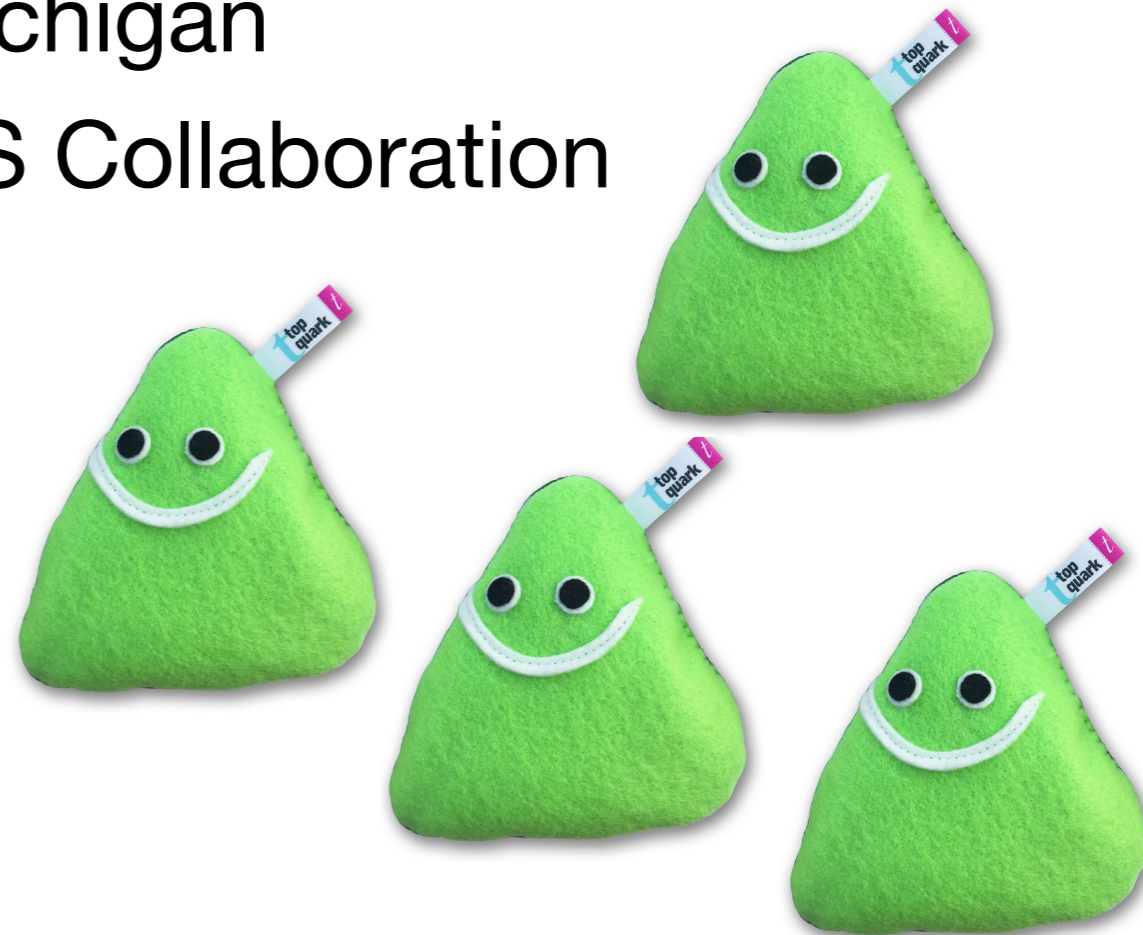


Search for  $t\bar{t}H/A \rightarrow t\bar{t}t\bar{t}$  in pp collisions at  $\sqrt{s} = 13$  TeV with the ATLAS detector

Meng-Ju Tsai

University of Michigan

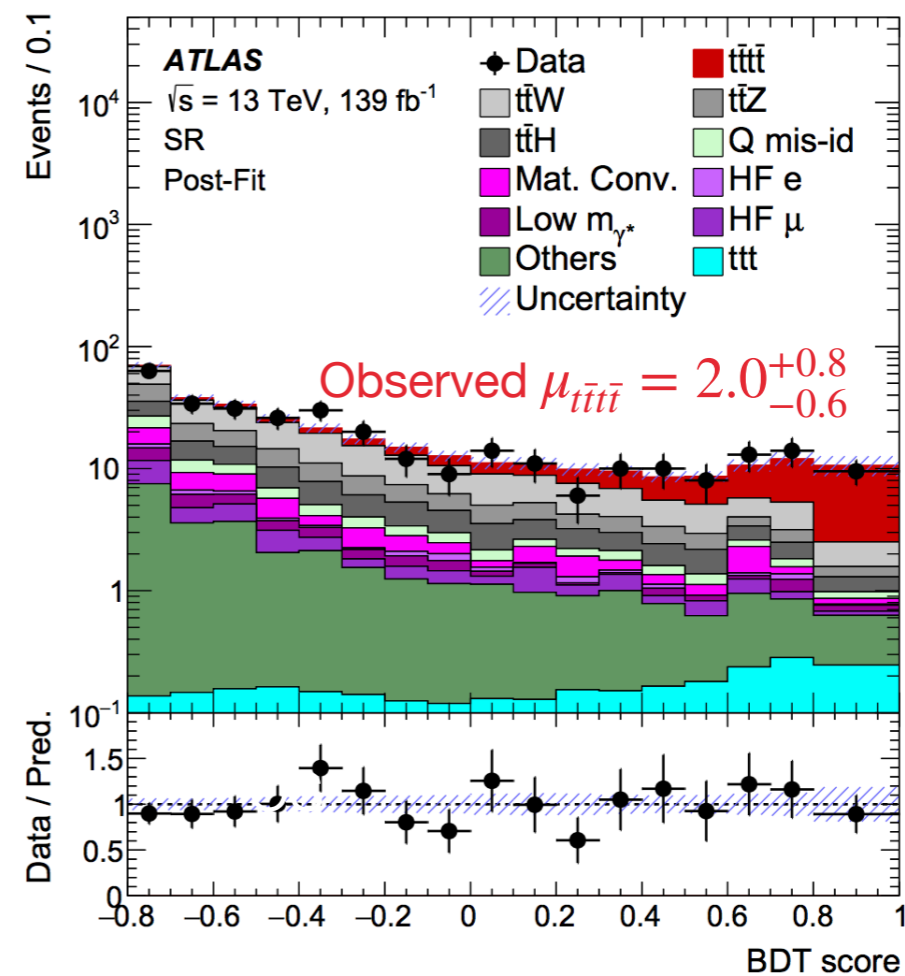
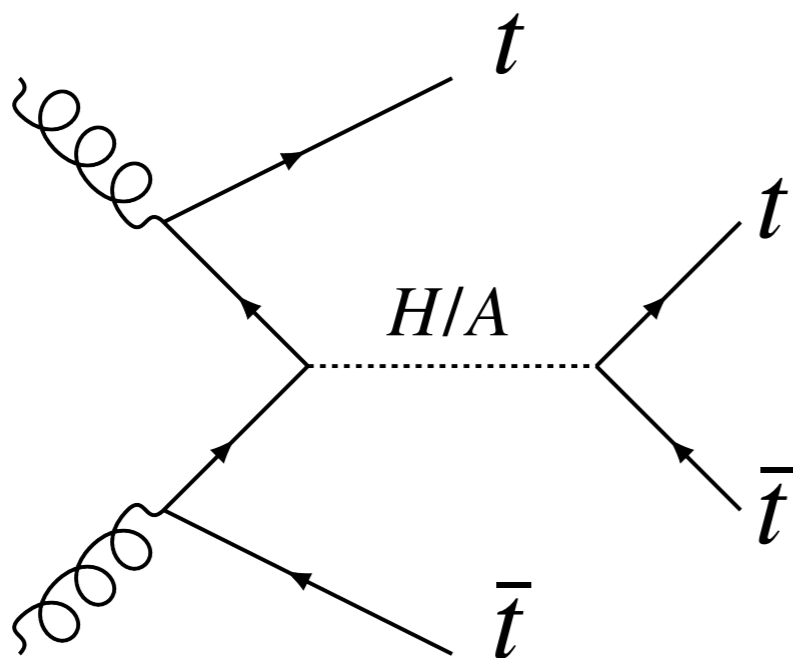
On behalf of the ATLAS Collaboration



# Motivation

- Evidence for SM  $t\bar{t}t\bar{t}$  production in the same-sign dilepton and multilepton (SSML) final states was achieved  $\rightarrow$  observed cross section:  $\sigma_{t\bar{t}t\bar{t}} = 24_{-6}^{+7}$  fb (twice of expectation!)
- This analysis searches for the **heavy Higgs boson in top-associated production mode ( $t\bar{t}H/A \rightarrow t\bar{t}t\bar{t}$ ) from the 2HDM model** in the same SSML final state
- Sensitive to the low  $\tan\beta$  region with small interference with SM background
- $A \rightarrow t\bar{t}$  decay mode dominates at the low  $\tan\beta$  region (so does  $H \rightarrow t\bar{t}$  in the alignment limit)

High jet, b-jet multiplicities and High  $H_T$

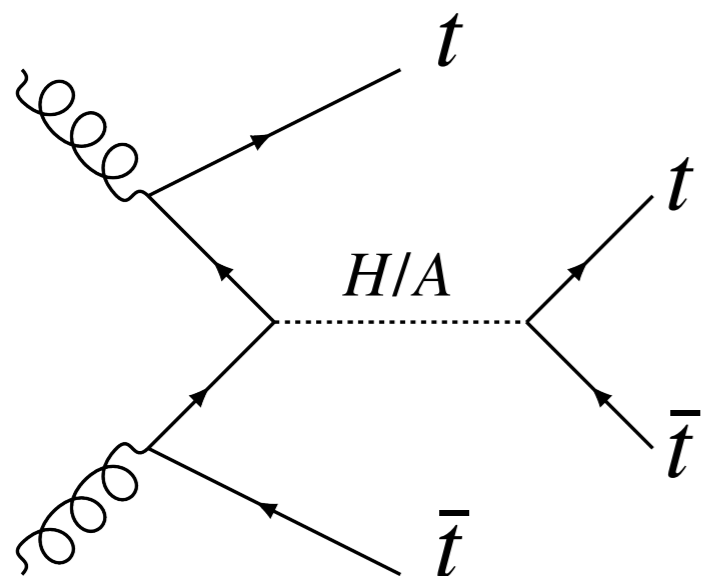


arXiv: 2007.14858

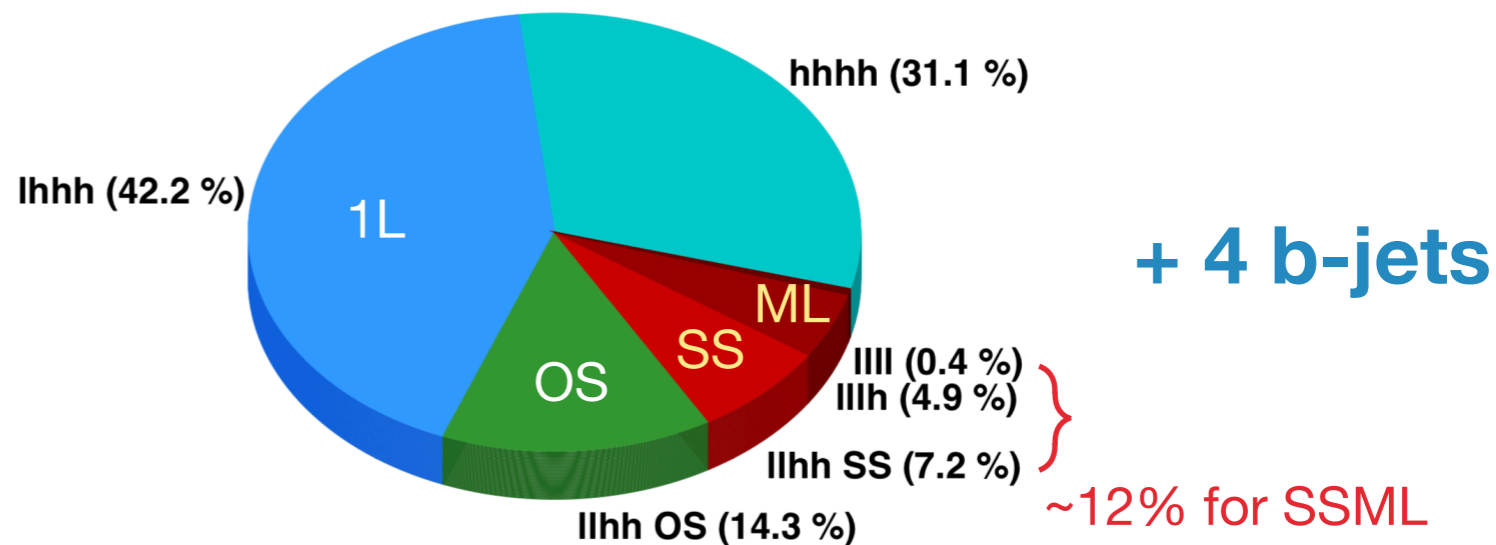
# Analysis overview

- Analysis strategy based on the SM  $t\bar{t}t\bar{t}$  analysis ([paper](#)) with additional upgrades
- Improved jet reconstruction (particle flow algorithm) and neural net based b-tagging algorithm
- Multivariate analysis (sequential BDTs) to extract BSM signals
- Main challenges
  - Small signal cross sections and background modeling in extreme kinematic regions
  - Hard to extract BSM  $t\bar{t}t\bar{t}$  from highly irreducible  $t\bar{t}V$  and SM  $t\bar{t}t\bar{t}$

High jet, b-jet multiplicities and High  $H_T$



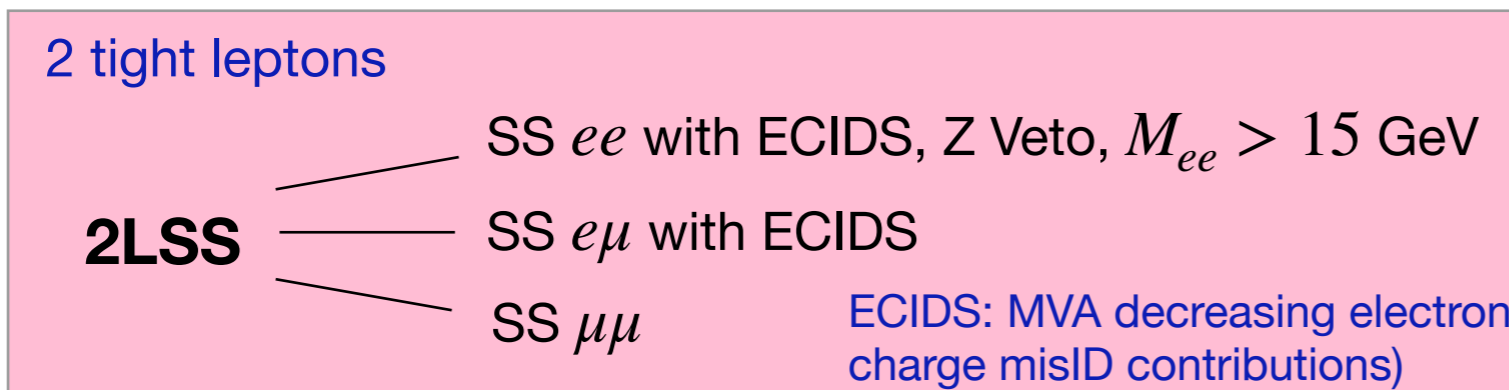
Decay product of 4 W bosons



# Analysis strategy

## Preselection

Both SS and ML,  $\geq 1$  b-tagged jet at 77% WP



## ✓ Signal background separation

- Events with  $N_{jets} \geq 6$ ,  $N_{b-jets} \geq 2$  and  $H_T = \sum p_T^\ell + \sum p_T^j \geq 500$  GeV are signal-enriched phase space (baseline SR) in which is used for **SM BDT** and **BSM parametrized-BDT (pBDT)** training with **XGBoost**

## ✓ Statistical analysis

- In data-blinded stage, estimate the normalization factor for backgrounds with unblinded CRs in the template fit
- Simultaneous fit of CRs and SRs with post-fit background model to get limits on the cross section for the BSM signals

# Backgrounds in SSML channel

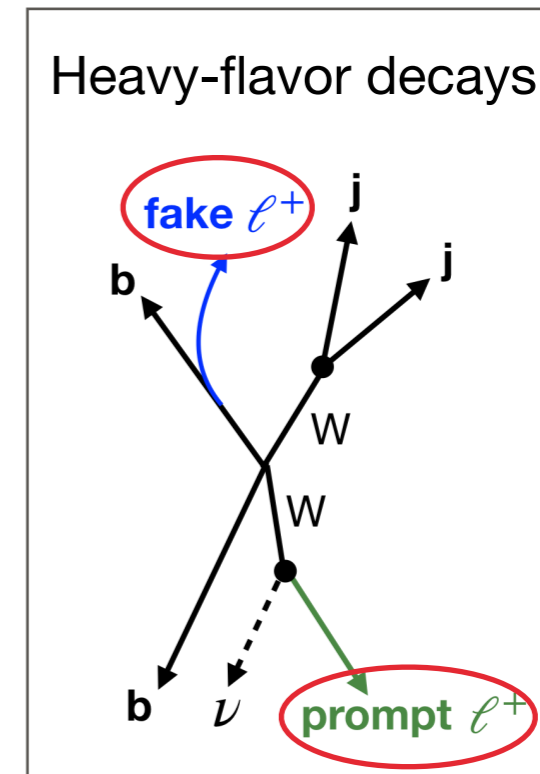
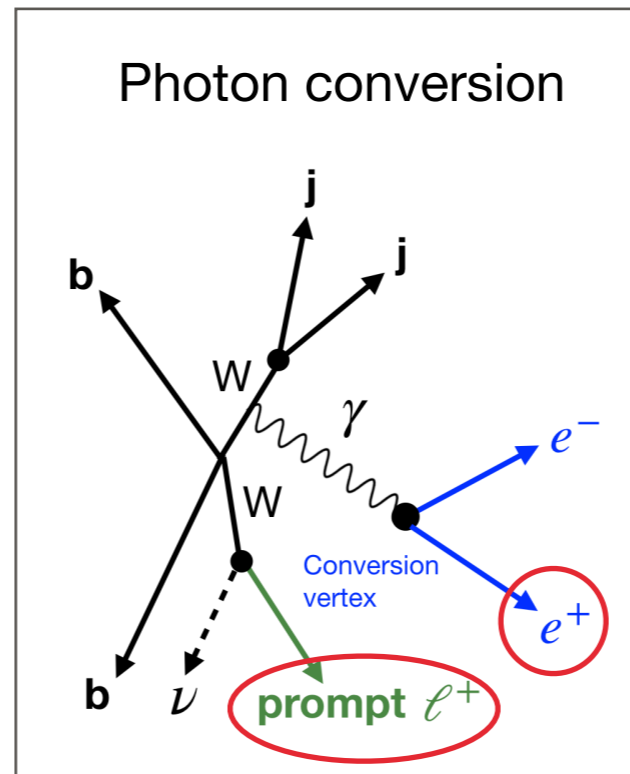
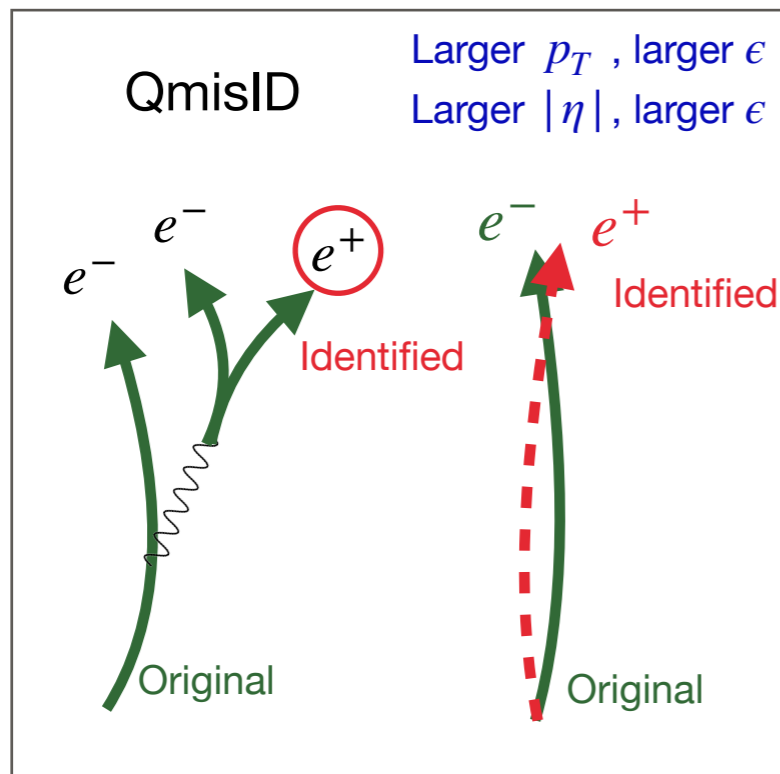
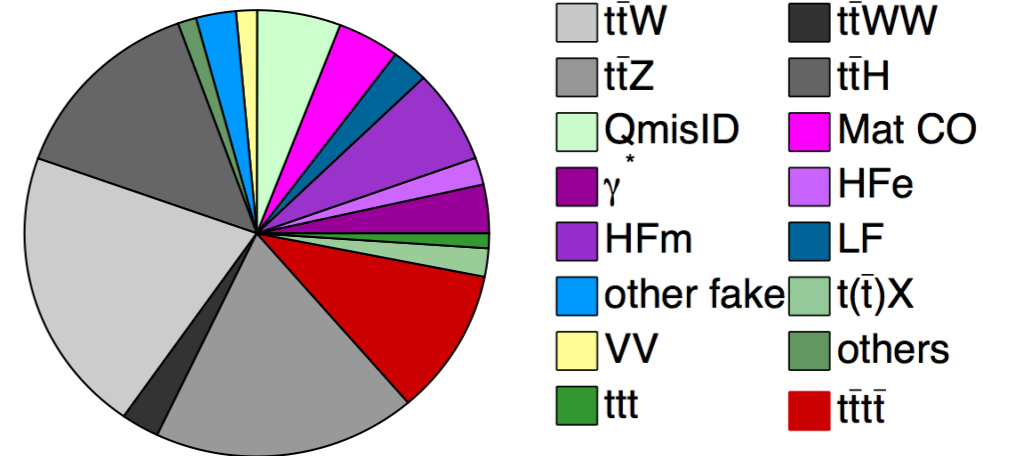
- **Physics processes: (~75%)**

- SM  $t\bar{t}t\bar{t}$  (**constrained to SM with 20% theory uncert.**)
- $t\bar{t}W$  NLO QCD (**MC corrected to data**)
- $t\bar{t}Z, t\bar{t}H$  and minor processes (**fully taken from MC**)

- **Instrumental and fake backgrounds: (~25%)**

- Charge mis-identification (**data-driven method**)
- Non-prompt leptons from heavy-flavor decays and photon conversion (**MC corrected to data in CRs**)
- Fake leptons from light mesons and quark/gluon jets, and minor backgrounds (**fully taken from MC**)

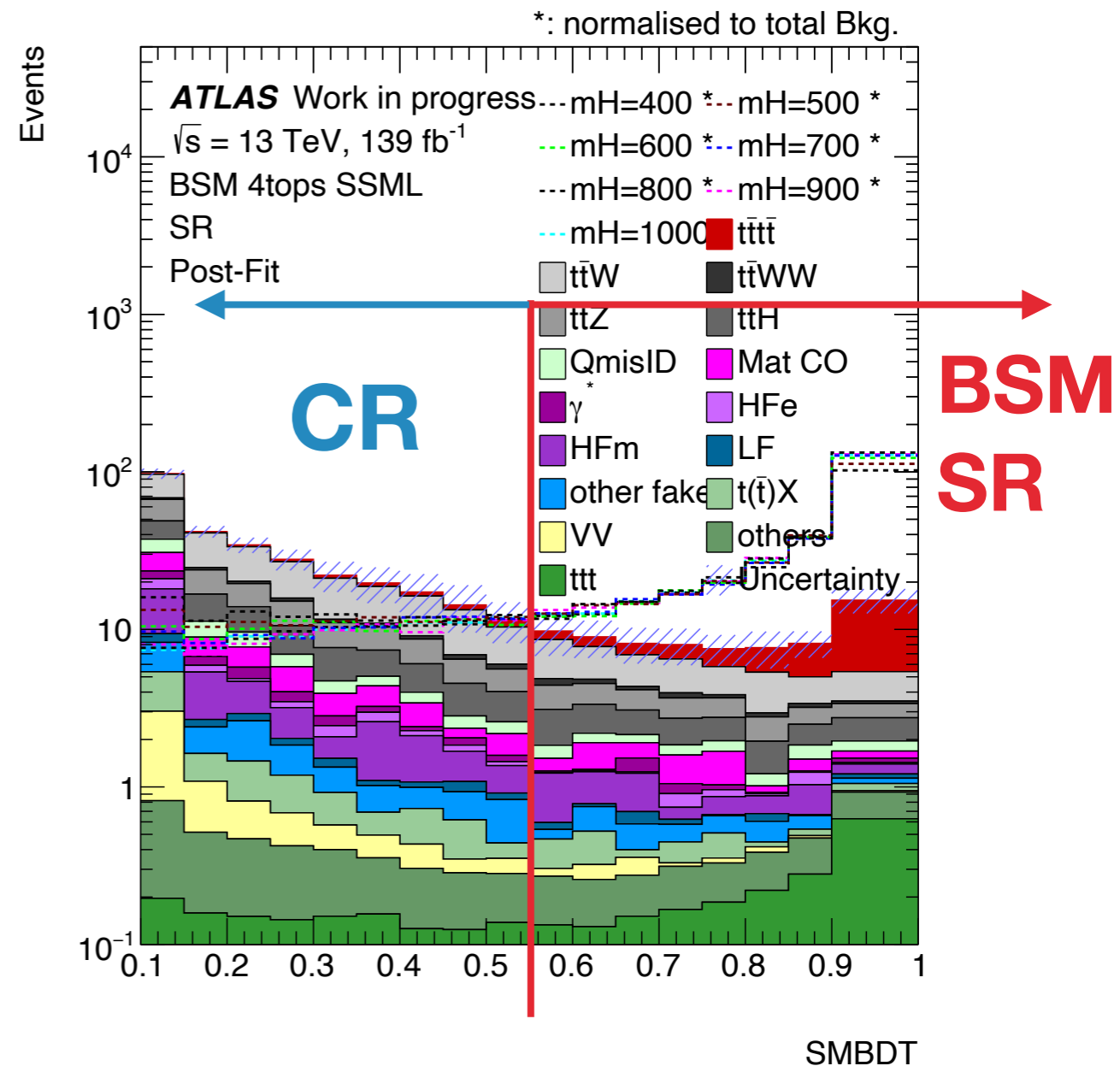
**Baseline SR (Pre-fit)**  $N_{jets} \geq 6 \ \& \ N_{b-jets} \geq 2 \ \& \ H_T \geq 500 \text{ GeV}$





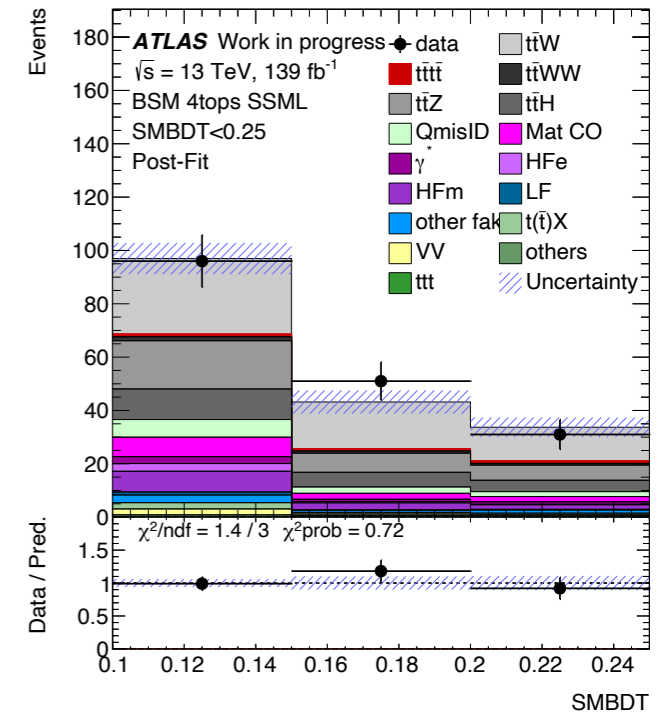
# Sequential BDT strategy - SM BDT

- **SM BDT** is trained with SM  $t\bar{t}t\bar{t}$  against SM backgrounds using **XGBoost** package
- Discriminate  $t\bar{t}t\bar{t}$ -like events from SM background
- Low SM BDT region for background modeling
- High SM BDT region for BSM signal extraction

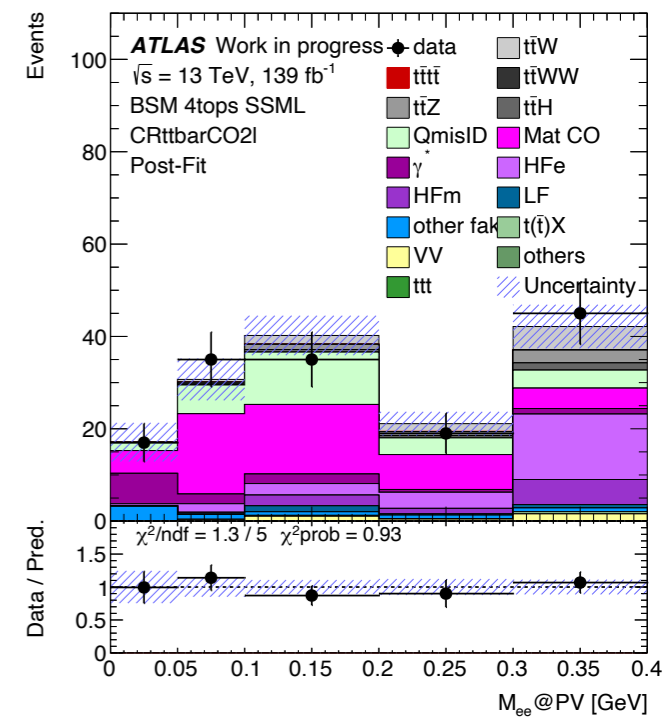
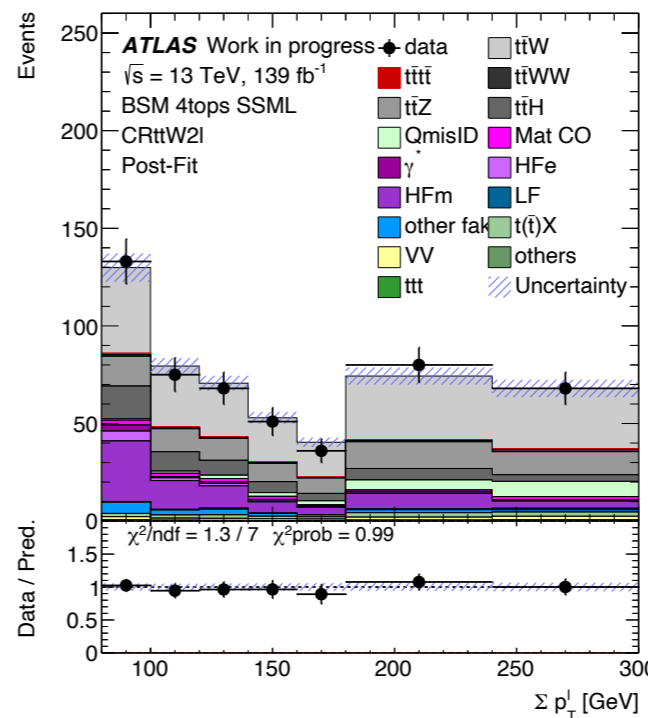
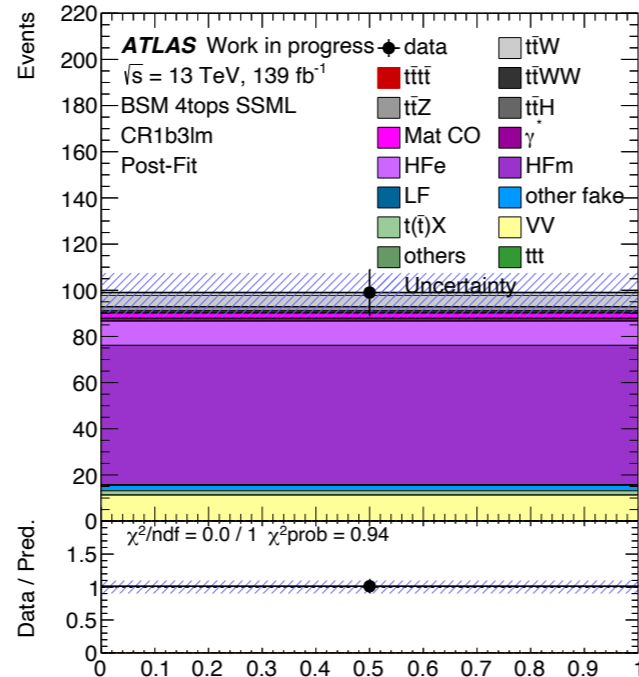
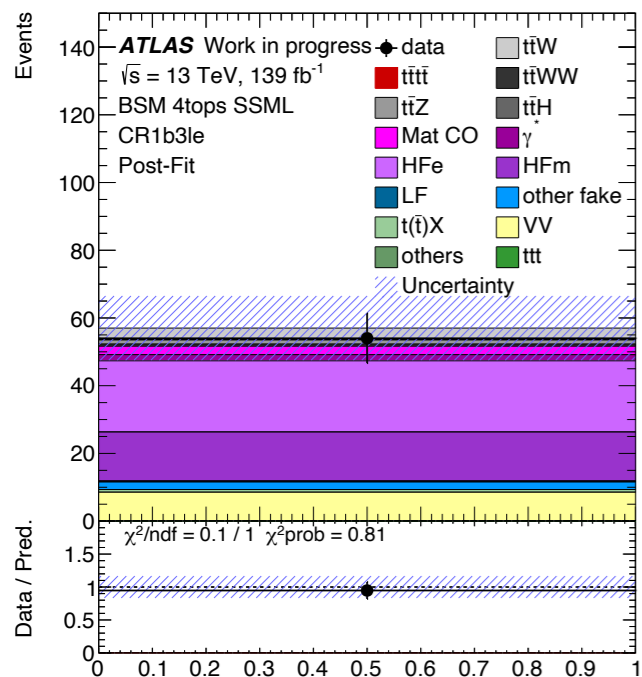


# Fake and $t\bar{t}W$ background estimation

- Template fit to data in 5 control regions to estimate the normalization factors of  $t\bar{t}W$  and fake backgrounds
- Low BDT control region was included to make the background modeling close to the unblinding fit setup  
 → unblind bins where signal contamination < 5% (**SM BDT < 0.25**) in the template fit

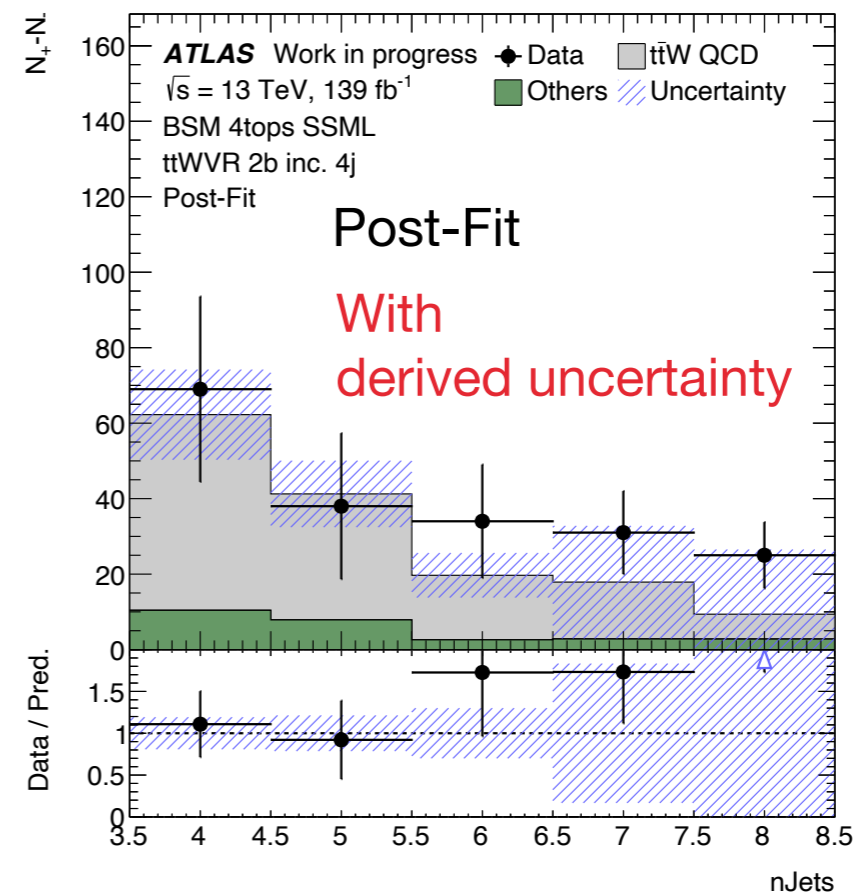
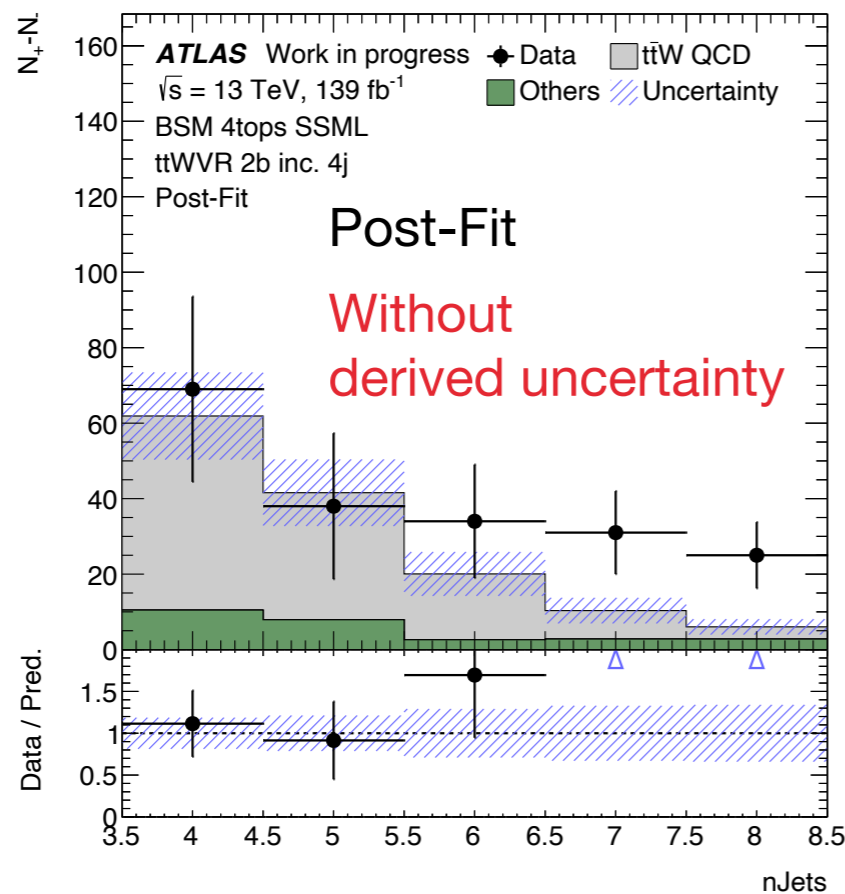


Parameter	NF $t\bar{t}W$	NF CO	NF $\gamma^*$	NF HF e	NF HF $\mu$
Value ( $\pm \text{Stat.} + \text{Syst.}$ )	$1.48^{+0.28}_{-0.27}$	$1.58^{+0.61}_{-0.53}$	$0.62^{+0.61}_{-0.42}$	$0.95^{+0.37}_{-0.35}$	$0.98^{+0.22}_{-0.20}$



# $t\bar{t}W$ validation region

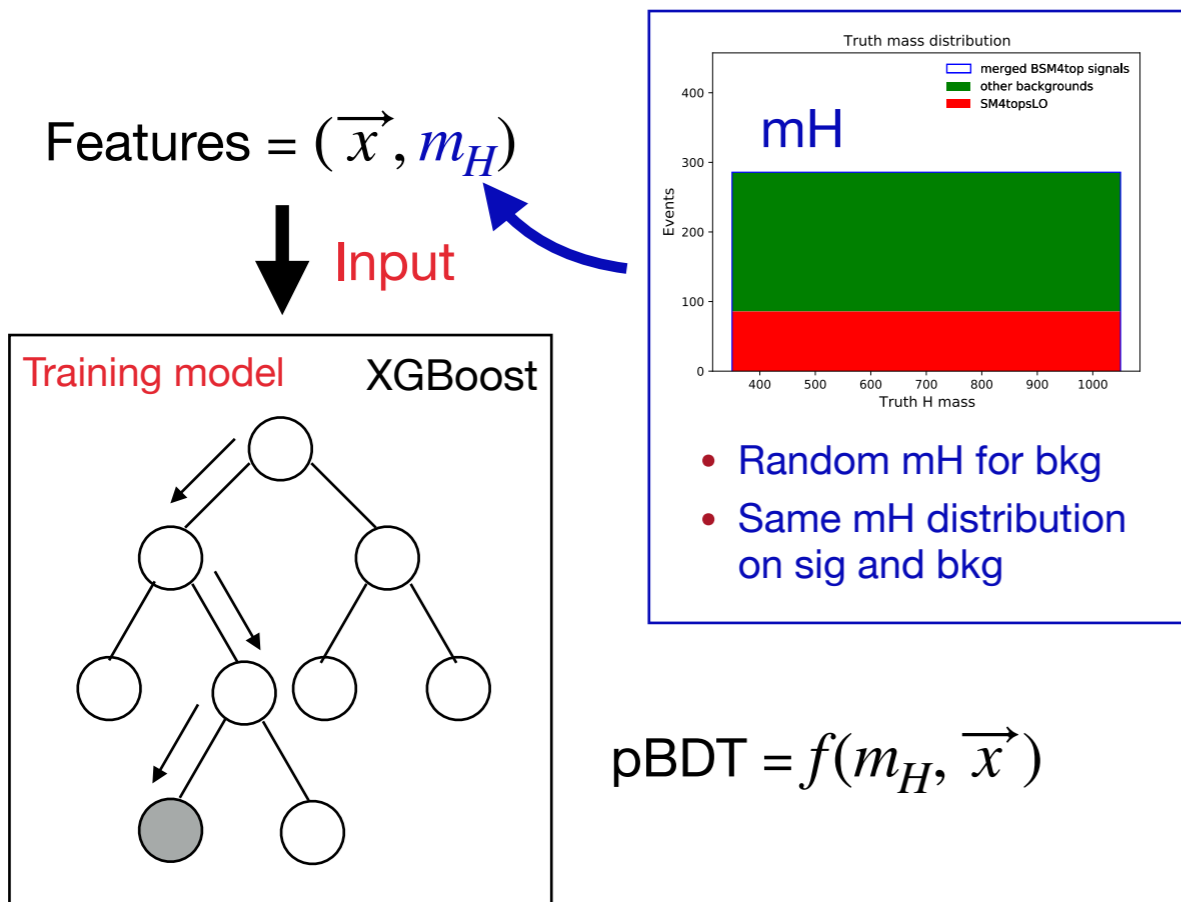
- $t\bar{t}W$  validation region: difference between positively and negatively charged leptons ( $N_{++} - N_{--}$ ) with selection  $N_{\text{jets}} \geq 4$  and  $N_{\text{b-jets}} \geq 2$  (no  $H_T$  selection)
- Large Data/MC discrepancy in the high  $N_{\text{jets}}$  region  $\rightarrow$  derive large uncertainty for  $t\bar{t}W$
- Working in progress to update the  $t\bar{t}W$  QCD+EW NLO samples





# Introduction to parameterized BDT

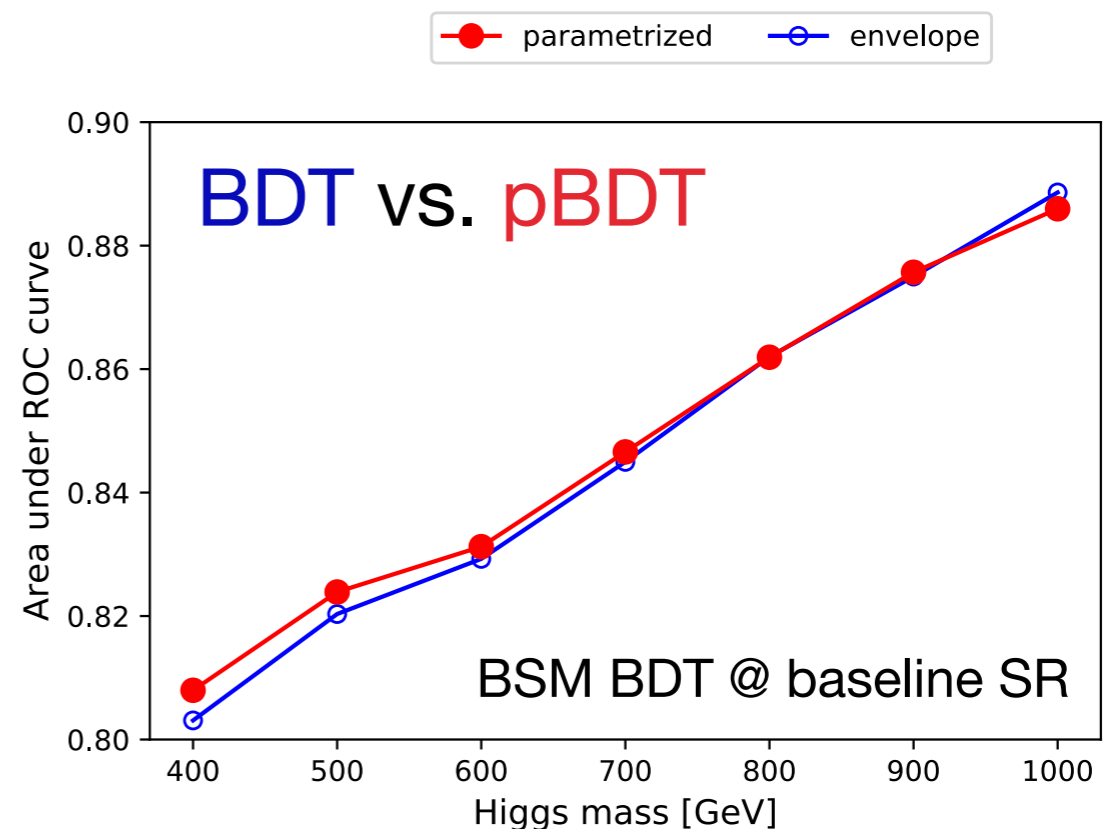
- Idea of pBDT is from parameterized neural network ([ArXiv:1601.07913](https://arxiv.org/abs/1601.07913))



$$pBDT = f(m_H, \vec{x})$$

BDT Application

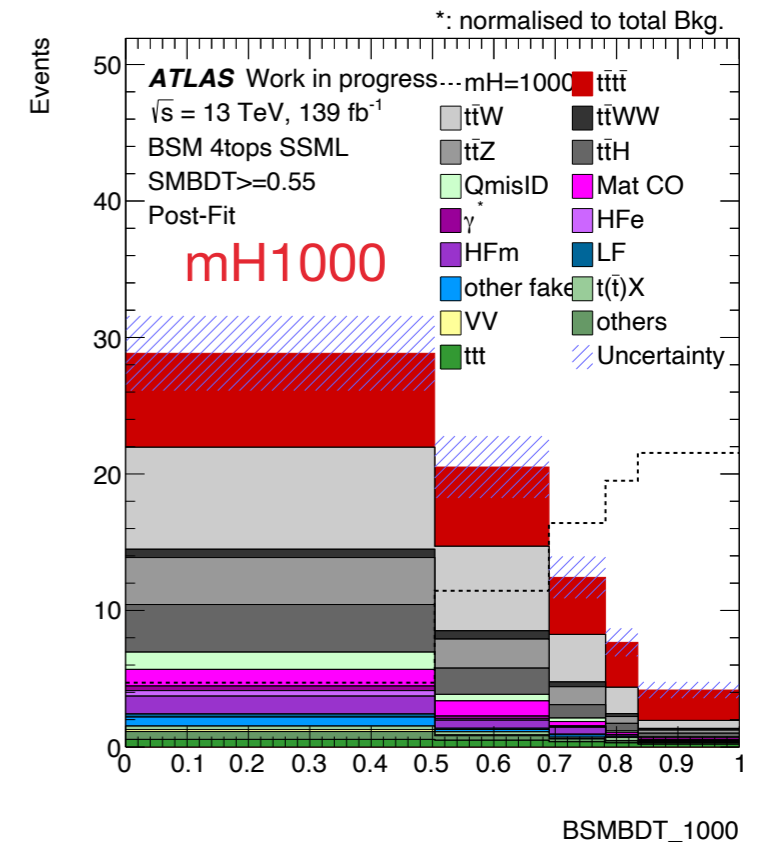
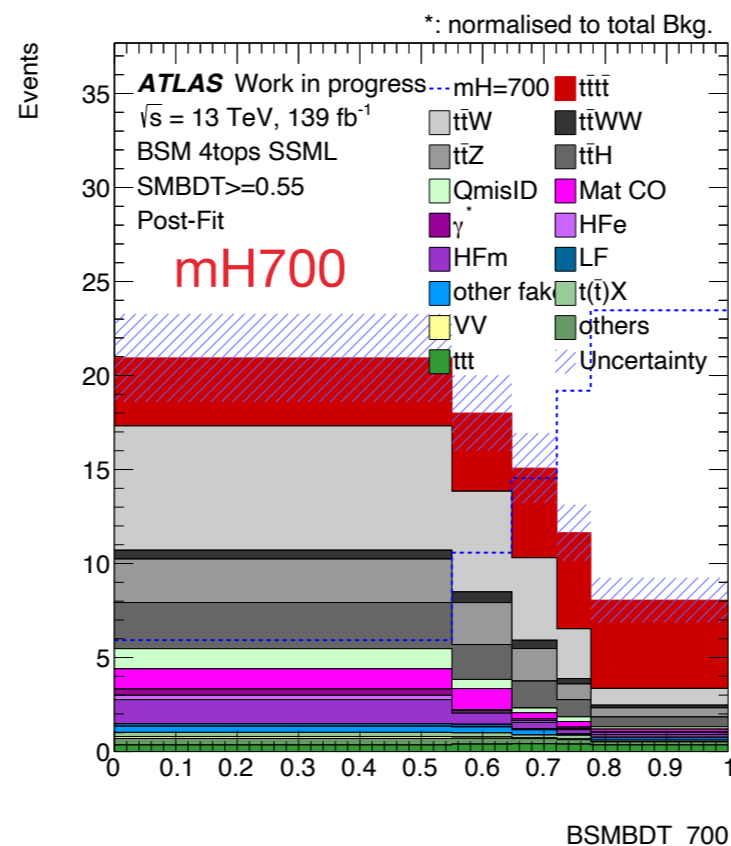
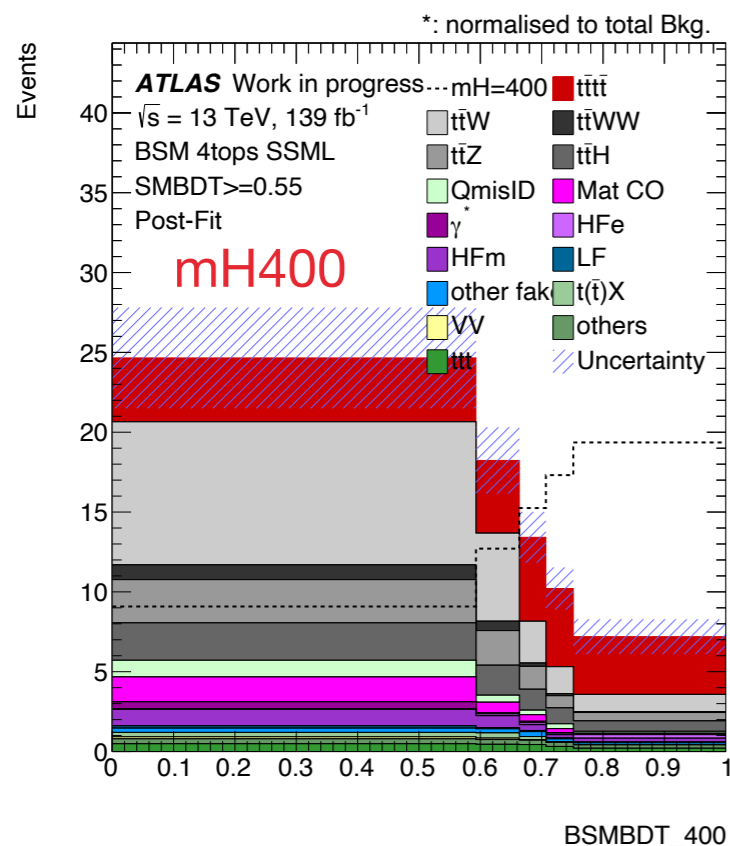
“BSM BDT @  $m_H=400$  GeV”  
 $= f(\vec{x}, m_H = 400 \text{ GeV})$



- BDT at each mass point:** each signal point against all SM backgrounds
- Parameterized BDT:** merged signals against all SM backgrounds

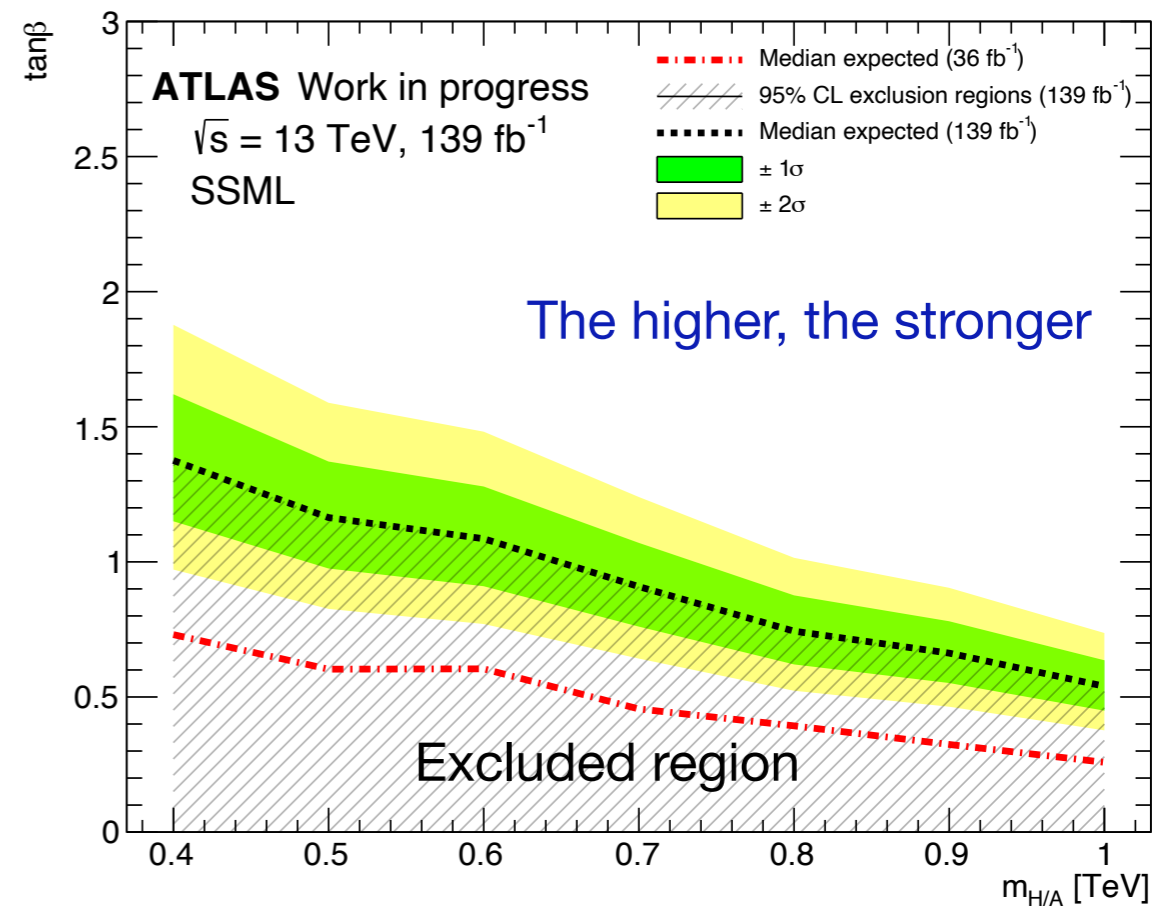
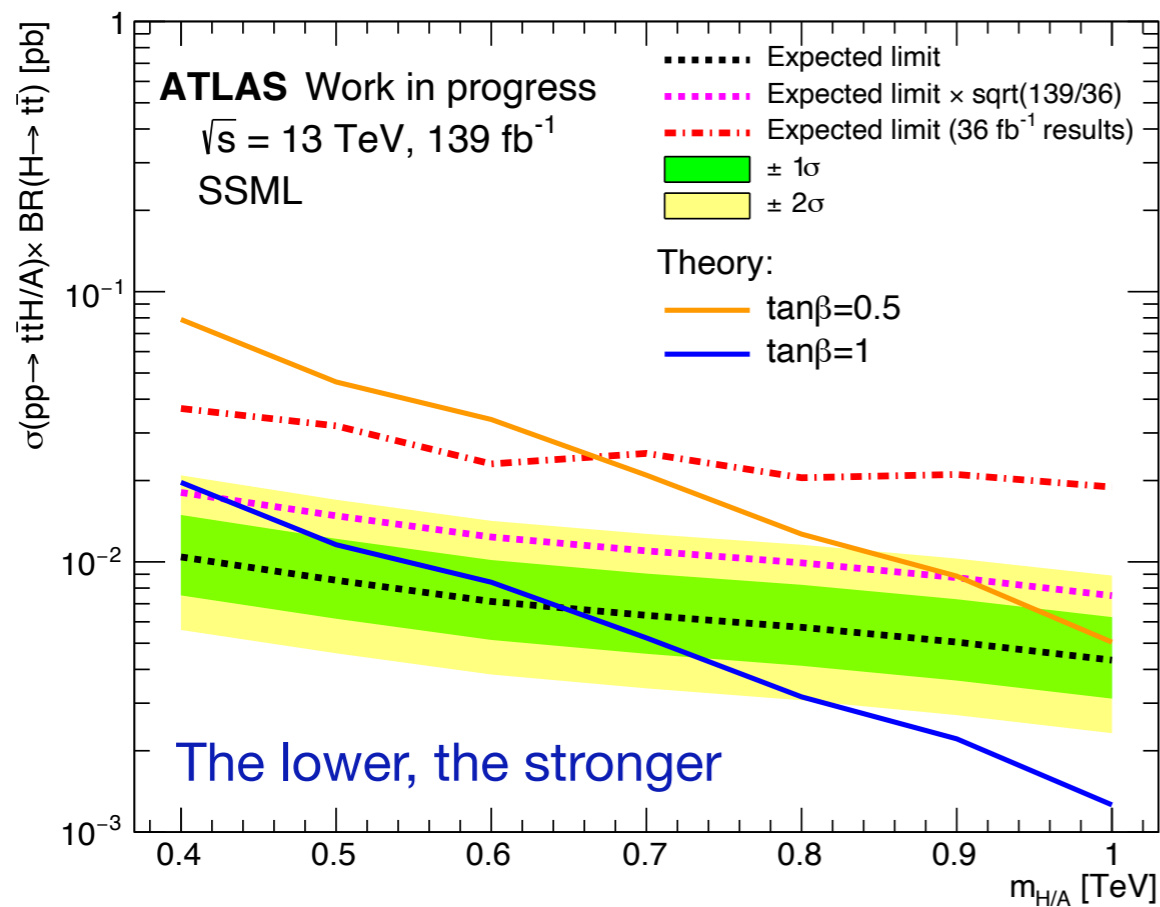
# Sequential BDT strategy - BSM pBDT

- **BSM pBDT** is trained with merged BSM  $t\bar{t}t\bar{t}$  and SM backgrounds using **XGBoost** package
- Reweigh background contribution to mimic the high SM BDT region in the training
- SM BDT is used as input variable for BSM pBDT



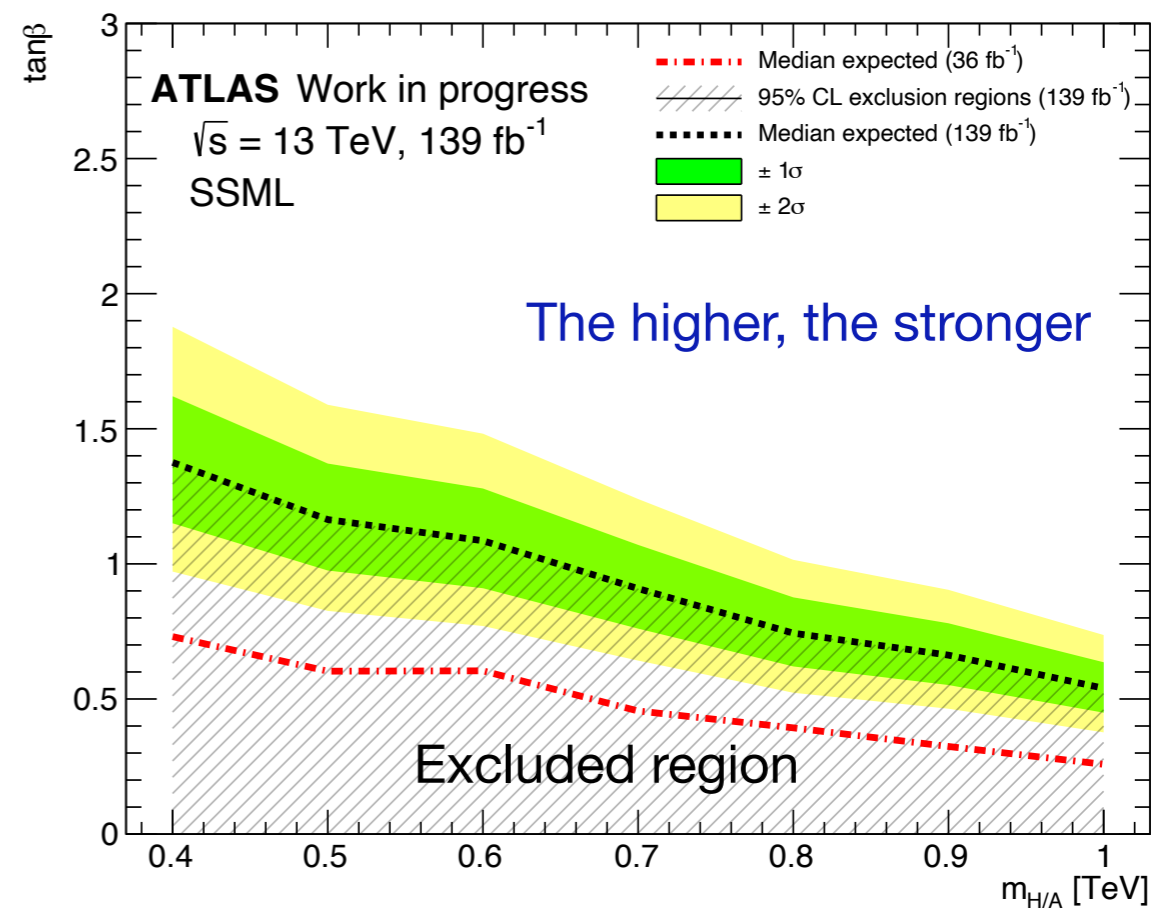
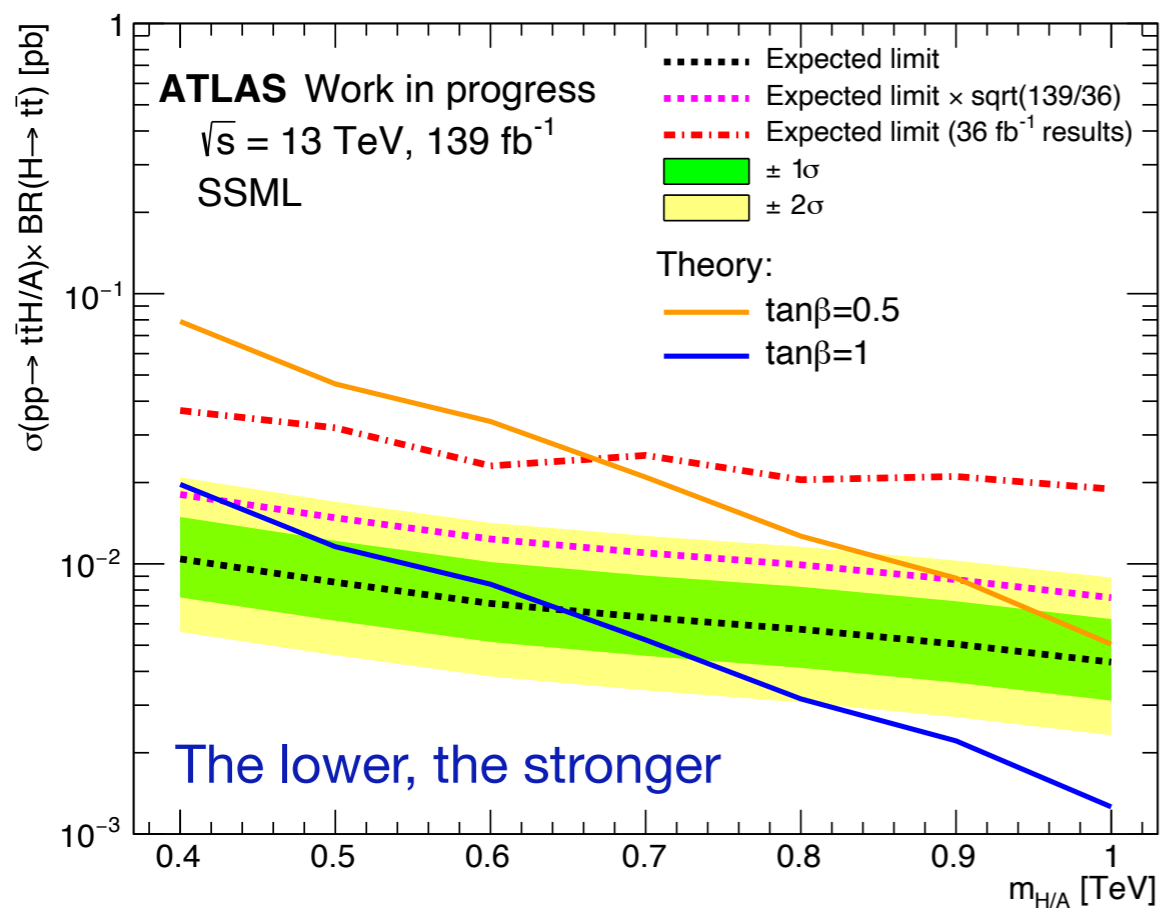
# Expected sensitivity

- About 75% stronger than ATLAS 36 fb<sup>-1</sup> SSML+b-jets analysis ([paper](#))
- Expected 50% stronger with statistical improvement (4 times more statistics, limits decrease by a factor of 2)
- Exclusion of  $\tan\beta = 1$  for  $m_H < 650$  GeV and exclusion of  $\tan\beta = 0.5$  for  $m_H < 1000$  GeV



# Summary and outlook

- Mass parameterized BDT is used to extract BSM signals in  $t\bar{t}t\bar{t}$  enriched phase space
- Same performance between pBDT and BDT at each mass point
- Work in progress to improve modeling of SM backgrounds ( $t\bar{t}W$ , SM  $t\bar{t}t\bar{t}$ , ...)
- Plan to update with  $t\bar{t}W$  QCD+EW NLO samples to have better description in the high  $N_{\text{jets}}$  region



# Backup



# Object selection

- Trigger and object definitions are inherited from SM  $t\bar{t}t\bar{t}$  analysis
- Single lepton and di-lepton triggers are used
- Same object definitions **but changing to PFlow jets and DL1r b-tagging**

Table 13: Overlap removal procedure

Reject	Against	Criteria
Electron	Electron	shared track, $p_{T,1} < p_{T,2}$
Muon	Electron	is Calo-Muon and shared ID track
Electron	Muon	shared ID track
Jet	Electron	$\Delta R < 0.2$
Electron	Jet	$\Delta R < 0.4$
Jet	Muon	NumTrack < 3 and (ghost-associated or $\Delta R < 0.2$ )
Muon	Jet	$\Delta R < \min(0.4, 0.04 + 10 \text{ GeV}/p_T(\mu))$

Ntuple production with AnalysisBase 21.2.120

Object	Electrons		Muons		Jets	b-jets	$E_T^{miss}$
	Loose	Tight	Loose	Tight			
$p_T$ [GeV]	> 10	> 28	> 10	> 28	> 25	> 25	$-\sum$ (calib. lep & jet & non-matched tracks at PV)
$ \eta $	< 1.37 or 1.52-2.47		< 2.5		< 2.5	< 2.5	
ID quality	mediumLH ECIDS ( $ee, e\mu$ )	tightLH ECIDS ( $ee, e\mu$ )	medium		Anti- $k_t$ (R=0.4) <b>EMPFLOW</b> with Jet cleaning + JVT tight	<b>DL1r 77%</b> (CDI: 2020-03-11-Sh228.v3)	Calo-based (Tight WP)
Isolation	none	FCtight	none	FCTightTrackOnly			
Track vertex: $ d_0/\sigma_{d_0} $ $ \Delta z_0 \sin \theta $	< 5 < 0.5 mm		< 3 < 0.5 mm				

# Template fit method

- **Fit to data in CRs to estimate the  $t\bar{t}W$  and fake backgrounds**
  - Simultaneously free float the NFs of backgrounds with shape from MC in a fit to data to correct the background estimation
  - 5 Normalization factors: fake electron from HF (HFe), fake muon from HF ( $HF\mu$ ), material conversion (Mat CO), virtual photon conversion ( $\gamma^*$ ) and  $t\bar{t}W$
- CRlowBDT was included in the template fit in the SM 4top analysis to make the background modeling close to the unblinding fit setup → **unblind bins where signal contamination < 5% (SM BDT < 0.25) in this region and include in the template fit**

Currently Included in the template fit

Region	Channel	$N_{\text{jets}}$	$N_{\text{b-jets}}$	other selection	Fitted variable
CRttbarCO	SSee    SSem	$4 \leq N_j < 6$	$\geq 1$	$0 < M_{ee@ConvV} < 0.1$ $200 < H_T < 500$	$M_{ee@PV}$
CR1b3Le	eee    eem		$= 1$	$100 < H_T < 250$	1 bin
CR1b3Lm	emm    mmm		$= 1$	$100 < H_T < 250$	1 bin
CRttW2L	SSem    SSmm	$\geq 4$	$\geq 2$	$M_{ee} < 0$ or $M_{ee} > 0.1$ , $ \eta(e)  < 1.5$ for $N_b = 2$ , $H_T < 500$ or $N_j < 6$ for $N_b \geq 3$ , $H_T < 500$	$\sum p_T(lep)$
CRlowBDT	SS+3L	$\geq 6$	$\geq 2$	$H_T > 500$ , SM BDT < 0.55 <sup>†</sup>	SM BDT
BSM SR	SS+3L	$\geq 6$	$\geq 2$	$H_T > 500$ , SM BDT $\geq 0.55$	BSM pBDT

\*  $M_{ee@ConvV}(@PV)$  is defined as the invariant mass between the track associated to  $e$  and the closest track at conversion (primary) vertex

† In the template fit, SM BDT < 0.25 is included in the fit. In the realistic Asimov fit, SM BDT < 0.55 is included

# Input variable for SM BDT and BSM pBDT

- Input variable of SM BDT inherited from SM 4top analysis with additionally decorrelating the **sum of continuous b-tagging scores** into two variables to have better separation between 4top-like events and  $t\bar{t}W \geq 7$ -jet events
- SM BDT is used in BSM pBDT training

Input variables (SM BDT)
jet_sum_DL1r_Continuous_leadingFour
nJets
deltaR_ll_min
leading_jet_pT
lep_0_pt
leading_bjet_pT
deltaR_ll_sum
HT_jets_noleadjet
jet5_pt
deltaR_bj_min
met_met
deltaR_lb_max
jet1_pt

Input variables (BSM pBDT)
deltaR_ll_min
deltaR_ll_sum
SphericityJets
SphericityXYJets
met_met
HT_all
<b>SMBDT</b>
mH

MVA strategies	BDTs used	Samples used	Parametrized
Sequential BDT	SM BDT	SM 4top vs. bkg	✗
	BSM pBDT	Merged BSM vs. SM 4top (& bkg)	✓

# Fit setup

## ✓ Plain Asimov Fit: (S+B and SRCR with Asimov data)

- Probe sensitivity (limits) and expected constraints on fitted variables based on pre-fit background model

## ✓ Real data CRs-Only Fit: (BOnly and unblinded CROnly, fit with real data)

- Derive a "close-to-final" post-fit background estimation w/o looking at data in SR
- Template fit method: CRttbarCO, CR1b3Le, CR1b3Lm, CRttW2L

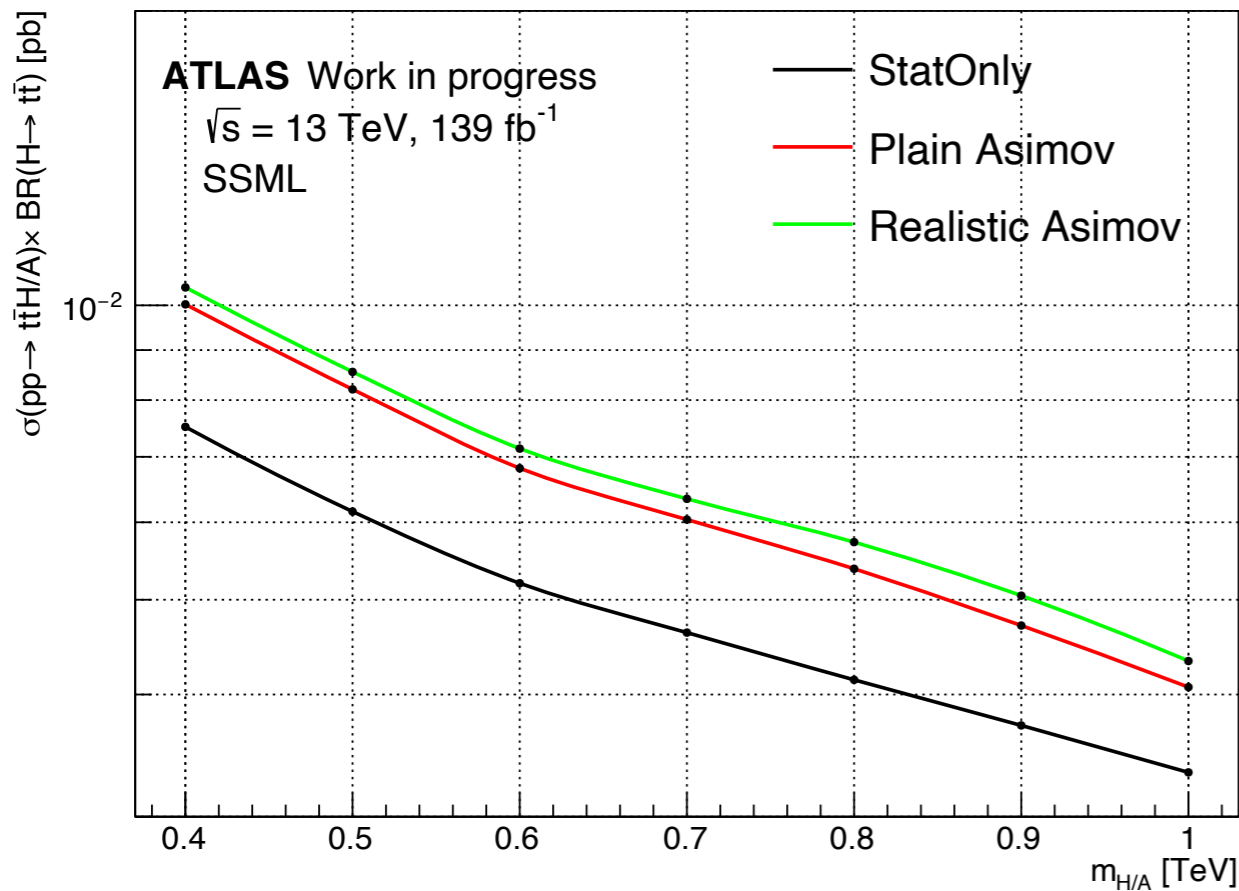
## ✓ Realistic Asimov Fit: (S+B and SRCR with Asimov data with $\mu_{\text{BSM}t\bar{t}\bar{t}}$ free-floating with nominal set to 0)

- Use pseudo-data corresponding to post-fit background prediction from Real CRs-Only fit
- Obtain the "close-to-final" values of sensitivity (limits), NP pulls and constraints w/o looking at data in any regions

Fit setup	Control regions	Signal regions	Fitted parameters	$\mu_{\text{BSM}t\bar{t}\bar{t}}$	Purpose
Plain Asimov	MC	MC	$\alpha_{\text{PlainAsimov}}$	free [ $\mu=0$ ]	expected constraints and sensitivity
Real CRs-Only	data	not included	$\alpha_{\text{CRs-Only}}$	not included	post-fit bkg
Realistic Asimov	MC[ $\alpha_{\text{CRs-Only}}$ ]	MC[ $\alpha_{\text{CRs-Only}}$ ]	$\alpha_{\text{RealisticAsimov}}$	free [ $\mu=0$ ]	expected sensitivity (post-fit bkg injected)

# Systematic uncertainties

- Hard to separate SM  $t\bar{t}t\bar{t}$ , 3top and  $t\bar{t}W \geq 8$  jets events from BSM  $t\bar{t}t\bar{t}$  signals
- Large impact from SM  $t\bar{t}t\bar{t}$ , 3top and  $t\bar{t}W \geq 8$  jets related systematic uncertainties



Pre-fit impact on  $\mu$ :  
 $\theta = \hat{\theta} + \Delta\theta$  (light blue)  
 $\theta = \hat{\theta} - \Delta\theta$  (light cyan)

Post-fit impact on  $\mu$ :  
 $\theta = \hat{\theta} + \Delta\hat{\theta}$  (dark blue)  
 $\theta = \hat{\theta} - \Delta\hat{\theta}$  (dark cyan)

—●— Nuis. Param. Pull

