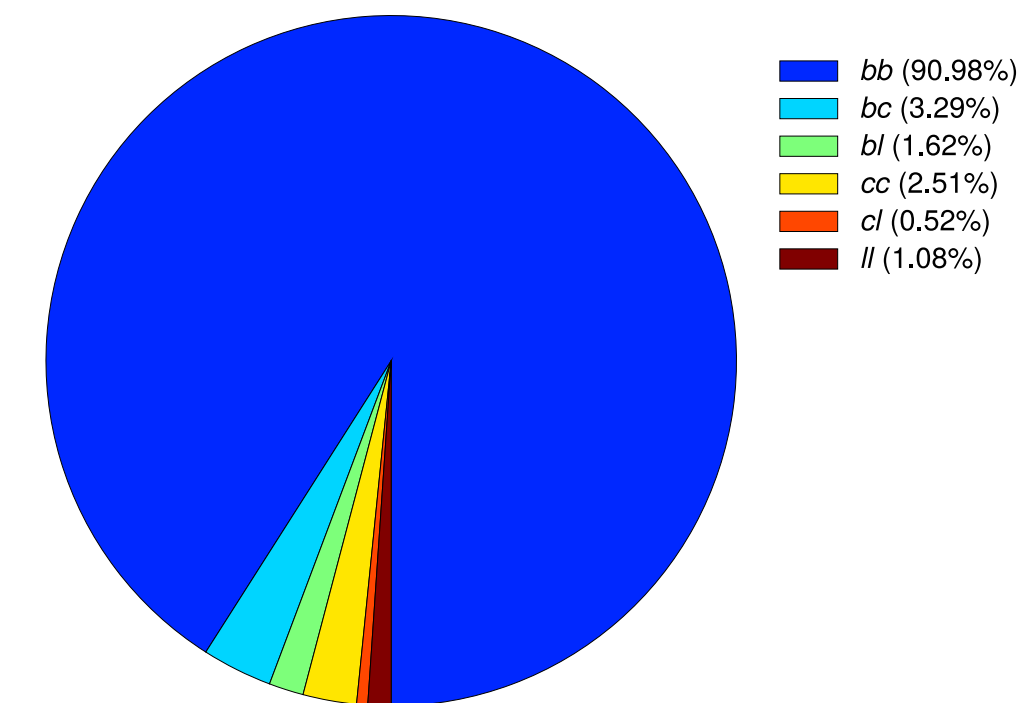
True jet flavor	$\epsilon$ ( 2 reco b   true ij): GN2	$\epsilon$ ( 2 reco b   true ij): DL1r	GN2 / DL1r
bb	$77\% \times 77\% = 59\%$	$77\% \times 77\% = 59\%$	1
bc	$77\% \times 8.3\% = 6.4\%$	$77\% \times 16.6\% = 12.8\%$	0.5
bl	$77\% \times 0.14\% = 0.11\%$	$77\% \times 0.33\% = 0.25\%$	0.44
cc	$8.3\% \times 8.3\% = 0.69\%$	$16.6\% \times 16.6\% = 2.76\%$	0.25
cl	$8.3\% \times 0.14\% = 0.01\%$	$16.6\% \times 0.33\% = 0.05\%$	0.2
ll	$0.14\% \times 0.14\% = 0.0002\%$	$0.33\% \times 0.33\% = 0.001\%$	0.2



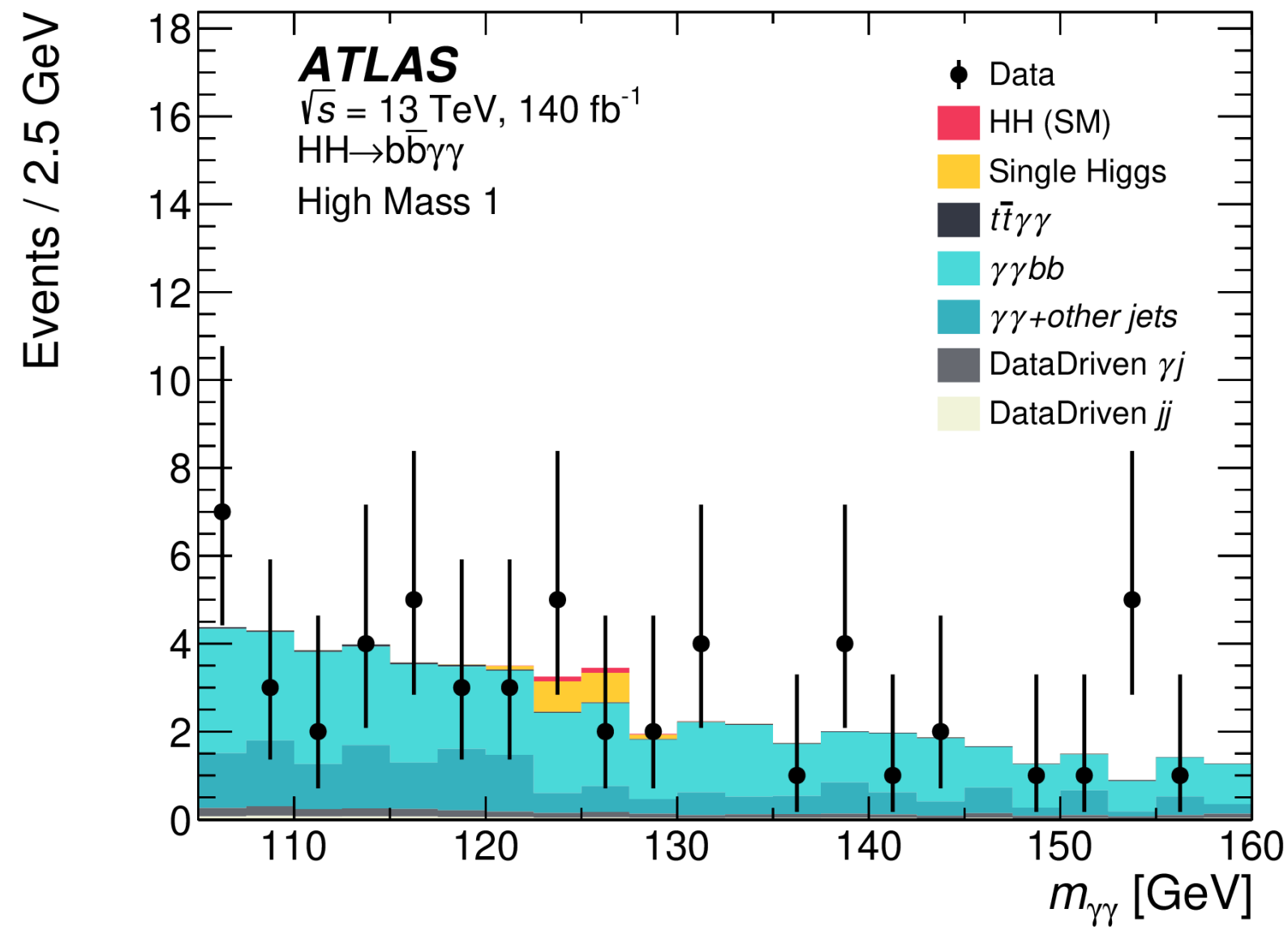
ggH sample (DL1r 77% WP)



ggH sample (GN2 77% WP)



# Background contribution for $HH \rightarrow b\bar{b}\gamma\gamma$



## Two main backgrounds:

- Resonant background** → Estimated using MC.
  - Single Higgs production + jets, where  $H \rightarrow \gamma\gamma$ .
  - Main contribution from ggH, ttH and ZH production modes.
- Continuum background** → Estimated using data (mostly from  $m_{\gamma\gamma}$  sidebands region).
  - Main contribution from di-photon production + additional jets.
  - Single photon + jets and multijet background (where 1 or 2 jets are misidentified as photons) account for < 15%.

Ratio of the expected bkg. yields for each analysis category after replacing the DL1r WP with GN2 WP.

	High Mass 1	High Mass 2	High Mass 3	Low Mass 1	Low Mass 2	Low Mass 3	Low Mass 4	$b\bar{b}\gamma\gamma$ preselection
<b>Samples</b>								
<b>ggH</b>	0.899244	0.914420	0.959389	0.767223	0.909515	0.958215	0.902911	0.838291
<b>ttH</b>	0.867658	0.864079	0.863263	0.887443	0.900553	0.906607	0.900554	0.912129
<b>ZH</b>	0.968675	0.969956	0.978217	0.959593	0.969878	0.970003	0.978440	0.943449
<b>Other H</b>	0.862848	0.836129	0.856764	0.853913	0.852318	0.917792	0.970914	0.839883
<b><math>\gamma\gamma</math>+jets</b>	0.833301	0.880967	0.907797	0.756044	0.812988	0.842126	0.817748	0.689437

## Caveat:

- The GN2 / DL1r expected yield ratio for the continuum bkg. was estimated relying on the true flavor fractions extracted from the  $\gamma\gamma$ +jets sample.
- This ratio is however used to rescale the continuum bkg. yields evaluated in data (following the recipe of the  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis).

→ Bkg. yields reduced of a factor of up to 25% (depending on the samples and the category)!

# Expected improvement on the statistical results

- Expected stat. only exclusion limits are set on the **di-Higgs signal strength** at **95% CL**.

## Expected upper limits on $\mu(\text{HH})$ @ 95% CL.

	Nominal (DL1r)	Improved (GN2)
$\mu(\text{HH})$	4.86	4.70



We observe a **3.3% improvement** in the expected stat. only limit on  $\mu(\text{HH})$ , coming **from the new bkg. rejections only!**



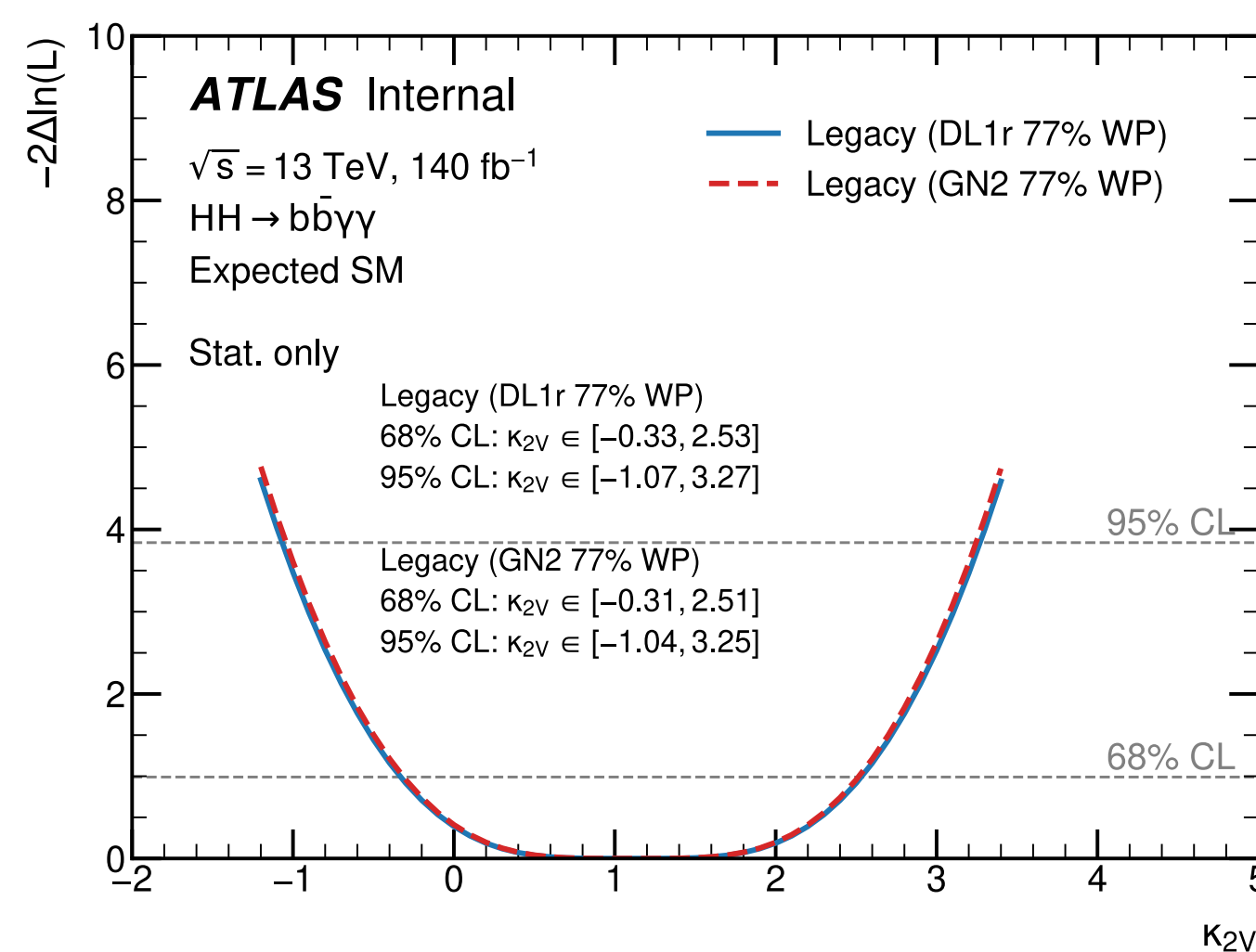
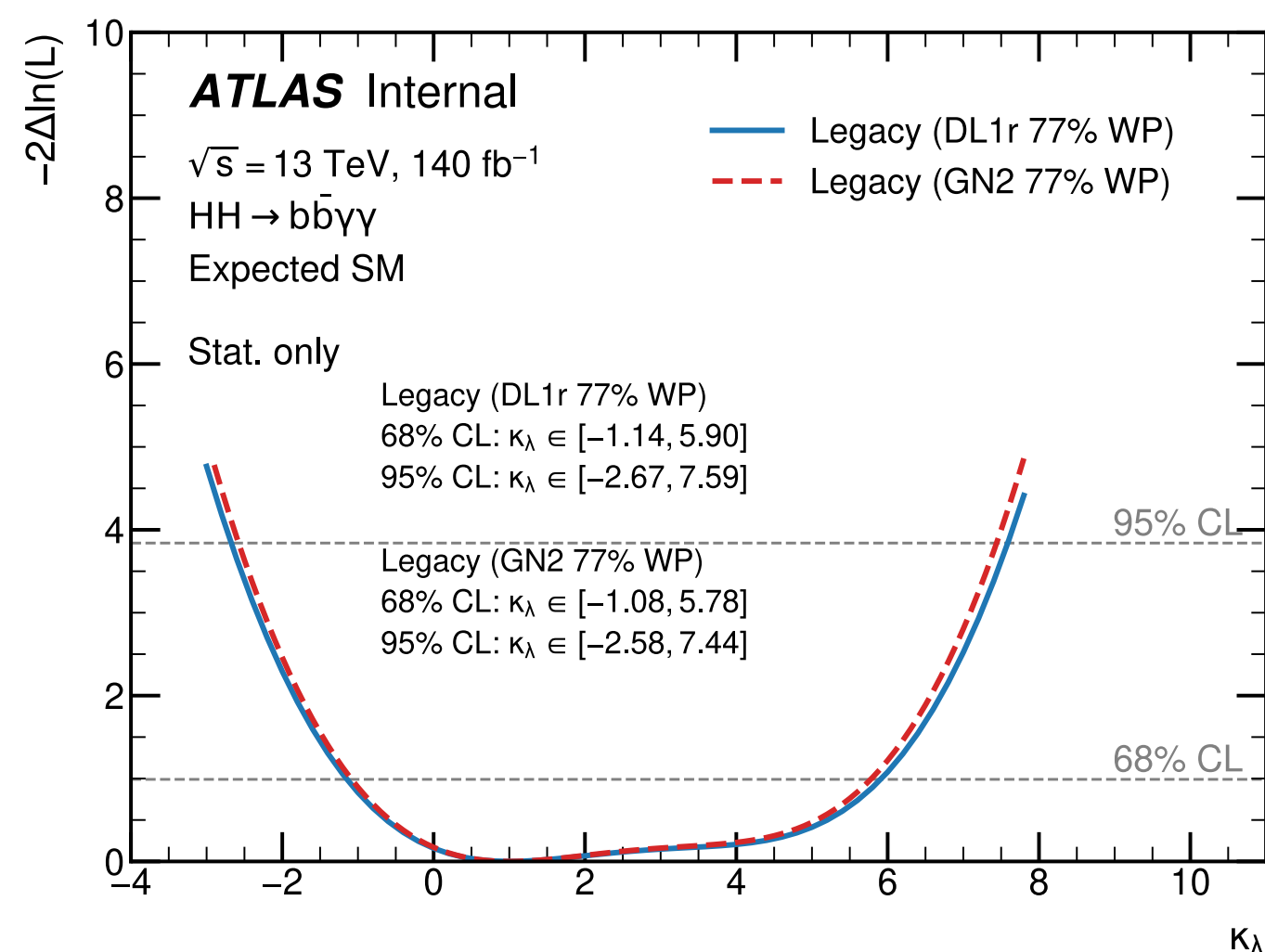
- No retraining of the Legacy BDT or redefinition of the categories involved here.  This improvement is very conservative!

- We also made assumption on the true flavor fraction composition from the MC, and on the GN2 / DL1r bkg. rejection estimated from MC.



Reasonable as long as Scale Factors for DL1r and GN2 are similar.

- We can also derive expected **68%** and **95% confidence intervals** for  $\kappa_\lambda$  and  $\kappa_{2V}$  via a **profile log-likelihood ( $-2\Delta \ln(\text{L})$ ) scan**.

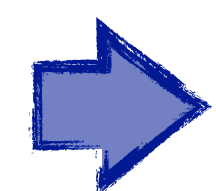


- Replacing the bkg. rejection efficiency of DL1r 77% WP with those of GN2 77% WP seems to **improve the expected constraints** on  $\kappa_\lambda$  by a relative factor of **2.3%**.

- The improvement for the  $\kappa_{2V}$  constraint is a bit **smaller** (1.1%).

# Summary

- A **new generation of GNN-based  $b$ -tagging algorithms** was developed in ATLAS.

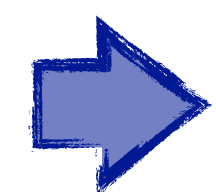


- The **GN2** algorithm is the **current state of the art!**

- We expect **at least a factor  $\times 2$  of improvement** in **fake  $b$ -jets background rejection** w.r.t. the older RNN-based taggers adopted in Run 2 (e.g. DL1r).

- **We look forward** to adopting the new tagger in our **di-Higgs analyses** (= especially in the  $b\bar{b}\gamma\gamma$ ,  $b\bar{b}\tau\tau$ , and  $b\bar{b}b\bar{b}$  channels)!

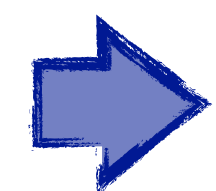
- Before being able to apply the new algorithms in physics analysis, we need to **calibrate** it!



**Calibration of the GN2 WPs = measuring the (mis-)tagging efficiency for  $b$ -jets,  $c$ -jets and light jets in real data!**

- I have been working on the **measurement** of the  **$b$ -jet efficiency**, using **di-leptonic  $t\bar{t}$  events!**

Three different analyses (proceeding in parallel), using each other's results as inputs.



- Efficiency measurement with 2022 data  $\sim$  ready!



- A first version of the efficiency and the corresponding SFs is ready, together with the systematic uncertainties!

- Still working through our to-do list for Run 2 and 2023 data.

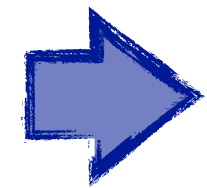


- We have nominal results for Run 2 (no systematic yet), and finalizing sample production for 2023!

**Timescale for releasing the recommendations:  
 $\sim 1$  month!**

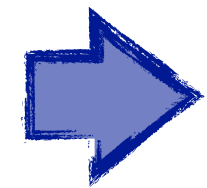
# Summary

- Finally, I have tried to perform a qualitative estimate of the improvement from adopting the GN2 tagger in the  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis.



- The analysis requires **exactly 2  $b$ -jets** selected by the **77% WP** of the **DL1r tagger**.
- **The pseudo-continuous  $b$ -tagging DL1r scores** for the jets are used as **input** for defining the **analysis categories!**

- We tried to adjust the expected background yields in the Legacy  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis, according to the expected fake  $b$ -jet rejection power of the GN2 77% WP compared with the DL1r 77% WP, and we repeated the statistical analysis to extract the expected results.



- We observed a  $\sim 3\%$  improvement, coming from the new bkg. rejection only!
- **Caveat:** we are being very conservative in this estimation.



- o No re-definition of the category is involved here (i.e. the categories are still relying on the DL1r scores).

- o We are only studying the effect of the improved bkg. rejection!

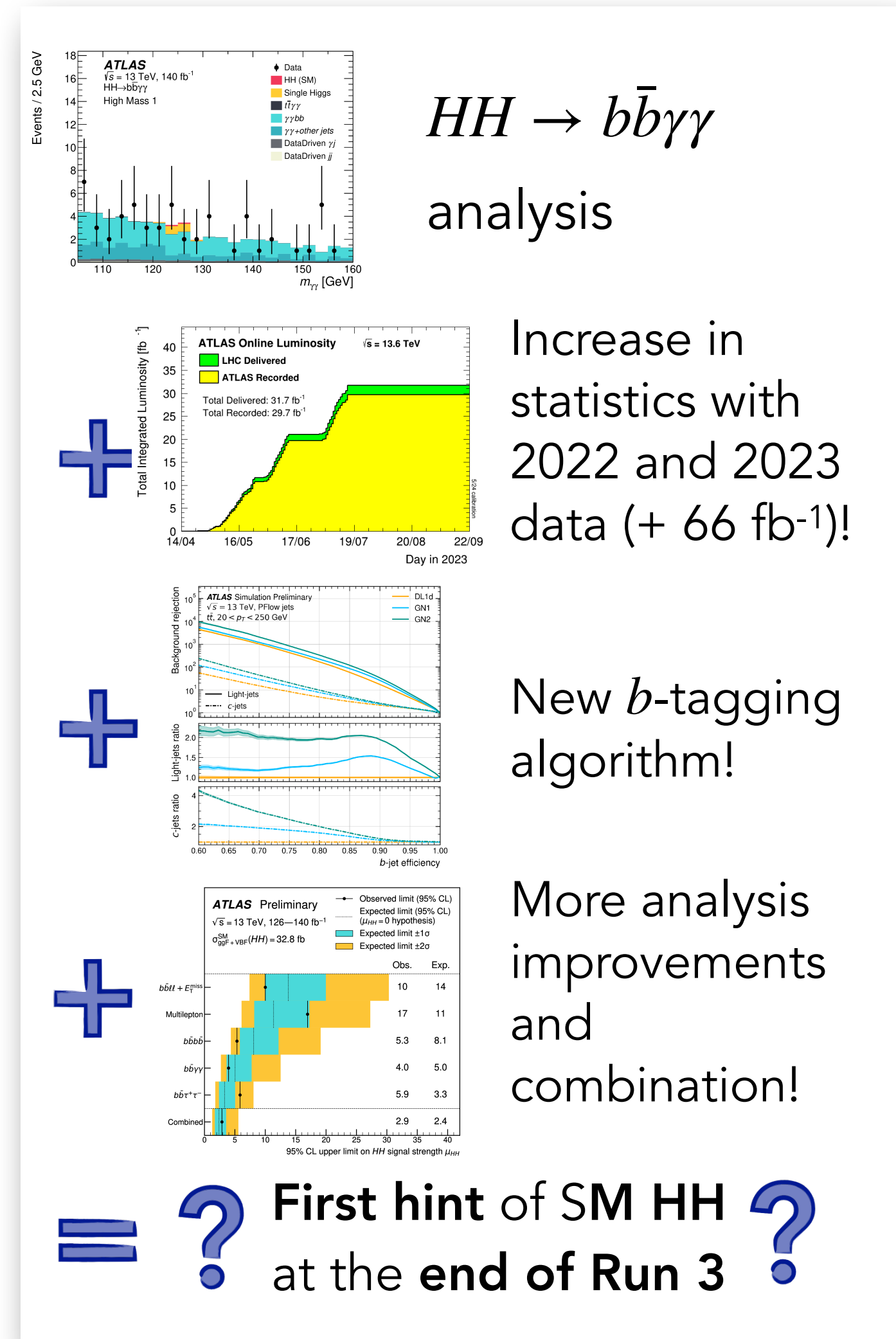


Since our analysis is statistically limited and our signal includes 2 true  $b$ -jets, using a higher eff. WP (e.g. 90%) will bring a much stronger improvement!

- New Run 2 + partial Run 3  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis is already testing the GN2 algorithm!



**Stay tuned for new di-Higgs results!**

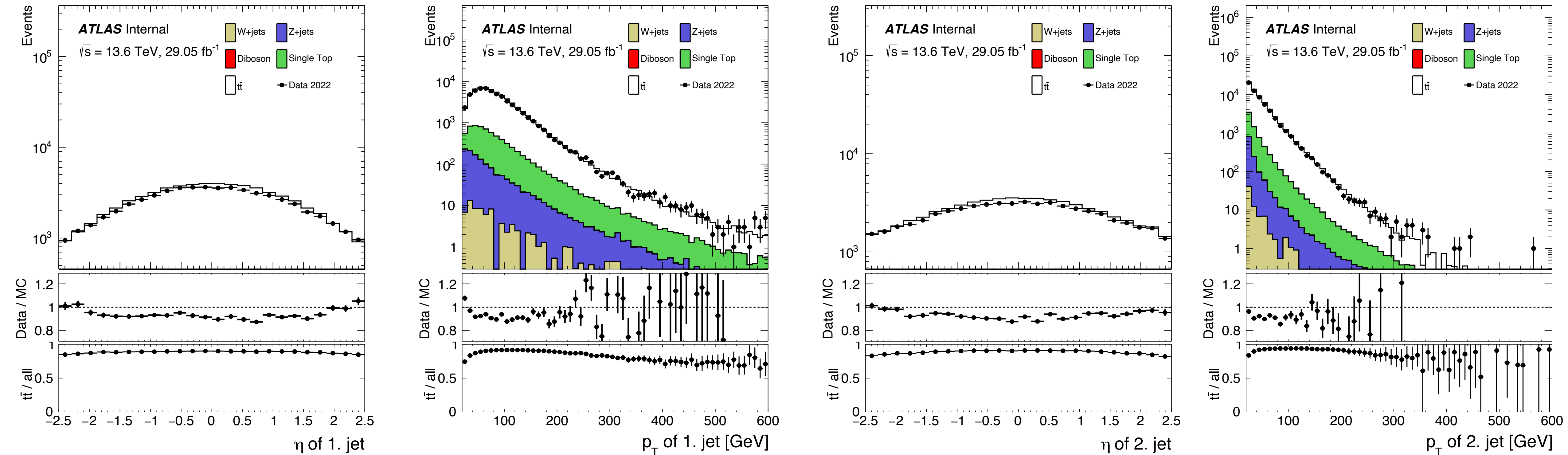


**Thank you for your attention!**

**Backup**

# $b$ -tagging efficiency measurement: data / MC agreement

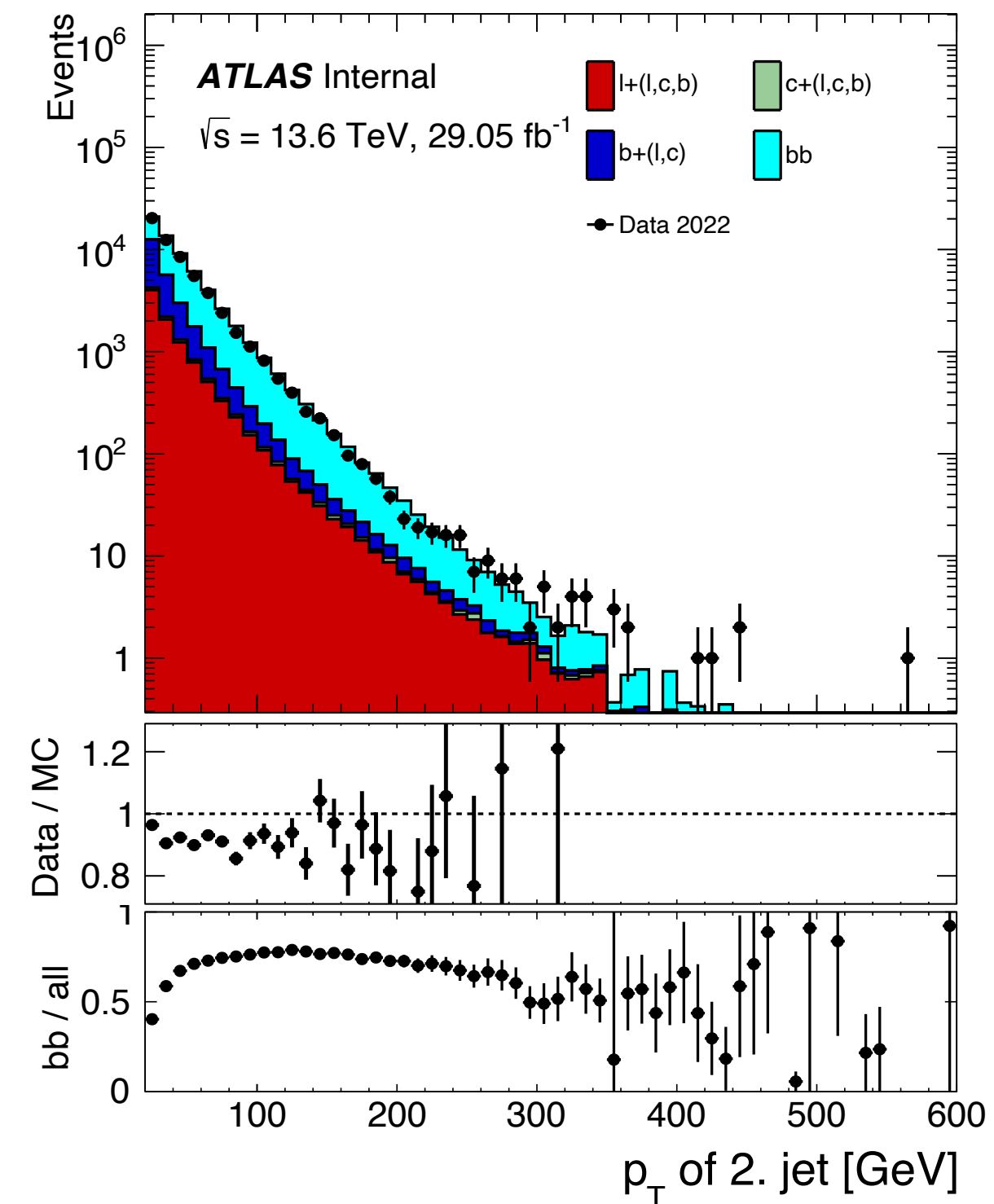
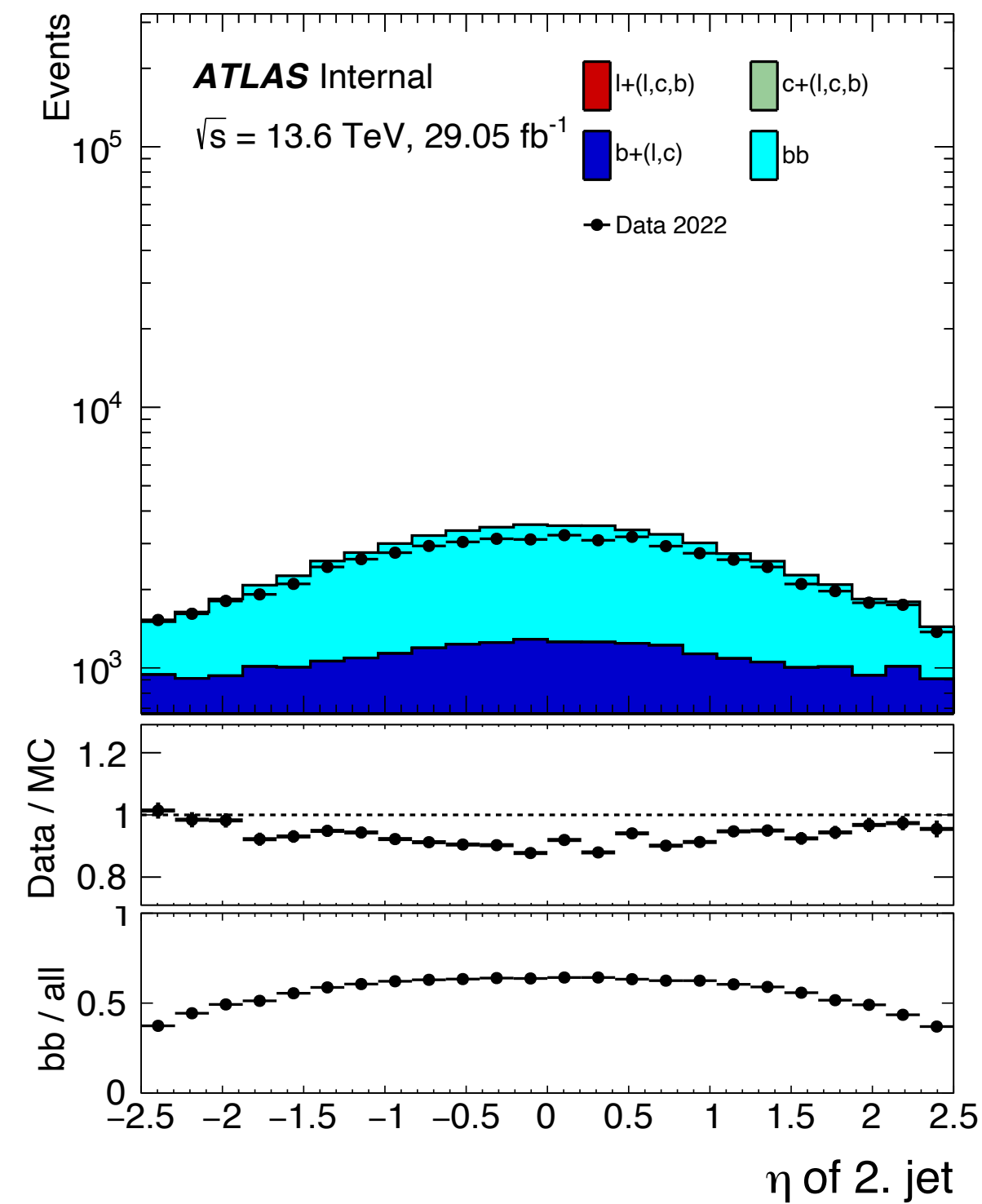
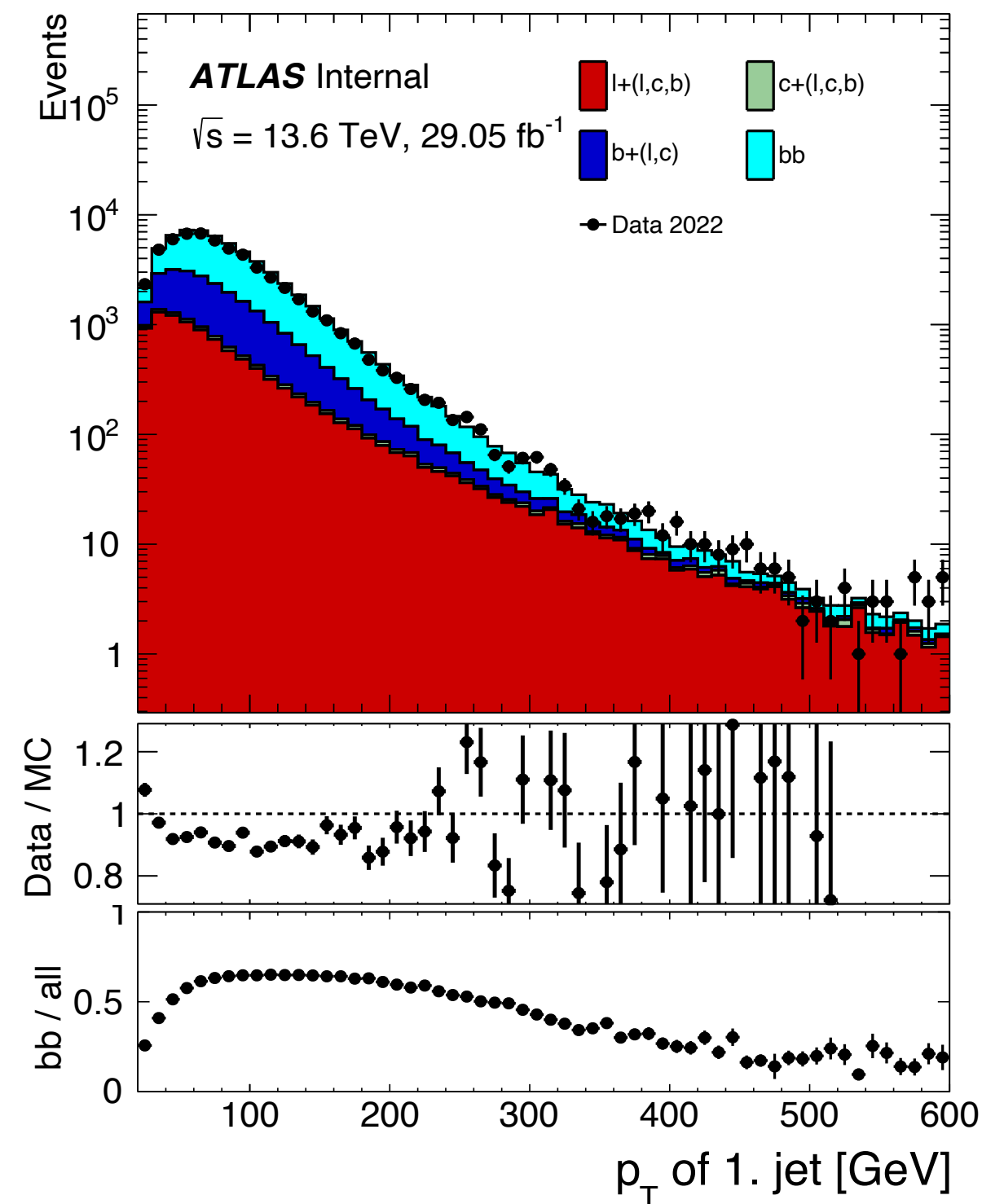
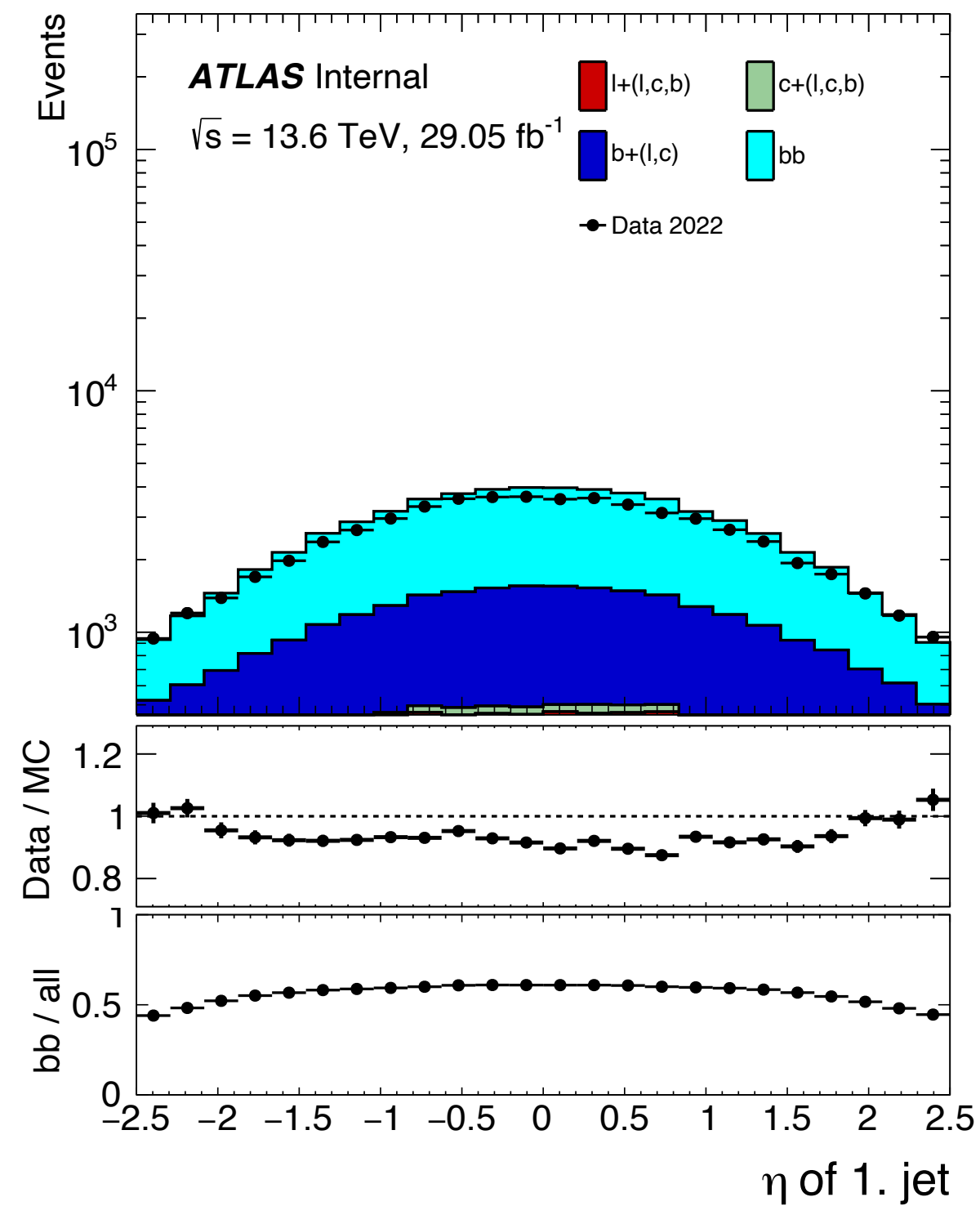
$\eta$  and  $p_T$  of the leading and subleading jets, showing the contribution from each physics process





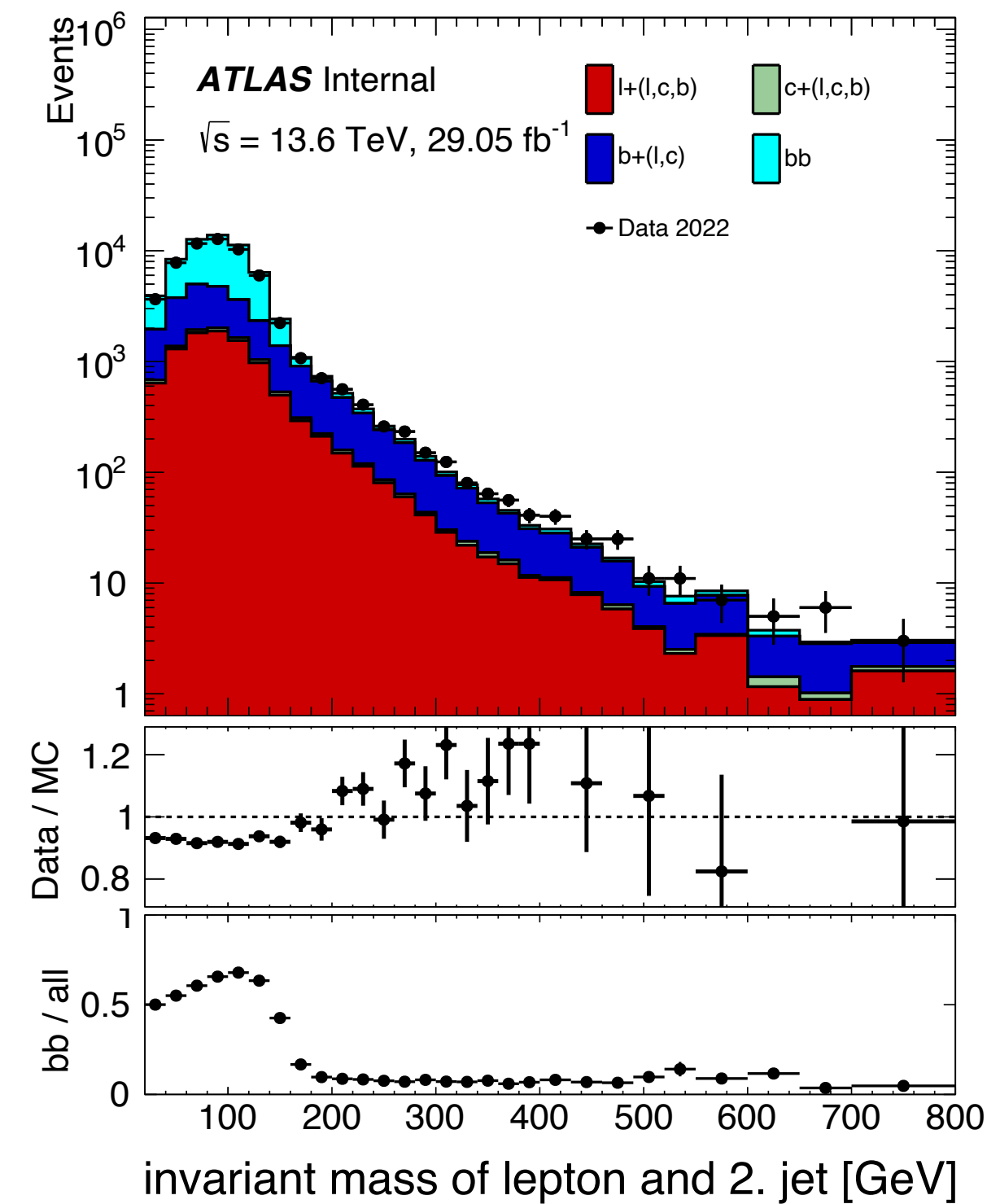
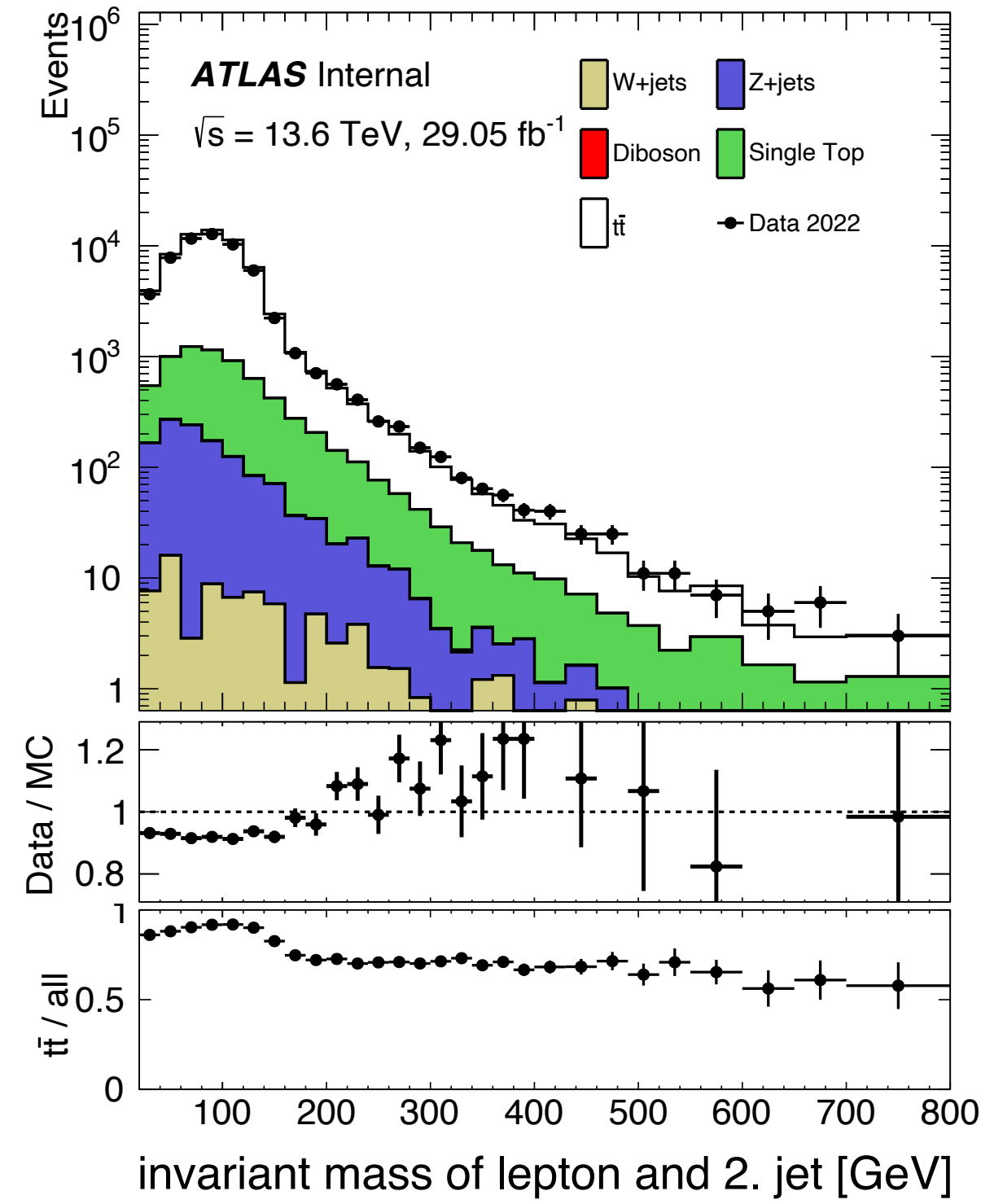
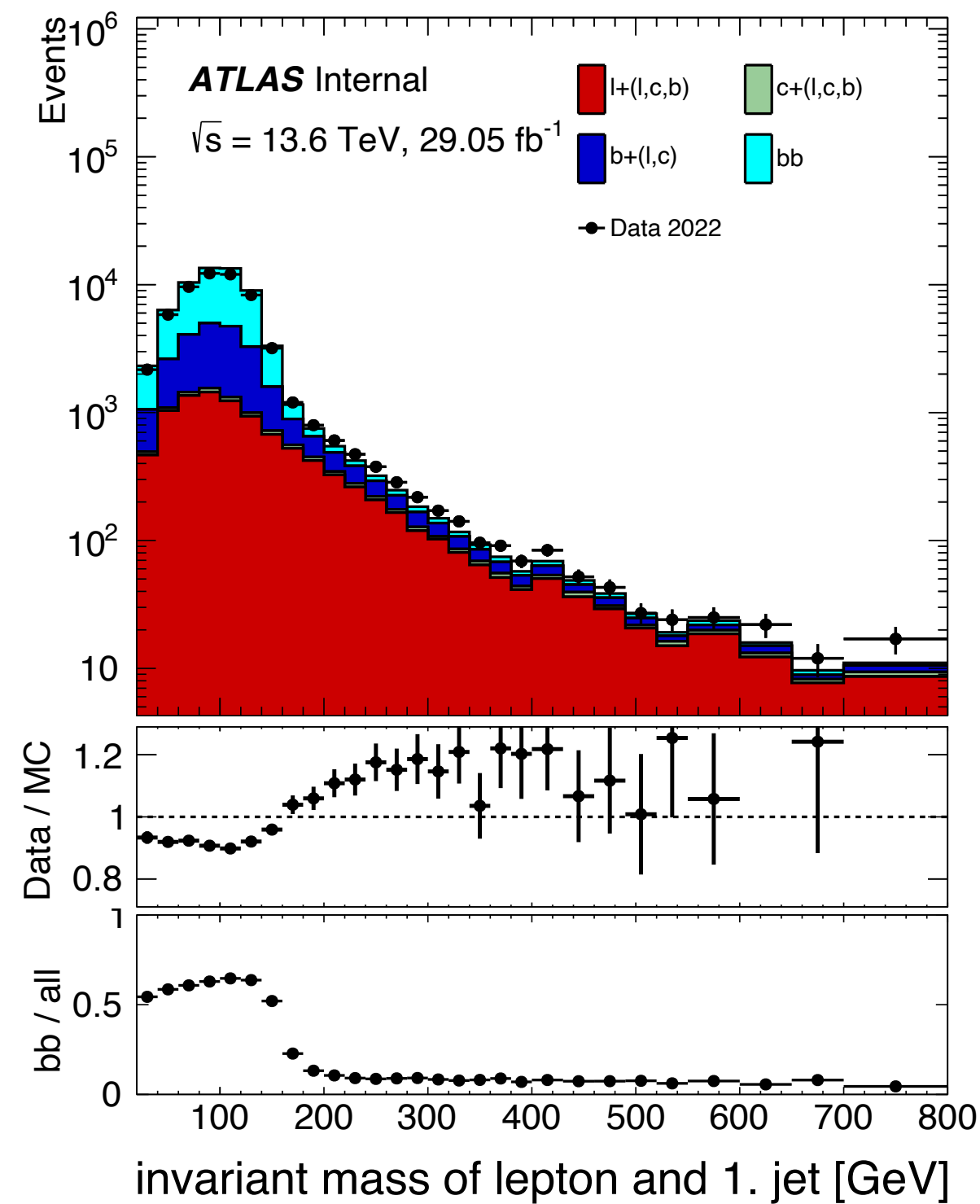
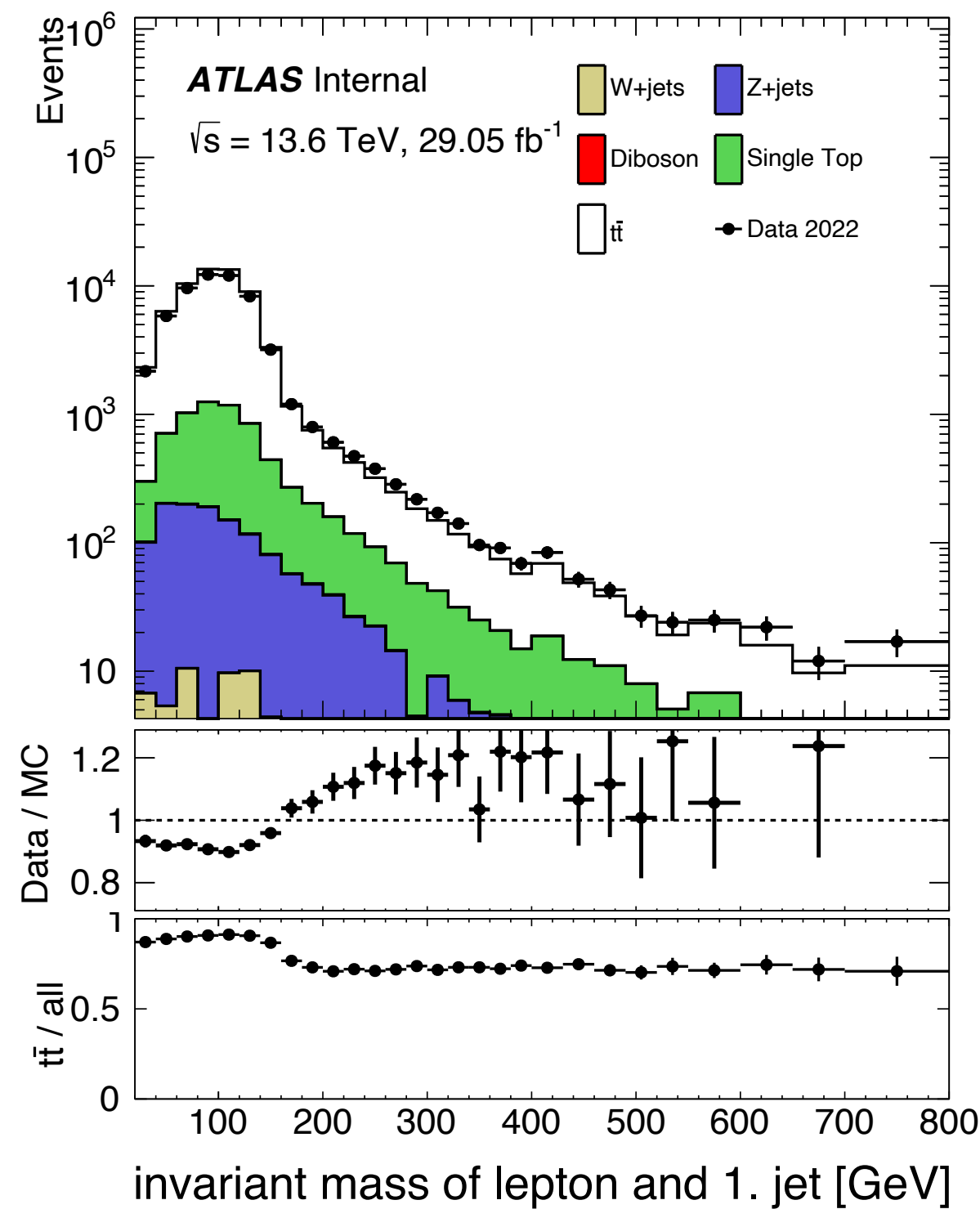
# $b$ -tagging efficiency measurement: data / MC agreement

$\eta$  and  $p_T$  of the leading and subleading jets, showing the contribution from the true jet flavor fractions



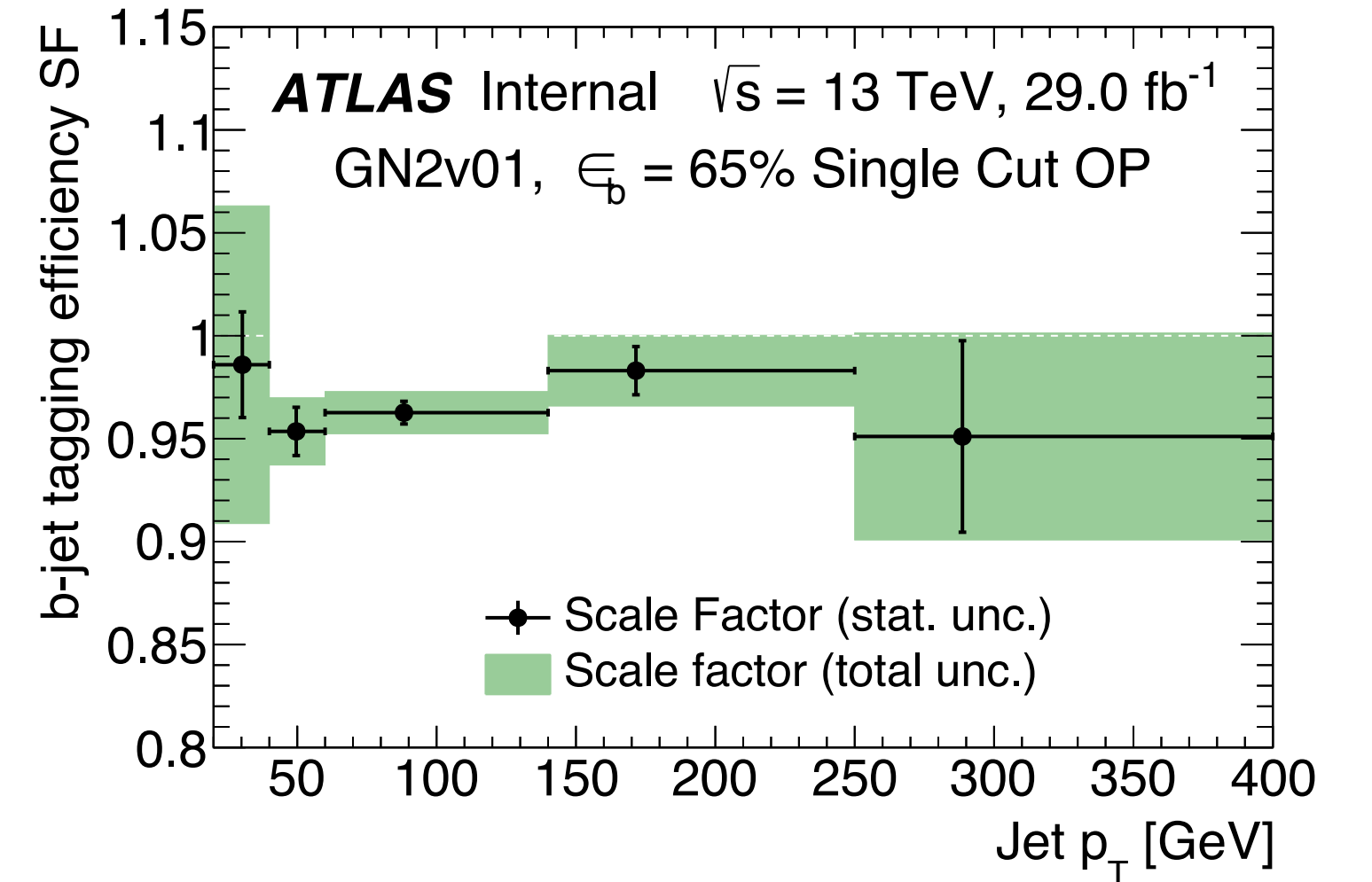
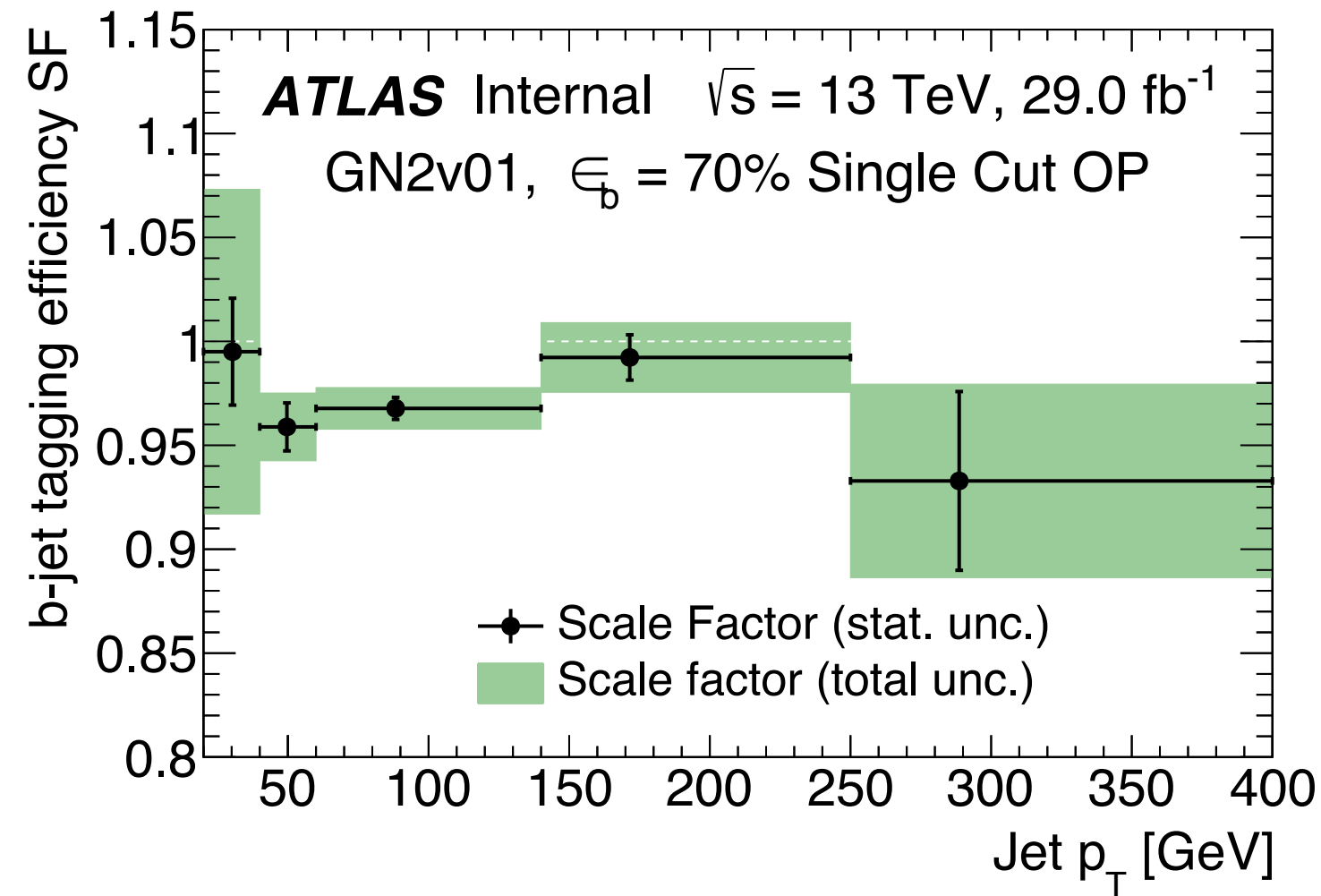
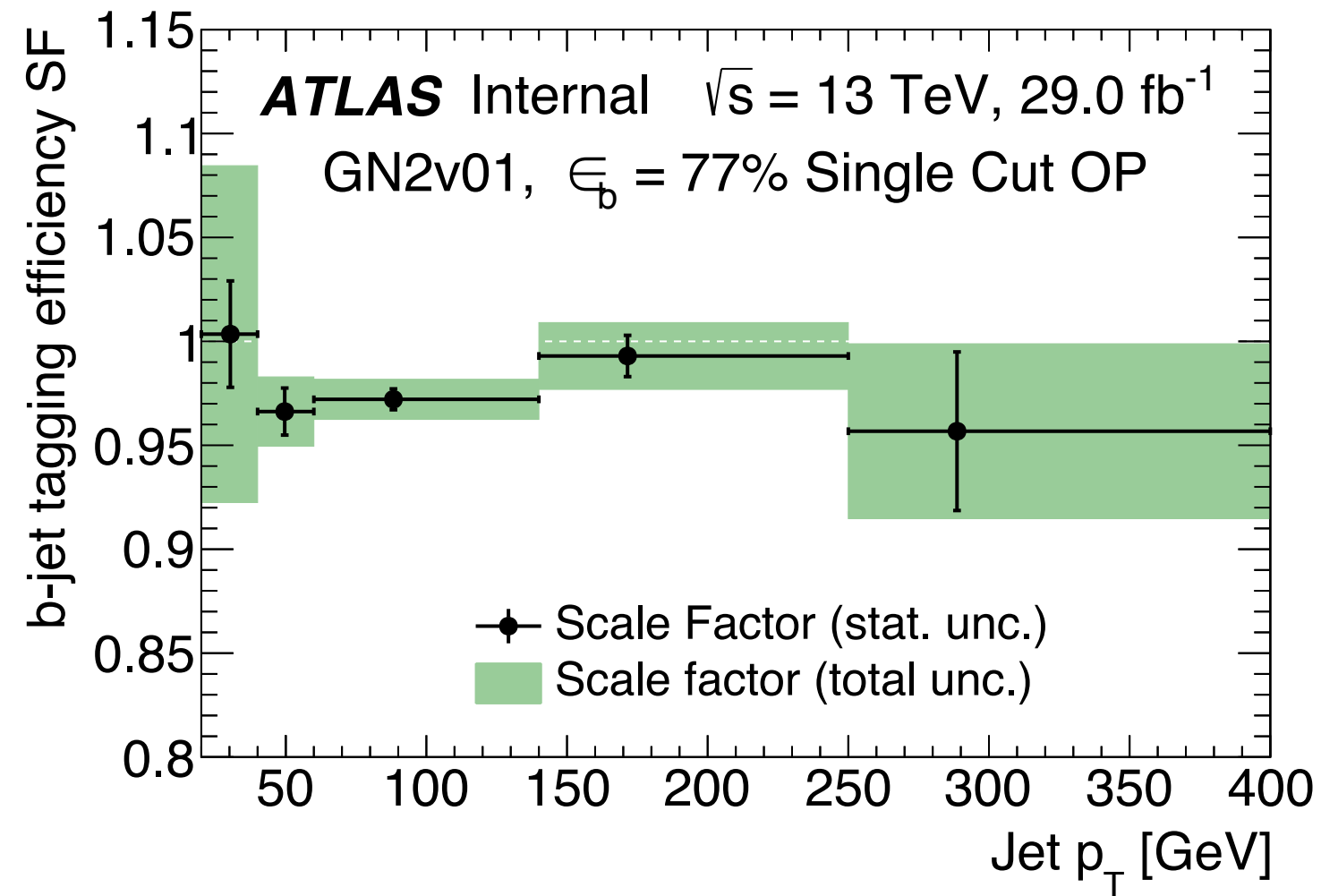
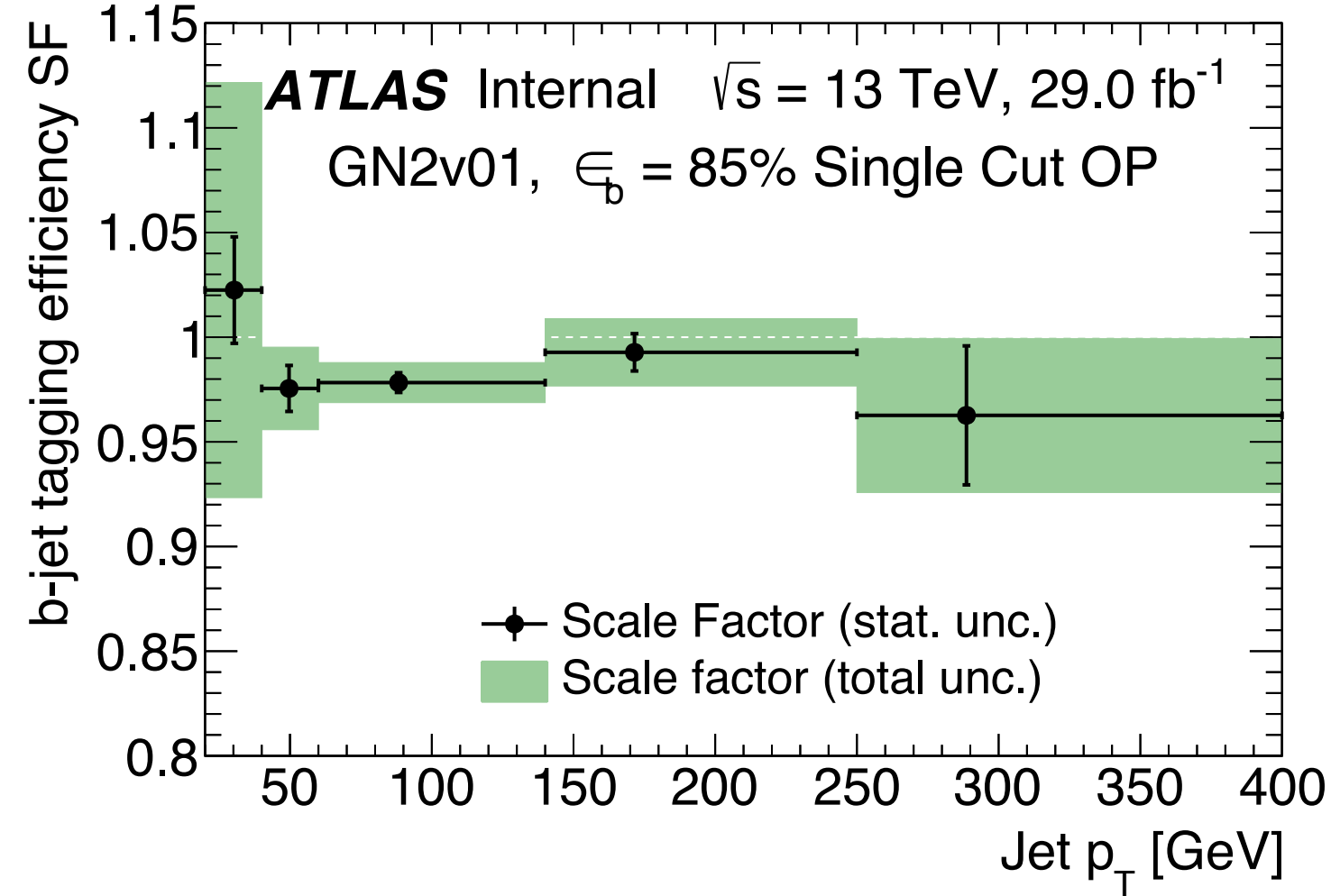
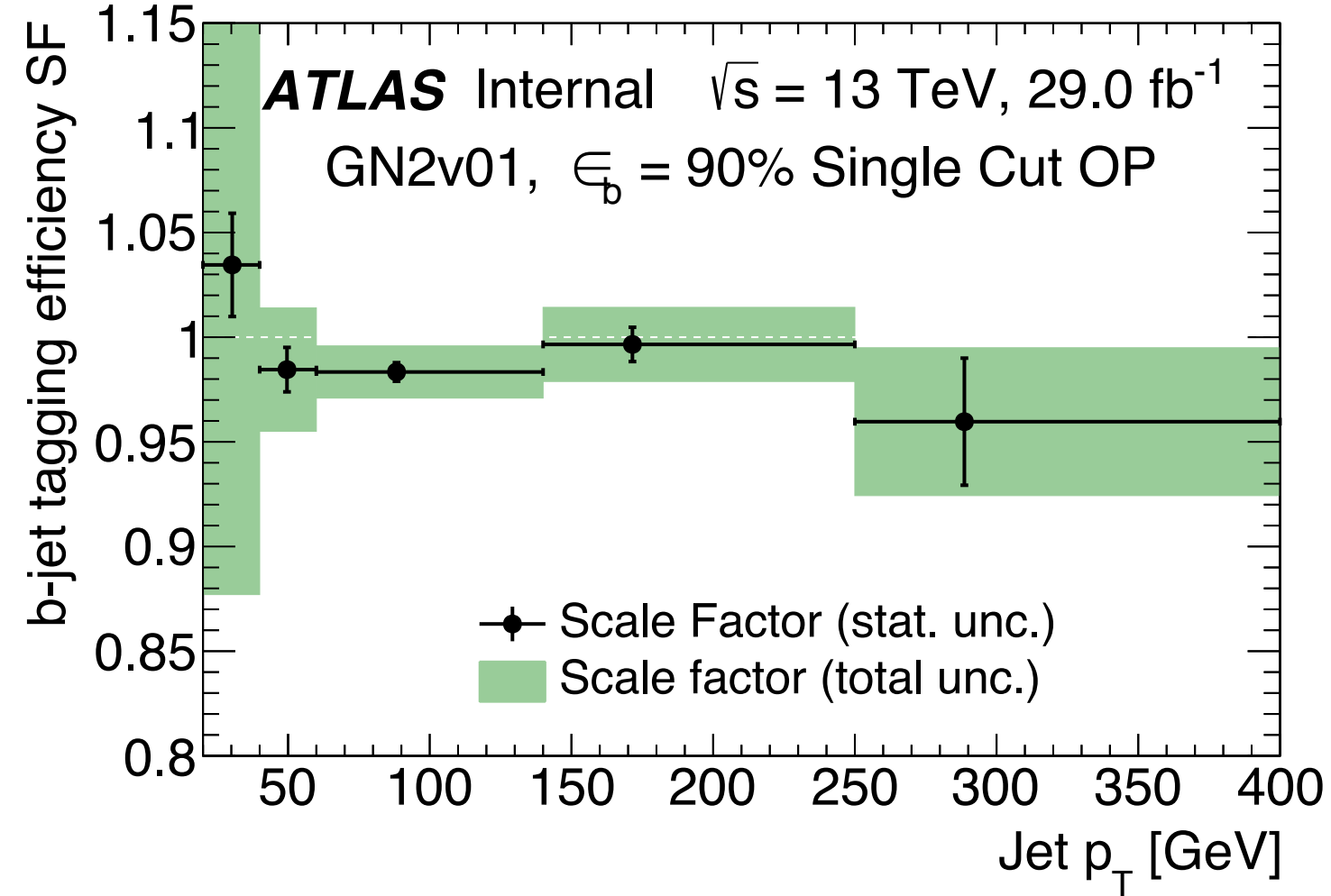
# $b$ -tagging efficiency measurement: data / MC agreement

## Lepton + jet invariant mass for leading and subleading jets



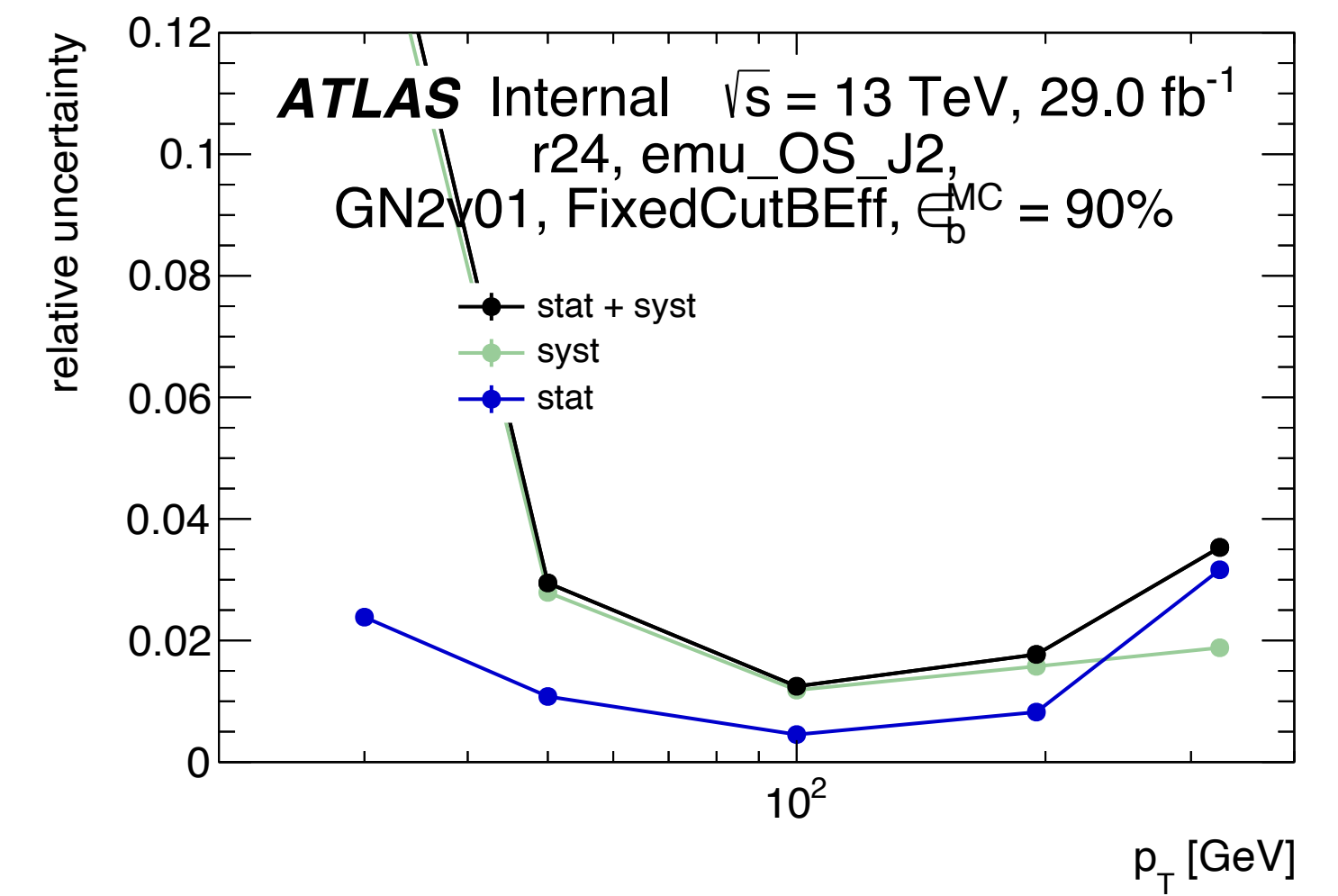
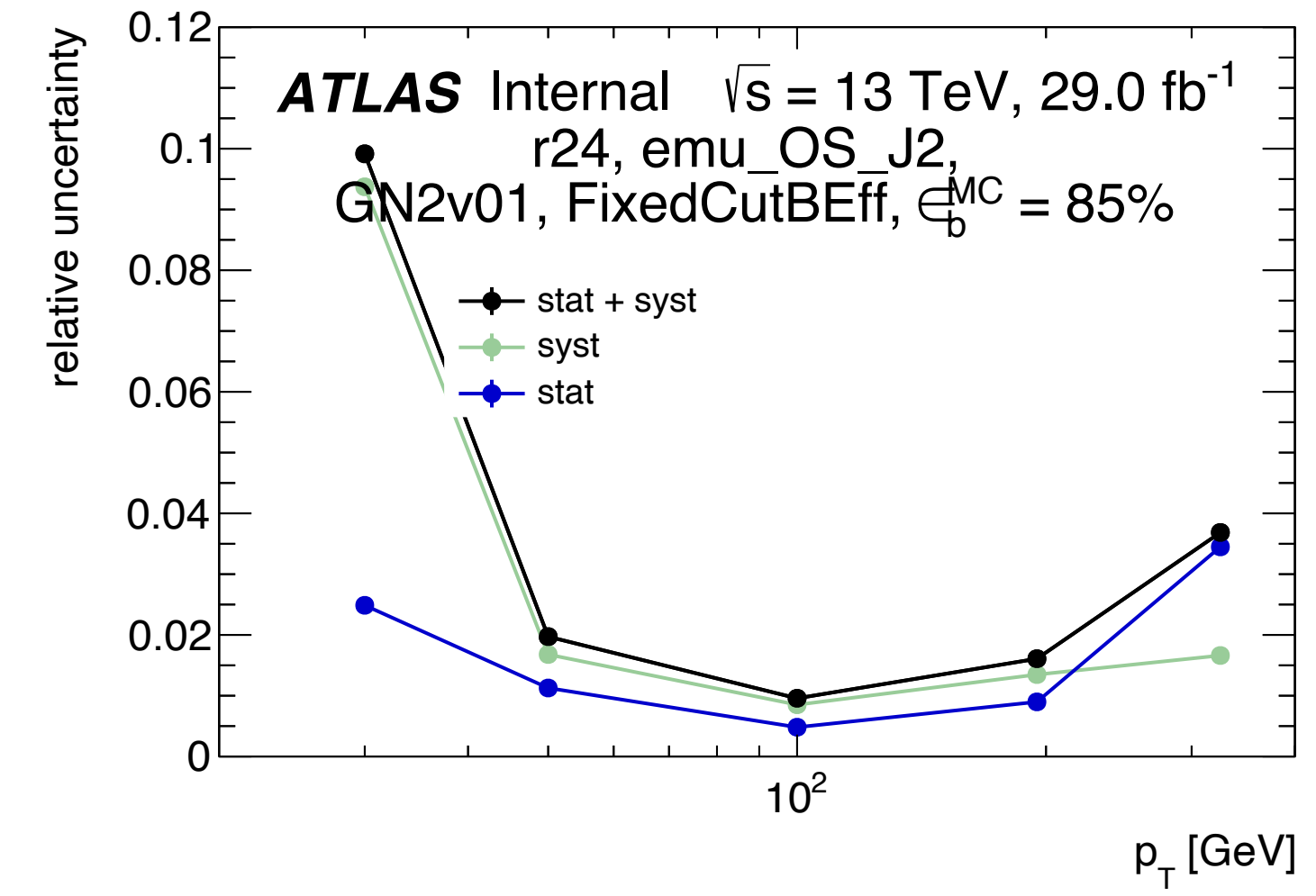
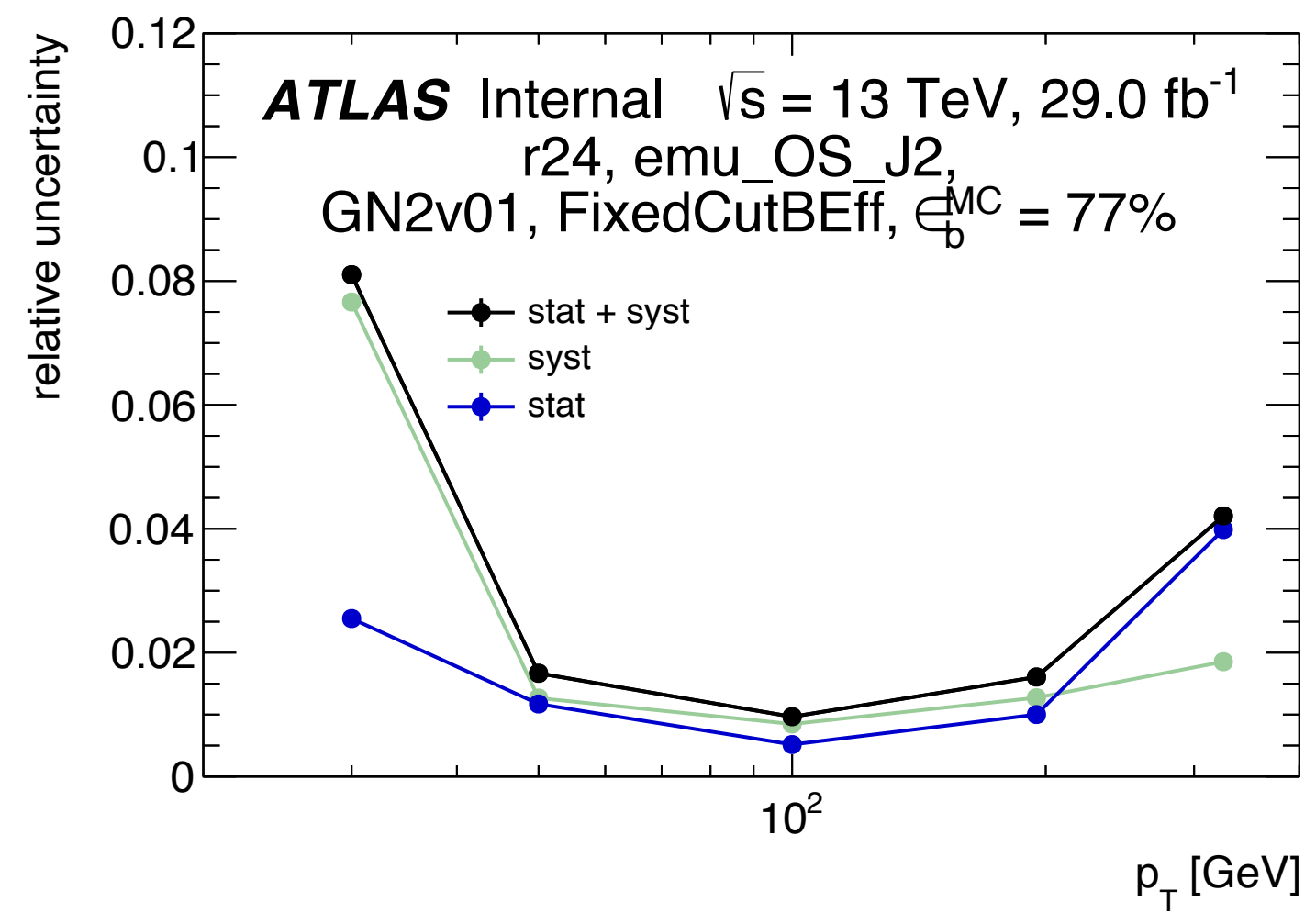
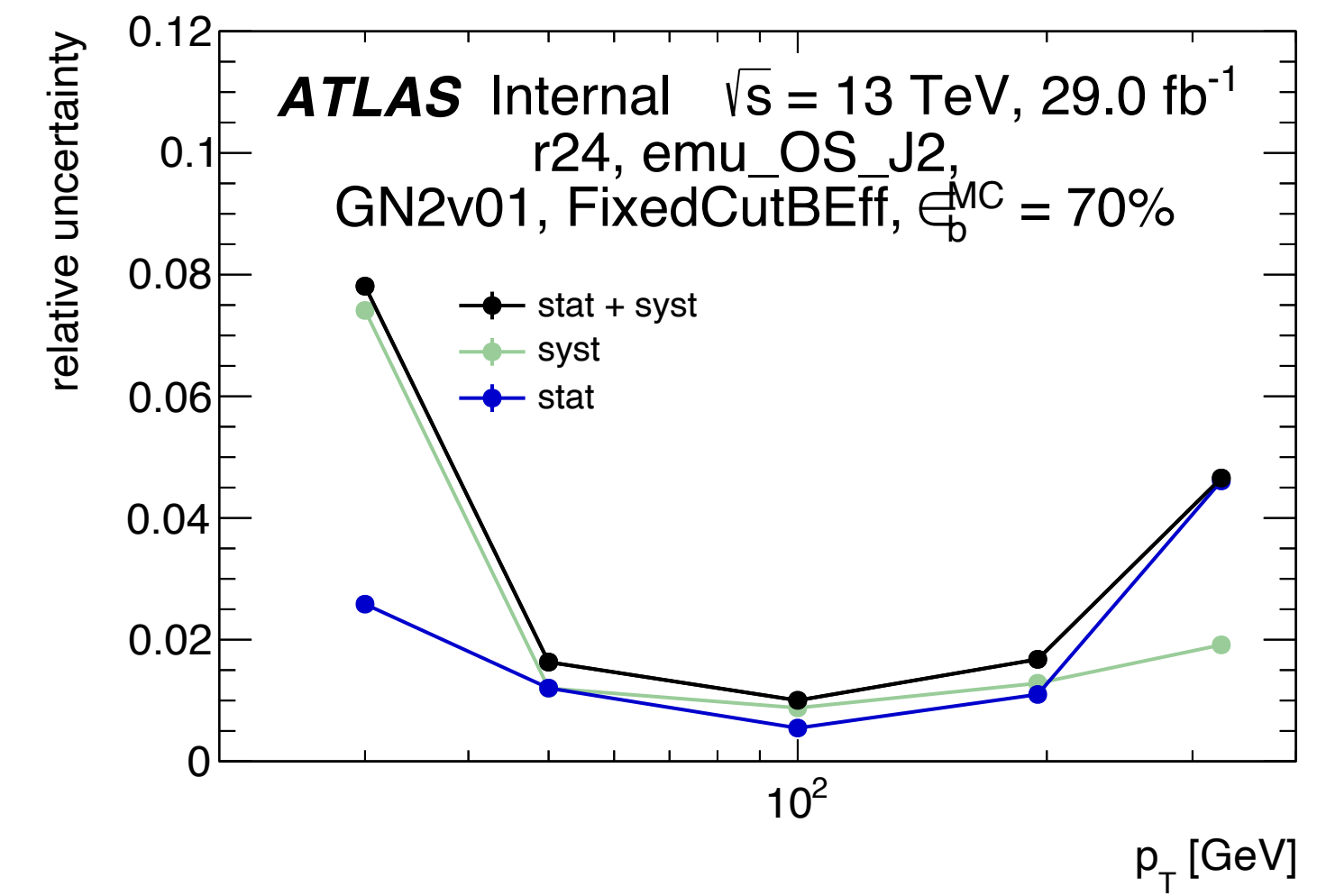
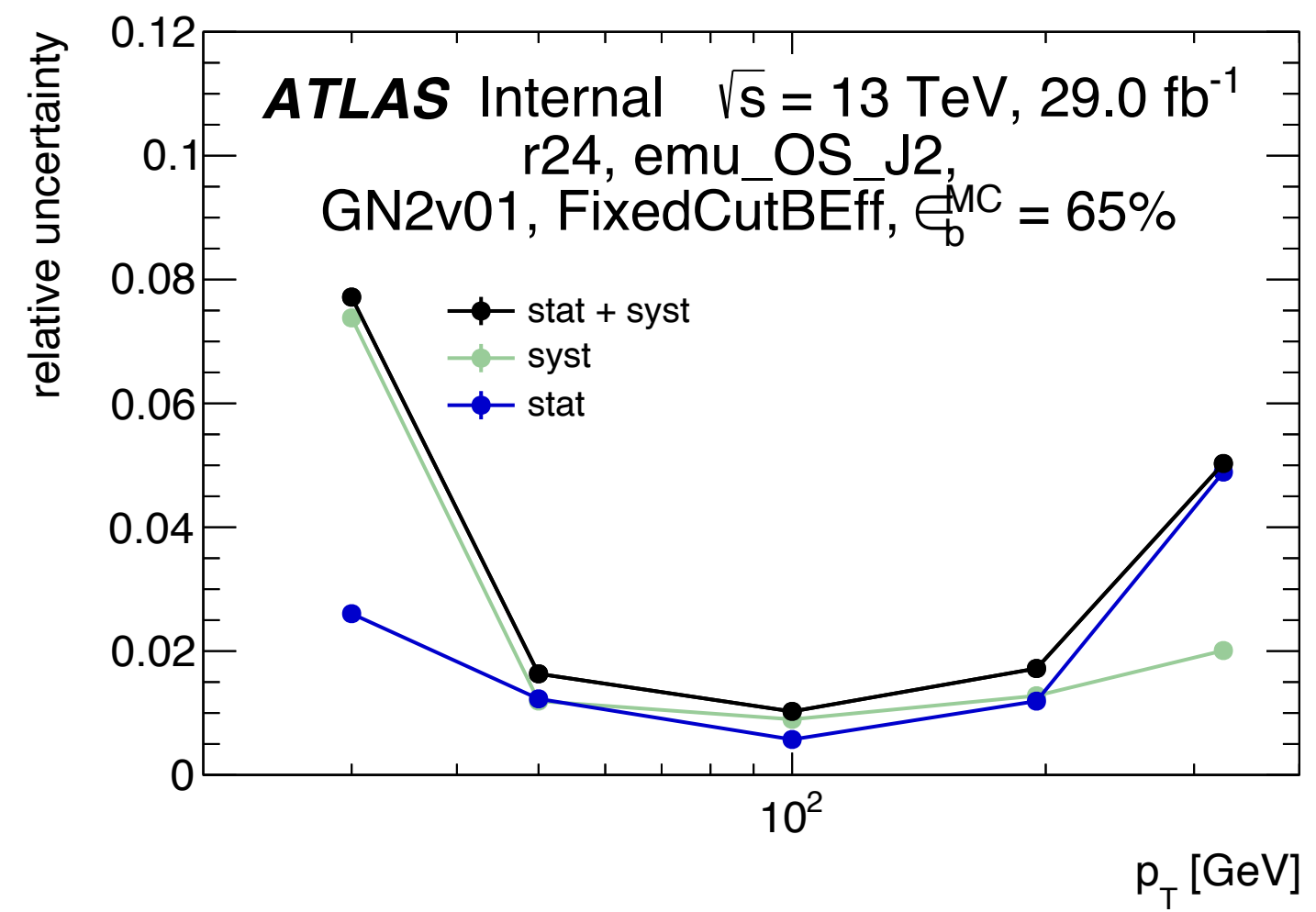
# $b$ -tagging efficiency measurement: Run 3

$b$ -tagging SFs as a function of the jet  $p_T$ , for the 65, 70, 77, 85, and 90% eff. WP of GN2



# $b$ -tagging efficiency measurement: Run 3

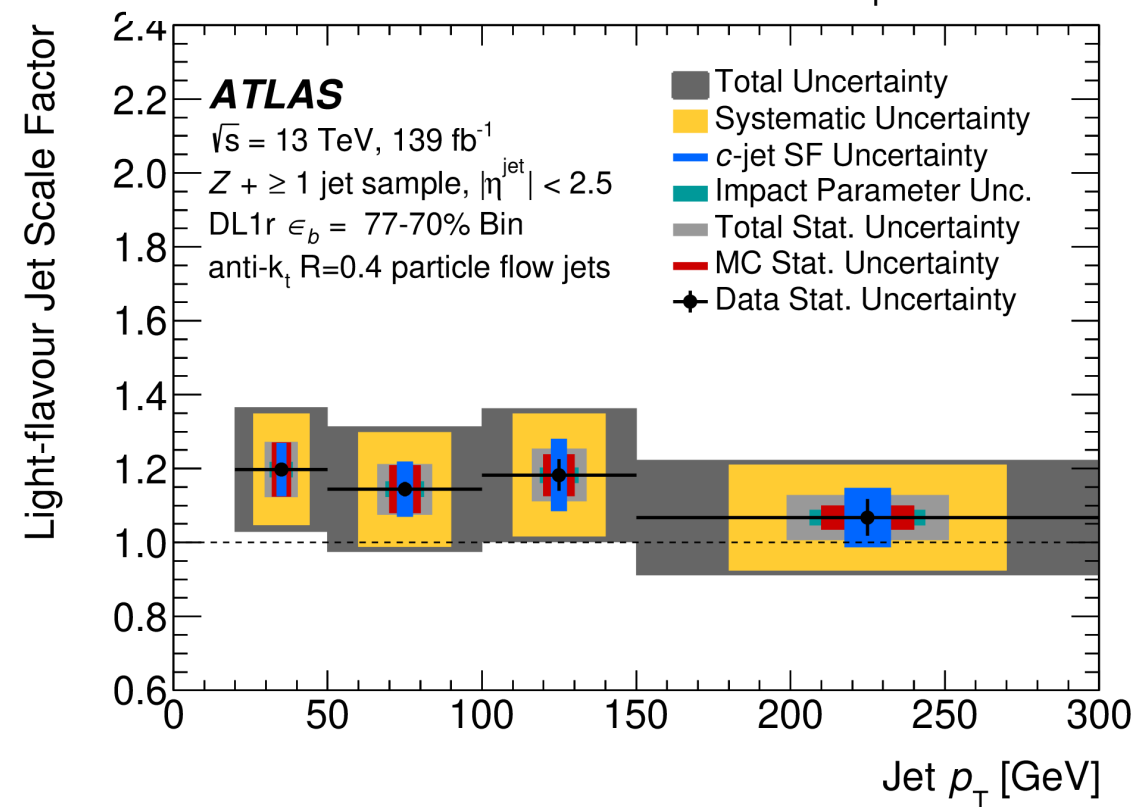
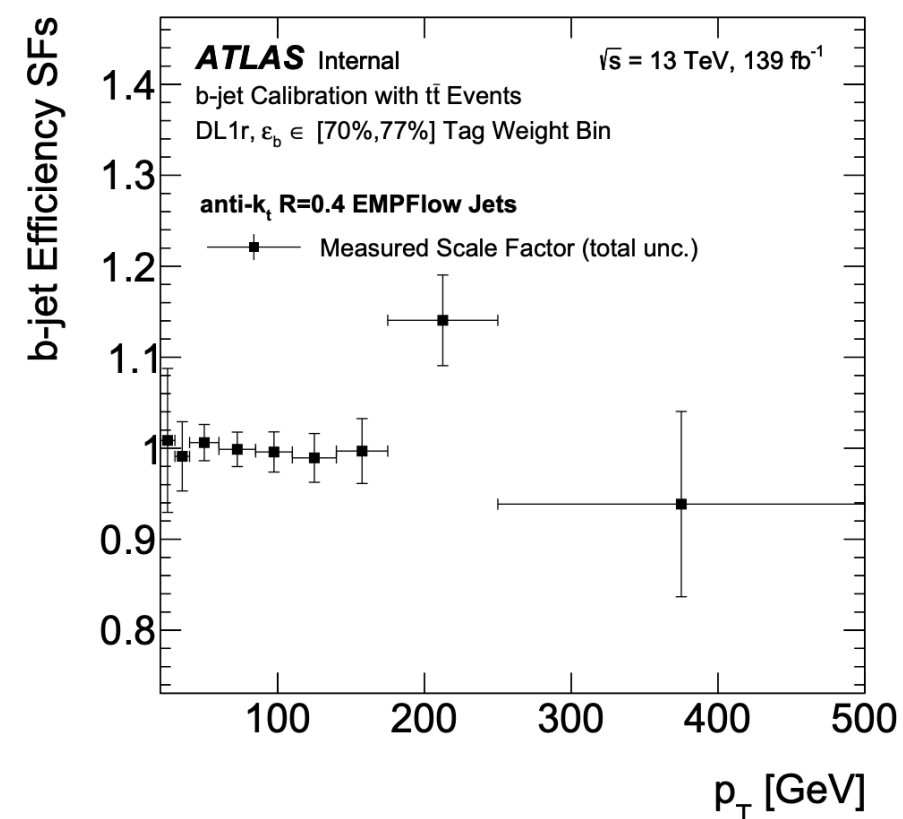
Relative uncertainty to the  $b$ -tagging SFs as a function of the jet  $p_T$ , for the 65, 70, 77, 85, and 90% eff. WP of GN2



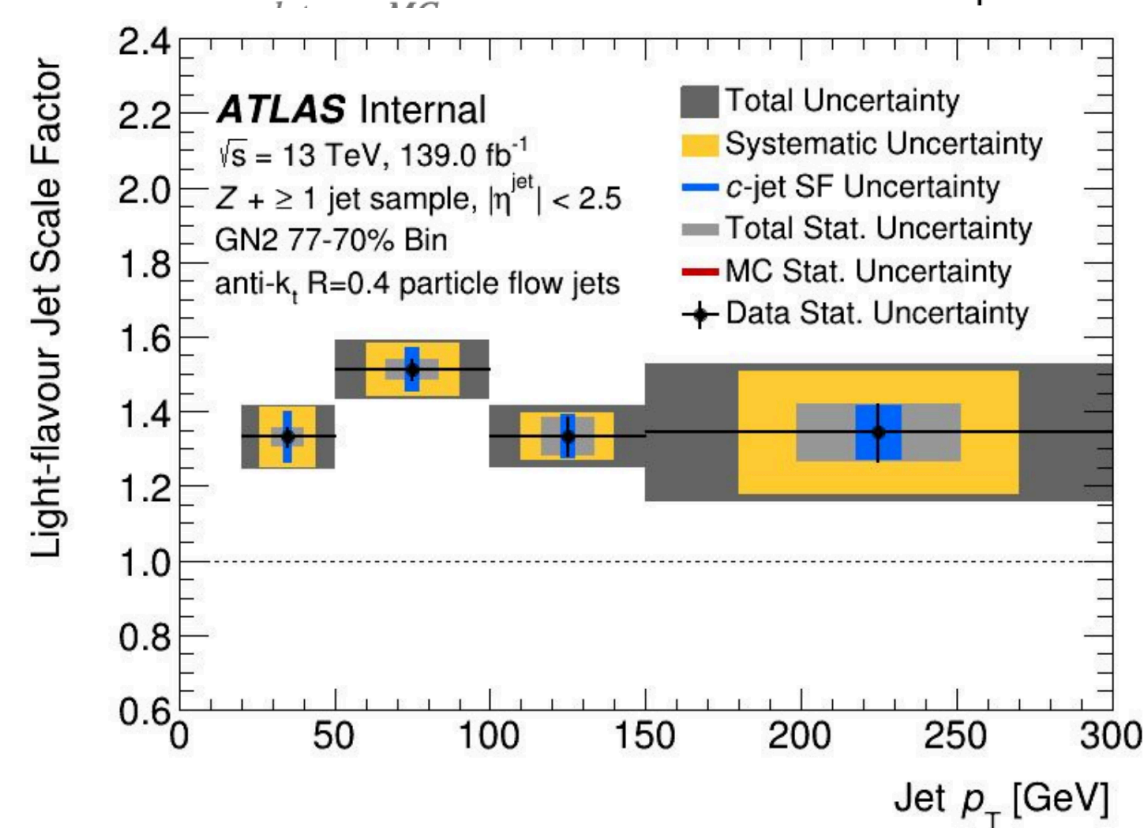
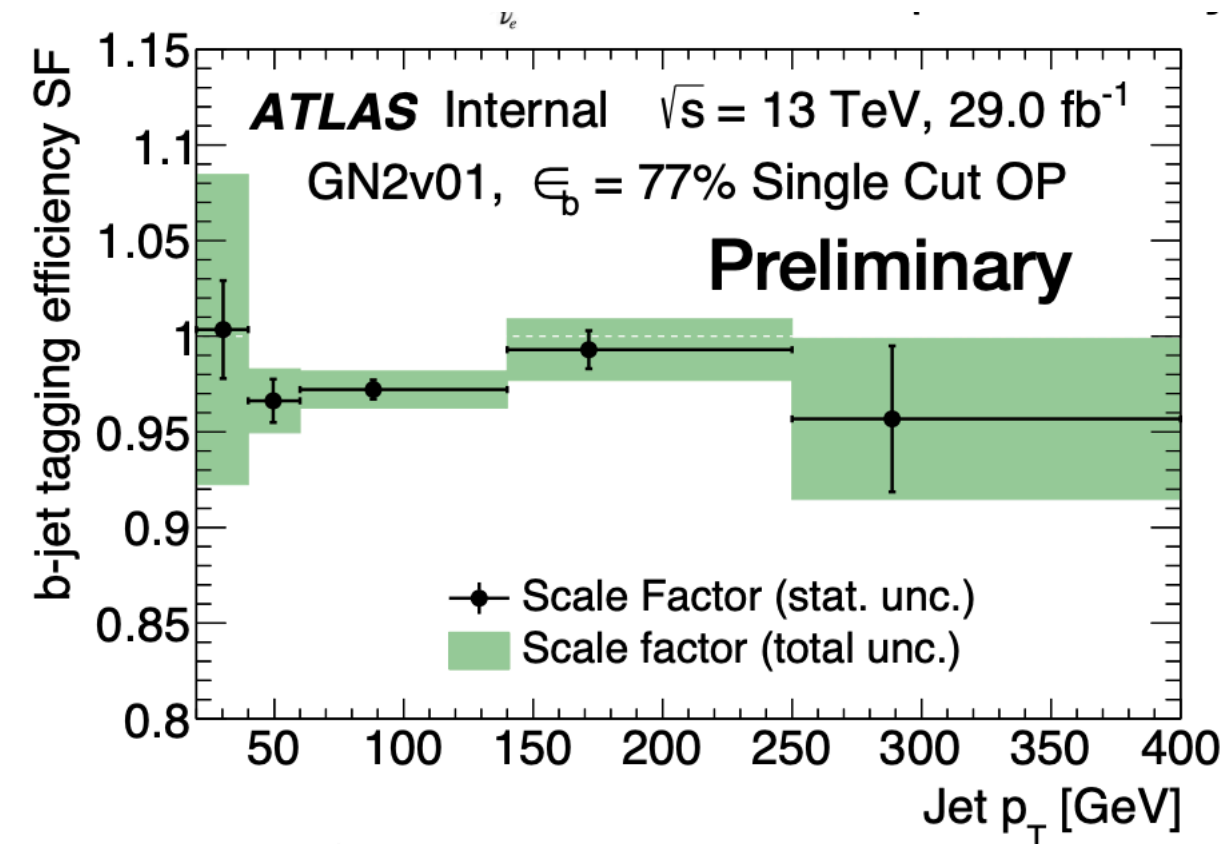
# $b$ -, $c$ -, and light jet scale factors for DL1r and GN2

- In order to estimate the ratio between the expected yields from the DL1r and GN2 77% eff. WPs, we relied on the  $c$ - and light-jet rejection curves (as a function of the  $b$ -jet efficiency) evaluated using MC samples.
- Trusting the ratios that we used to rescale the expected yields is a reasonable assumption, as long as the MC  $\rightarrow$  Data efficiency scale factors for  $b$ -,  $c$ -, and light jets are similar, between DL1r and GN2!

DL1r (Rel. 21)



GN2 (Rel. 25)



○ MC  $\rightarrow$  Data efficiency scale factors for  $b$ -jets are similar, between DL1r and GN2!

➡ Very close to 1 in both cases!

○ For light jets, the scale factors are close to 1.3 for GN2, while they are closer to 1 in the DL1r case.

➡ The two sets of Scale Factors differ of a factor between 10 and 20%, but they are compatible within their systematic uncertainties!

# True $bb$ / All purities in bkg. samples for the $HH \rightarrow b\bar{b}\gamma\gamma$ analysis

## True $bb$ / All purities (with DL1r-based b-tagging)

	High Mass 1	High Mass 2	High Mass 3	Low Mass 1	Low Mass 2	Low Mass 3	Low Mass 4	$b\bar{b}\gamma\gamma$ preselection
<b>Samples</b>								
<b>ggH</b>	0.843846	0.854195	0.929171	0.661551	0.848940	0.925240	0.845122	0.762660
<b>ttH</b>	0.741365	0.733432	0.734020	0.781049	0.805715	0.817540	0.805060	0.829875
<b>ZH</b>	0.954499	0.956188	0.968055	0.938304	0.951592	0.952504	0.962574	0.915358
<b>Other H</b>	0.757329	0.705959	0.736176	0.742982	0.734976	0.862540	0.957203	0.736247
<b><math>\gamma\gamma</math>+jets</b>	0.732071	0.801465	0.854433	0.625333	0.710732	0.754355	0.711985	0.543612

- The background samples for the  $HH \rightarrow b\bar{b}\gamma\gamma$  analysis include both true  $b$ -jets and  $c$  and light jets mistagged as  $b$ -jets.
- Given the 77% efficiency WP, the true  $bb$  purity in each sample strongly depends from the analysis category (= the pseudo-continuous DL1r score is used as input for the category definition).
  - ➔ **Resonant** background
    - ➔ - Between 60% and 95%, depending on the category and the production mode.
    - ZH has the highest purity (above 90% in each category!).
  - **Continuum** background
    - ➔ Between 60% and 85%, depending on the category.

# True $bb$ / All purities in bkg. samples for the $HH \rightarrow b\bar{b}\gamma\gamma$ analysis

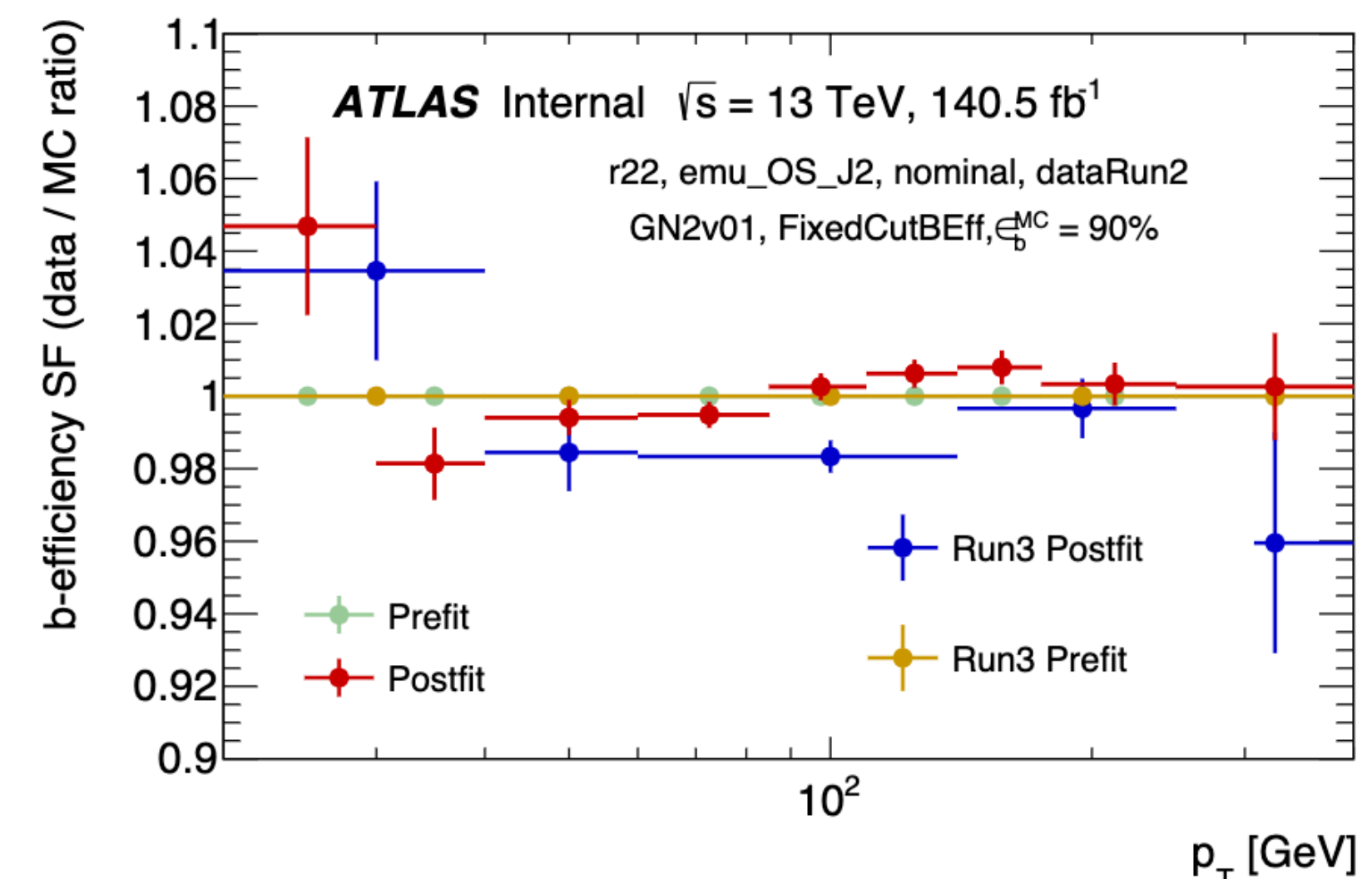
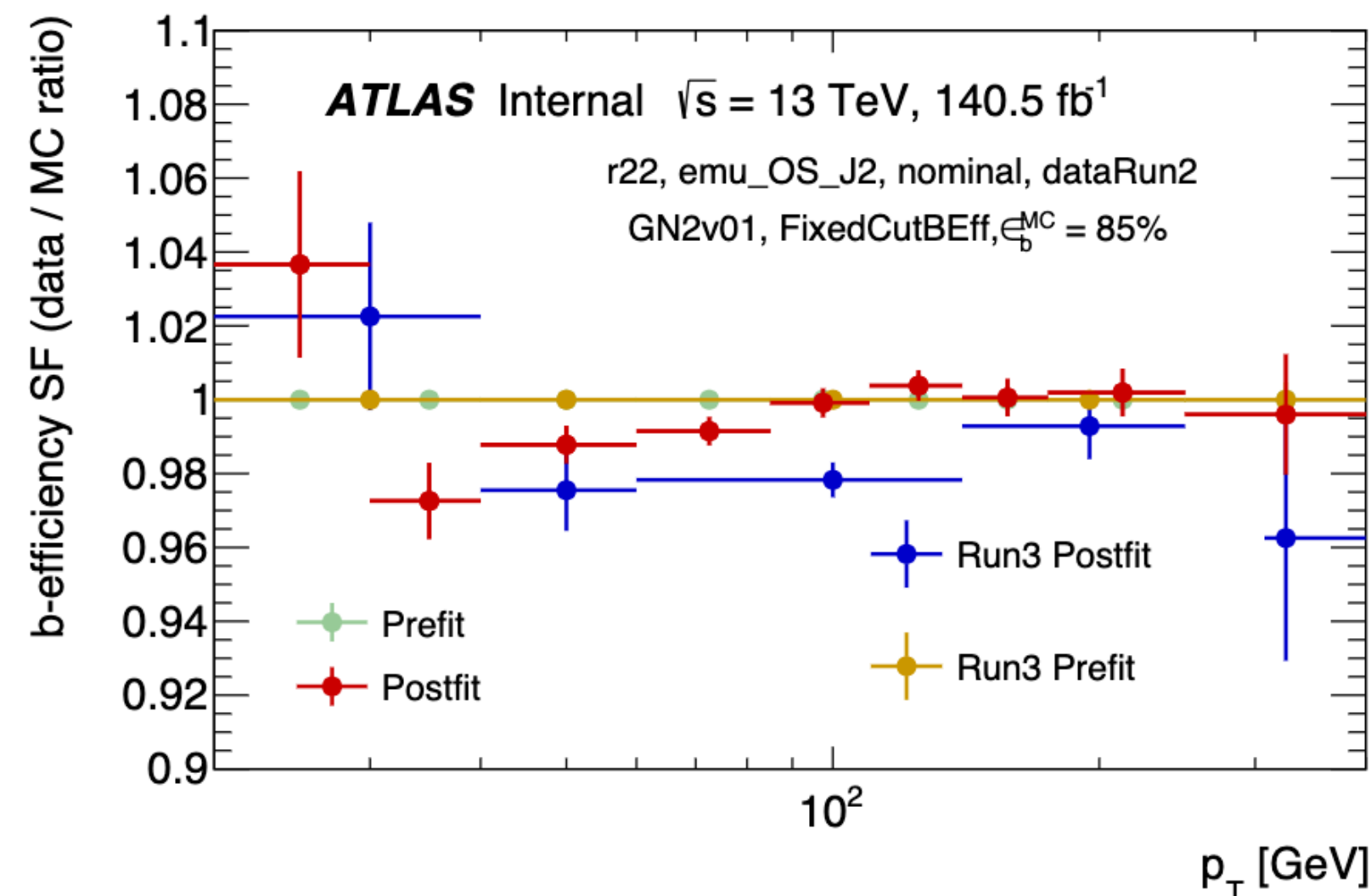
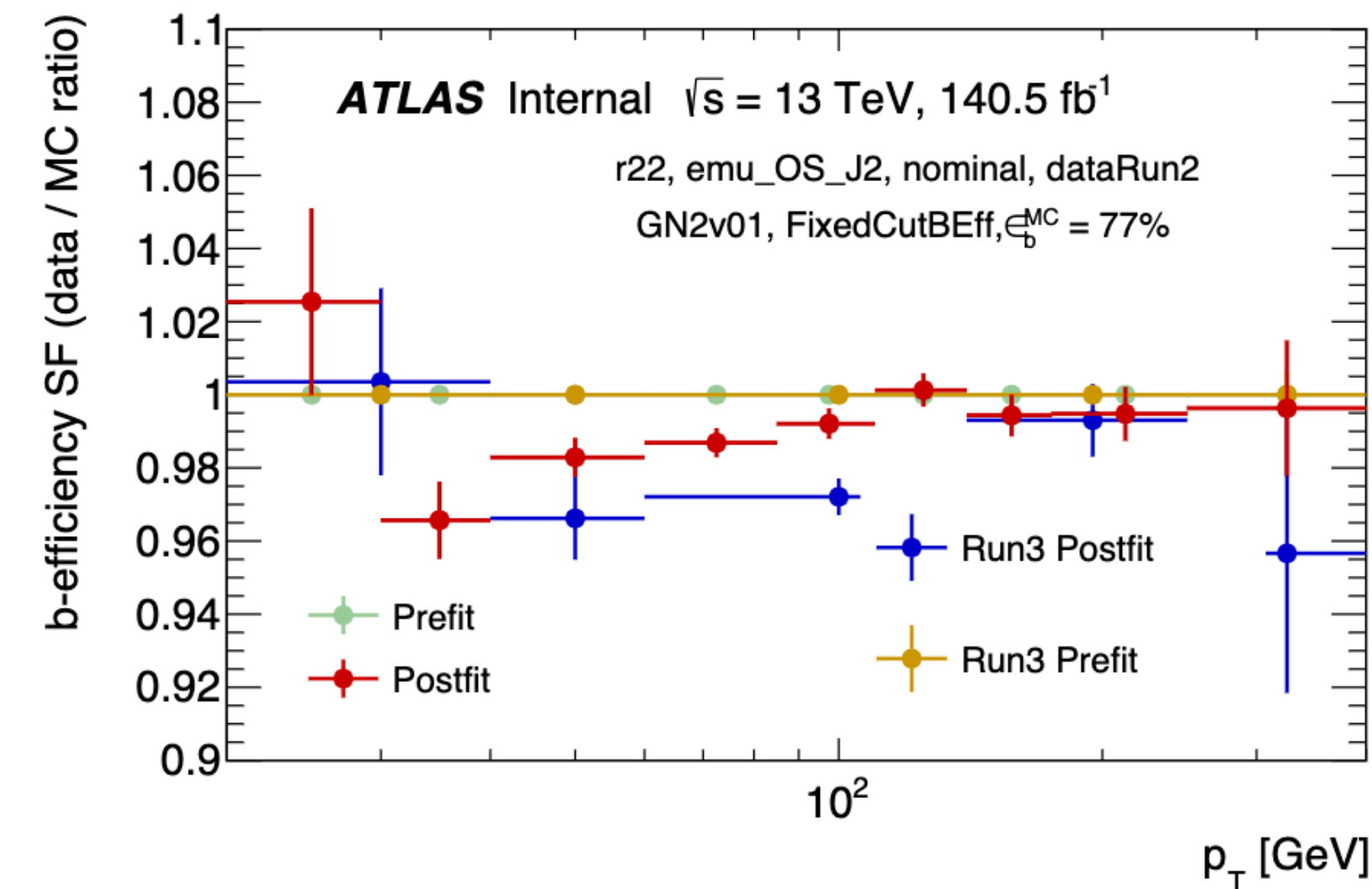
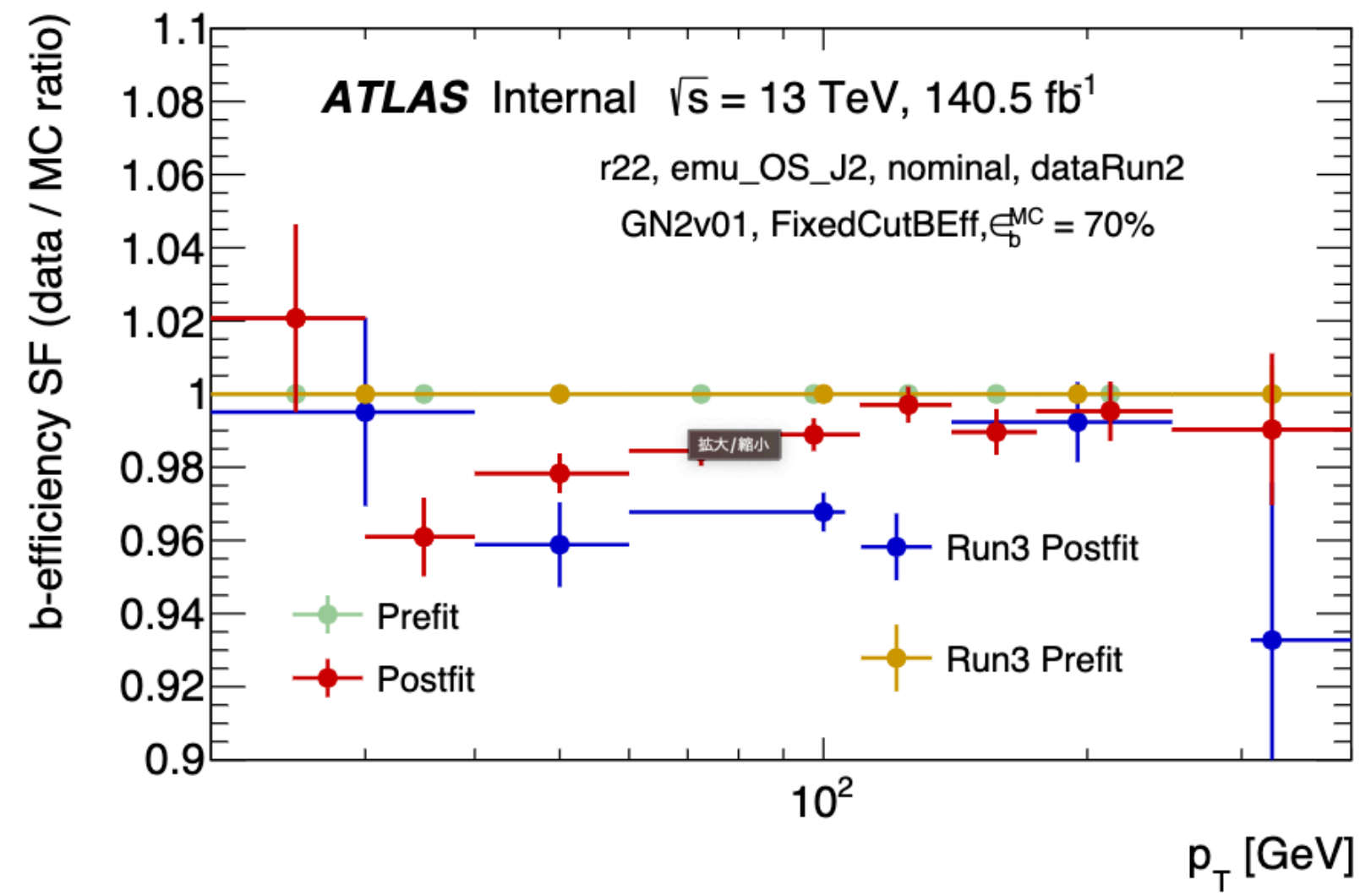
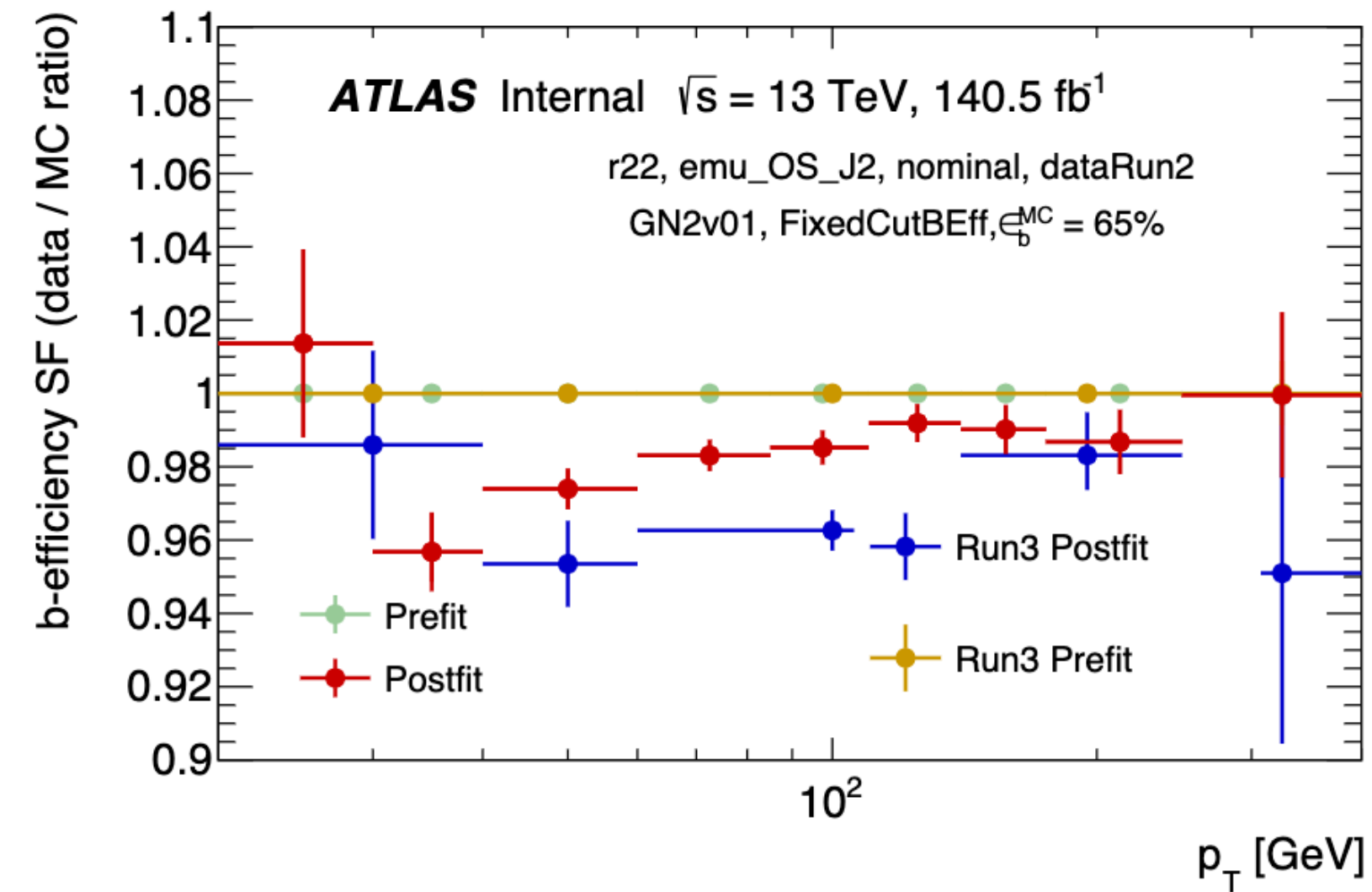
## True $bb$ / All purities (with GN2-based b-tagging)

	High Mass 1	High Mass 2	High Mass 3	Low Mass 1	Low Mass 2	Low Mass 3	Low Mass 4	$b\bar{b}\gamma\gamma$ preselection
<b>Samples</b>								
<b>ggH</b>	0.938394	0.934139	0.968502	0.862267	0.933399	0.965587	0.935997	0.909779
<b>ttH</b>	0.854443	0.848802	0.850285	0.880112	0.894690	0.901759	0.893961	0.909822
<b>ZH</b>	0.985365	0.985805	0.989611	0.977814	0.981146	0.981960	0.983784	0.970225
<b>Other H</b>	0.877708	0.844318	0.859252	0.870091	0.862325	0.939799	0.985878	0.876607
<b><math>\gamma\gamma</math>+jets</b>	0.878519	0.909757	0.941216	0.827112	0.874222	0.895774	0.870666	0.788487

- After replacing the bkg. efficiency for DL1r 77% WP with the estimated bkg. efficiency for the GN2 WP, we could improve the overall True  $bb$  purities in each analysis category, and for each sample!
- The margin of improvement depends from both the analysis category and the particular bkg. process.
  - ➔ **Resonant** background
    - ➔ - Improvement within 10% and ~ few %.
    - The new purity is always above 85% for each analysis category.
  - **Continuum** background
    - ➔ - Improvement between 10% and 20%.
    - The new purity is always above 80%!

# $b$ -tagging efficiency measurement: Run 2 VS Run 3

- We compared the nominal  $b$ -tagging efficiency SFs (no. systematic uncertainties yet), between full Run 2 and Run 3 (2022 data and mc23a only).
- For Run 3, SFs seems to be systematically larger w.r.t. Run 2 (except for low  $p_T$ ).
- Similar trend as a function of  $p_T$  w.r.t. observed in the past, for DL1d (see backup).





# $b$ -tagging efficiency measurement: DL1d tagger, Run 2

