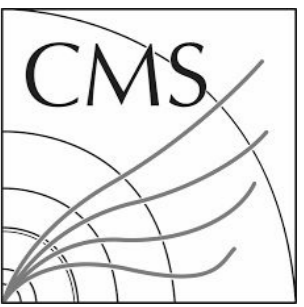


Workshop

Dubrovnik, HR
May 31th, 2022



ATLAS + CMS

Non-resonant

$HH \rightarrow bbbb$

Daniel Guerrero

and

Rafael Teixeira de Lima

on behalf of the ATLAS and CMS Collaborations

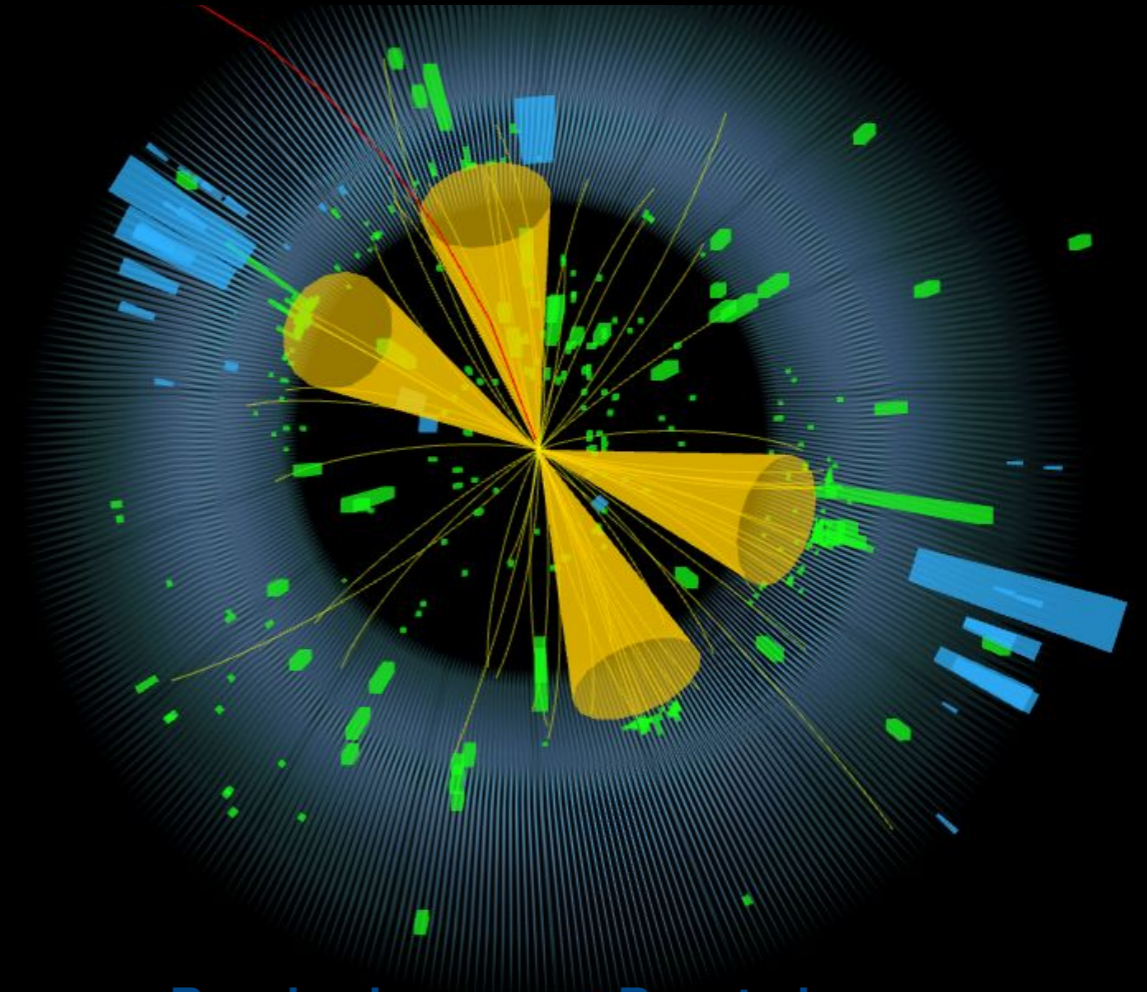


Outline



CMS Experiment at the LHC, CERN
Data recorded: 2016-Aug-13 15:04:59.113664 GMT
Run / Event / LS: 278802 / 7164845 / 11

- The $HH \rightarrow bbbb$ channel
- ATLAS and CMS Run-2 Overview
 - Jet tagging & Higgs reconstruction
 - Trigger
 - Event categorization
 - Background modeling
 - Discriminant observables
 - Systematics
 - Results: SM, anomalous $k1$ and $k2v$
- Conclusions

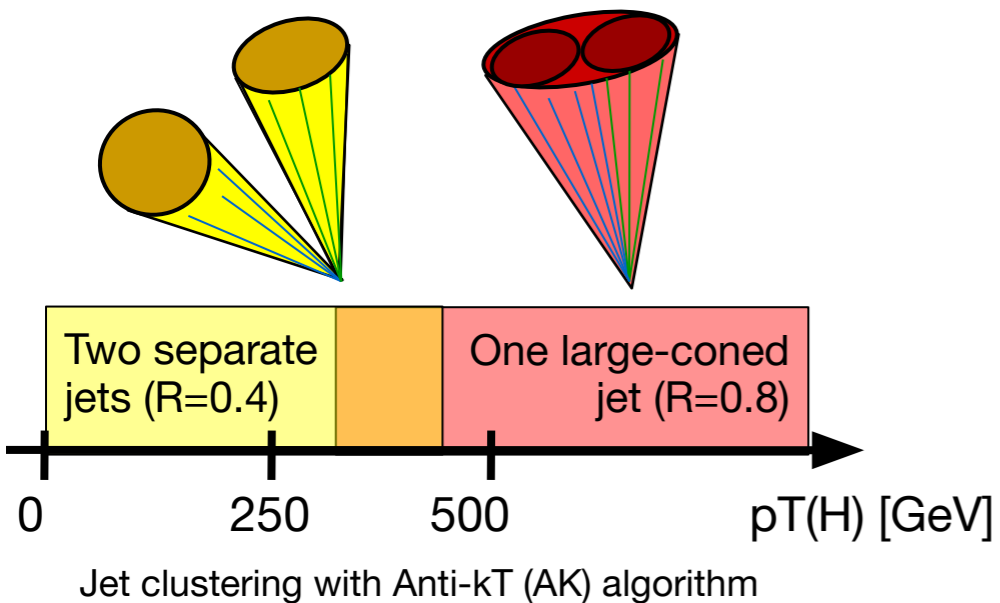


More details in references: **Resolved** [ATLAS \(CONF-2022-035\)](#) and **Resolved** [CMS \(ArXiv:2202.09617, ArXiv:2205.06667\)](#) **Boosted**
NEW RESULT! **Both submitted to PRL!**

The bbbb decay channel

It has the HH largest rate ~ 1500 SM events produced in the LHC Run-2 period ($L \sim 140 \text{ fb}^{-1}$)!
 But searching for signal events is challenged by
 the large production of multi-jet bkg. events (QCD multijet $\sim 90-95\%$, top quarks $\sim 5-10\%$)

$H \rightarrow bb$ experimental fingerprint depends on Higgs p_T



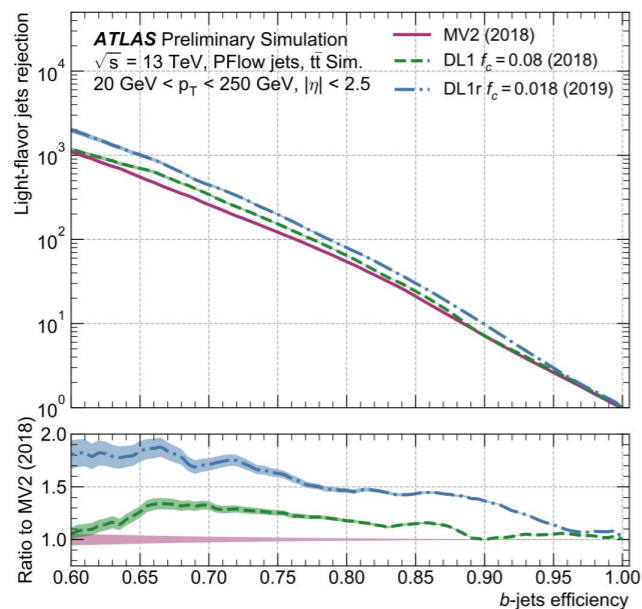
Experimental challenges:

- b quark jet or H jet identification w.r.t large usdg/c jet bkg
- Online trigger algorithms are complex
 - Depends on L1 seed, HLT tracking, jet reco/cal, b-tagging, etc
 - Constrained by L1 rate, HLT CPU limit & output rate
 - Consistency with offline algorithms (e.g. b-tagging)
- Higgs boson reconstruction affected by
 - Large jet combinatorics
 - Missing energy from neutrinos in semi-leptonic B decays
 - Jet constituents from ISR, FSR & Pile-up
- Precise model and rejection of multijet bkg are crucial

Small-R jet b-tagging (resolved)

The b-jet candidates are identified using machine learning algorithms

ATLAS



DL1r performance plots

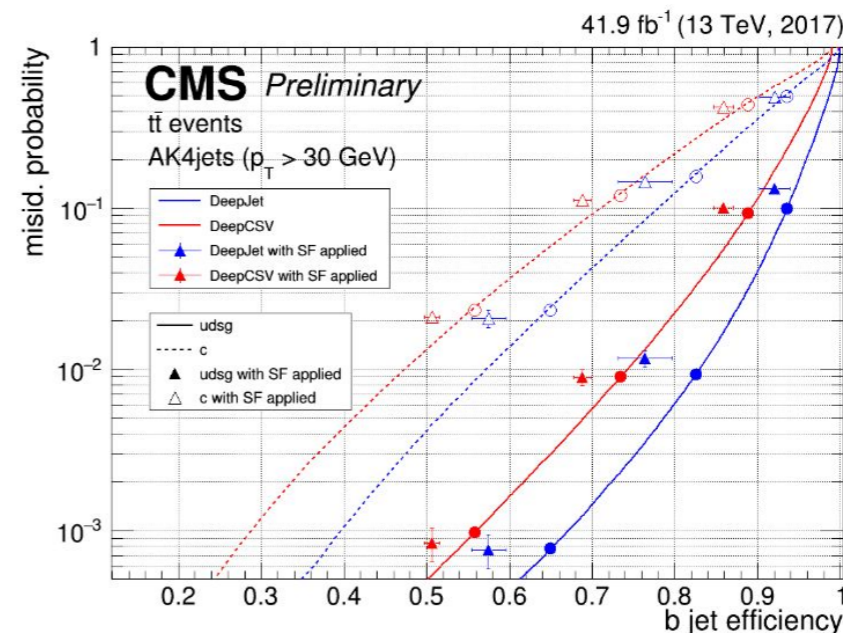
Best performing AK4 b-jet tagger: DL1r

- DNN + RNN (tracks' impact parameters)
- 10% improvement in efficiency for same bkg rejection

b-jet candidates:

- Pass DL1r at **77% b-jet efficiency** (measured in ttbar)
- Muon-in-jet information used to correct jet pT

CMS



DeepJet performance plots

Best performing AK4 b-jet tagger: DeepJet

- DNN multioutput classifier
- 10% eff. Improvement w.r.t previously used (DeepCSV)

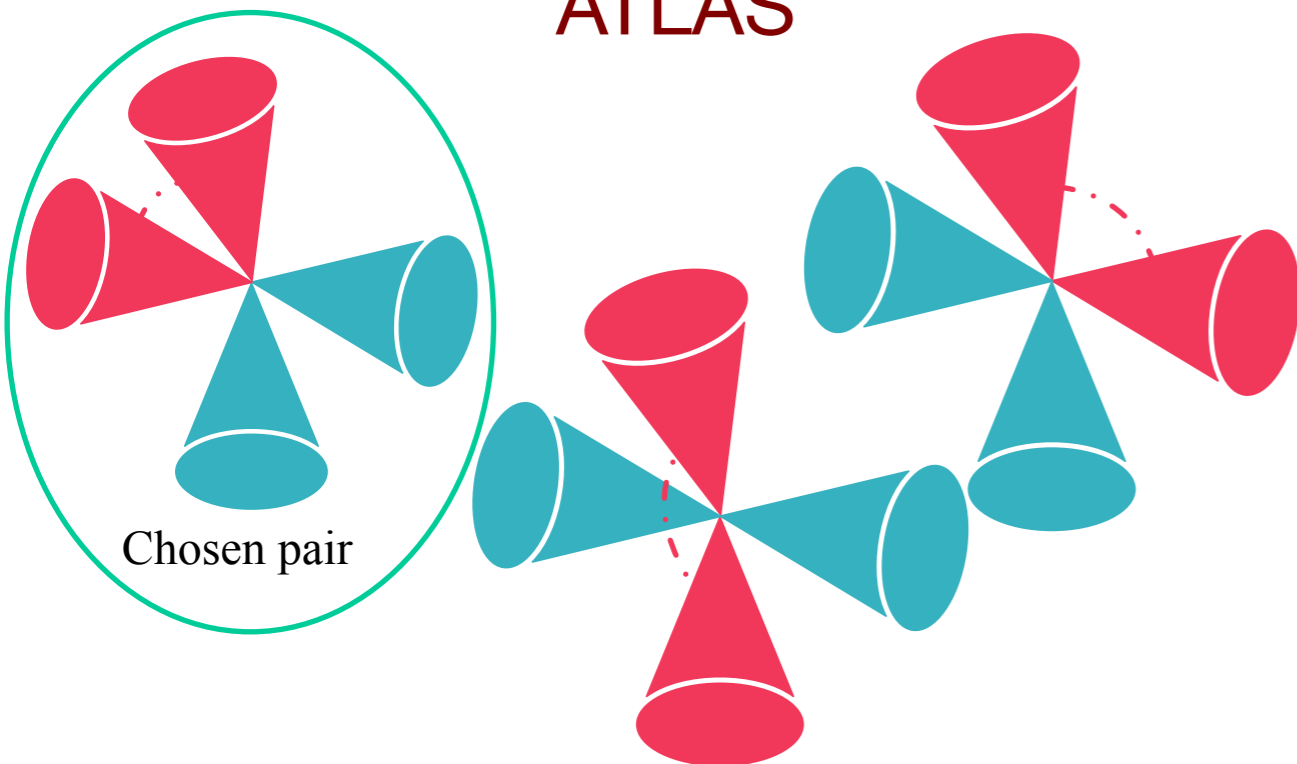
b-jet candidates:

- Pass DeepJet WP with **~75% efficiency** (measured in ttbar)
- DNN b-jet energy regression is used to correct jet pT

Higgs boson reconstruction (resolved)

The challenge: Four b-jet candidates → 3 pairing possibilities

ATLAS



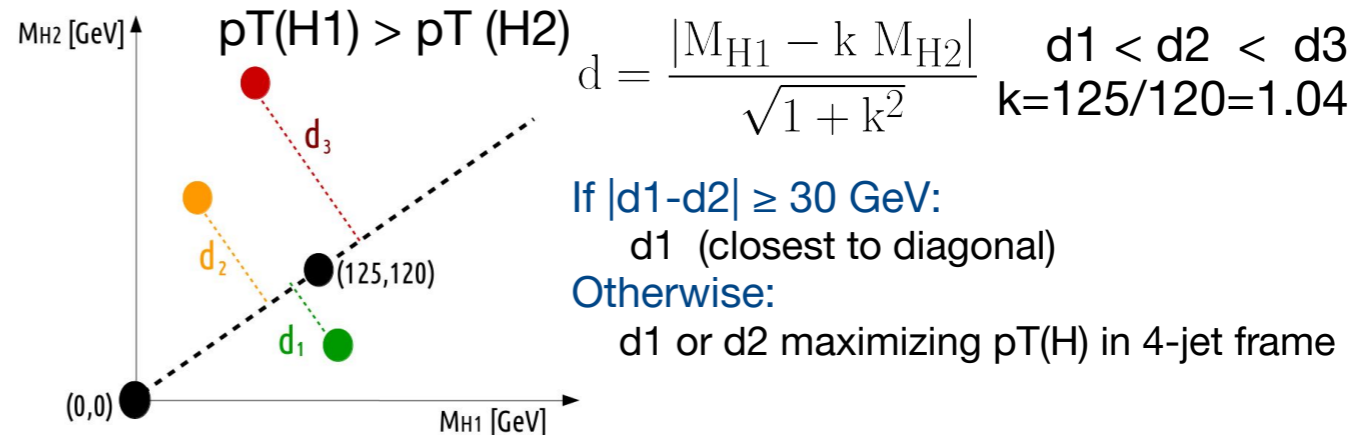
Choose pairing that minimizes ΔR distance
between jets in the leading Higgs candidate (H1)

90% accurate for SM ggF events

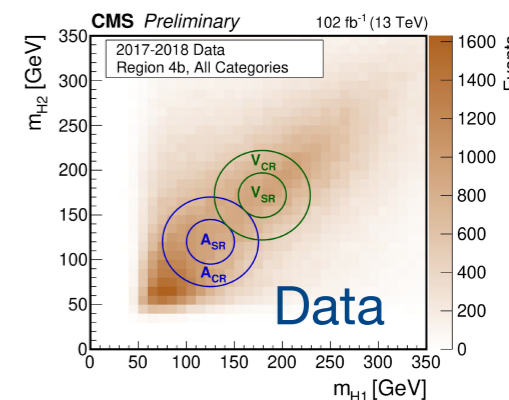
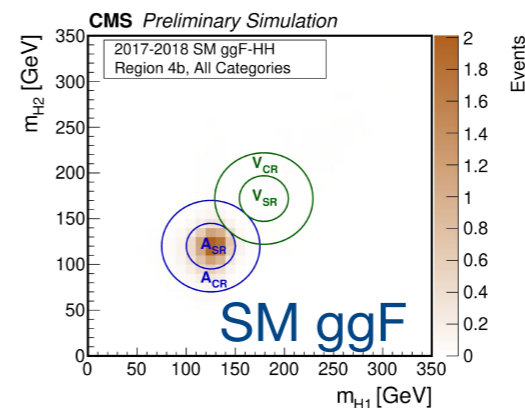
where 4 b-jets = 4 b-quarks from Higgs decays

CMS

Jet pairing based on $m_{H1}-m_{H2}$ plane information



Very good performance (e.g. 96% accurate for SM ggF)
Maximizing signal collection w/o bkg. sculpting near the m_H



H-tagging & reconstruction (CMS boosted)



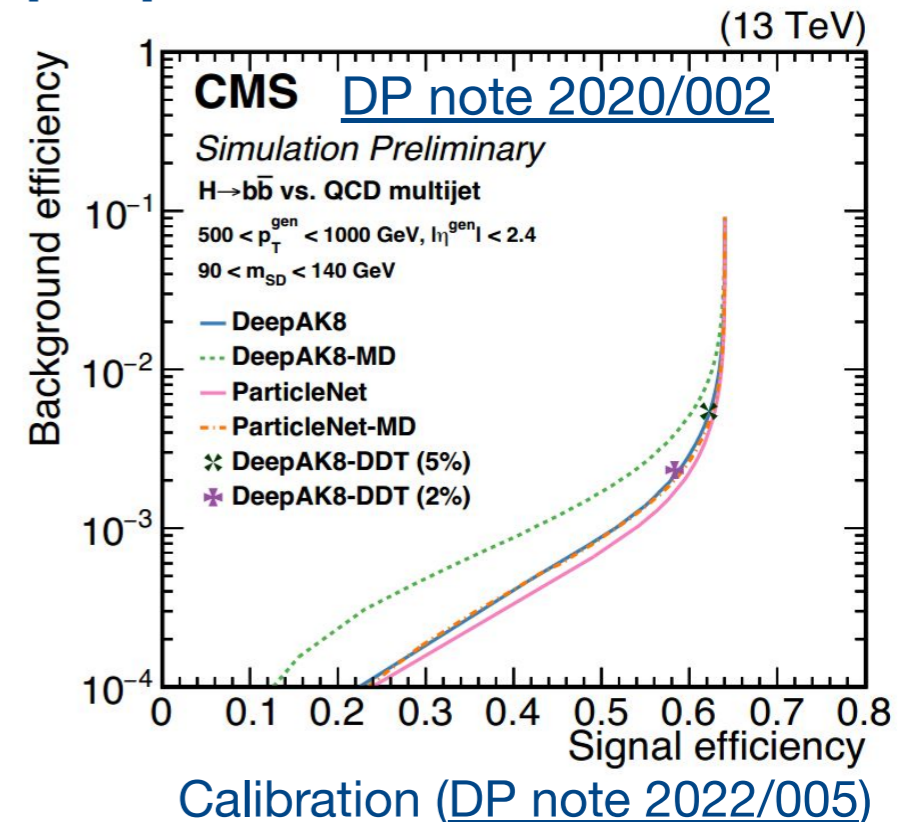
Best performing AK8 jet tagger: [ParticleNet](#)

- Architecture based on [Graph Neural Networks](#)
- Mass-decorrelated version identifies 2-prong hadronic decays of highly boosted particles (e.g. $X \rightarrow b\bar{b}$)

$$D_{bb} \text{ discriminant} = P[X \rightarrow b\bar{b}] / P[X \rightarrow b\bar{b}] + P[\text{QCD}]$$

- Input features:
 - Jet constituents PF candidates & secondary vertices
- Performance: 0.1% bkg eff. at 50% signal eff:
 - Bkg rejection improved by a ~factor of 2 per jet than previously
- Same architecture is used for regression of the jet mass (m_{reg})

More details on
[‘ATLAS+CMS jet and flavor tagging’ talk](#)
by Loukas tomorrow!



ATLAS

CMS

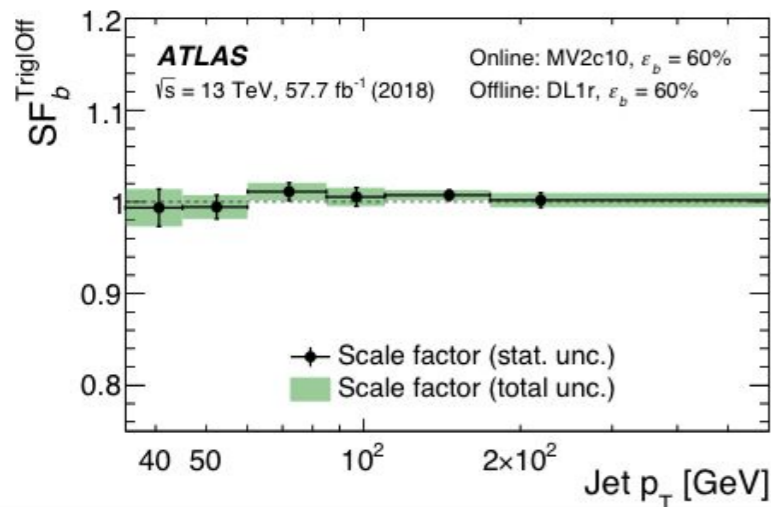
2b2j: 2 b-tagged jets + 2 extra jets

- $p_T > 35$ GeV for all jets
- b-jet eff. between 40% and 70%
- Important for low m_{HH} events

2b1j: 2 b-tagged jets + 1 extra jet

- $p_T > 35$ GeV for b-jets, $p_T > 100\sim 150$ GeV for extra jet
- b-jet eff. between 40% and 70%
- Important for high m_{HH} events

ATLAS b-jet trigger



Resolved:

Triggers require 4 jets with 3 b-tagged jets

■ 2016:

- 4j with $p_T > 45$ GeV
- 2j with $p_T > 30$ GeV, 2j with $p_T > 90$ GeV

■ 2017 (2018):

- 4j with $p_T > 40, 45, 60, 75$ GeV, $HT > 300$ (330) GeV

Boosted:

Combination of several single-jet and HT triggers:

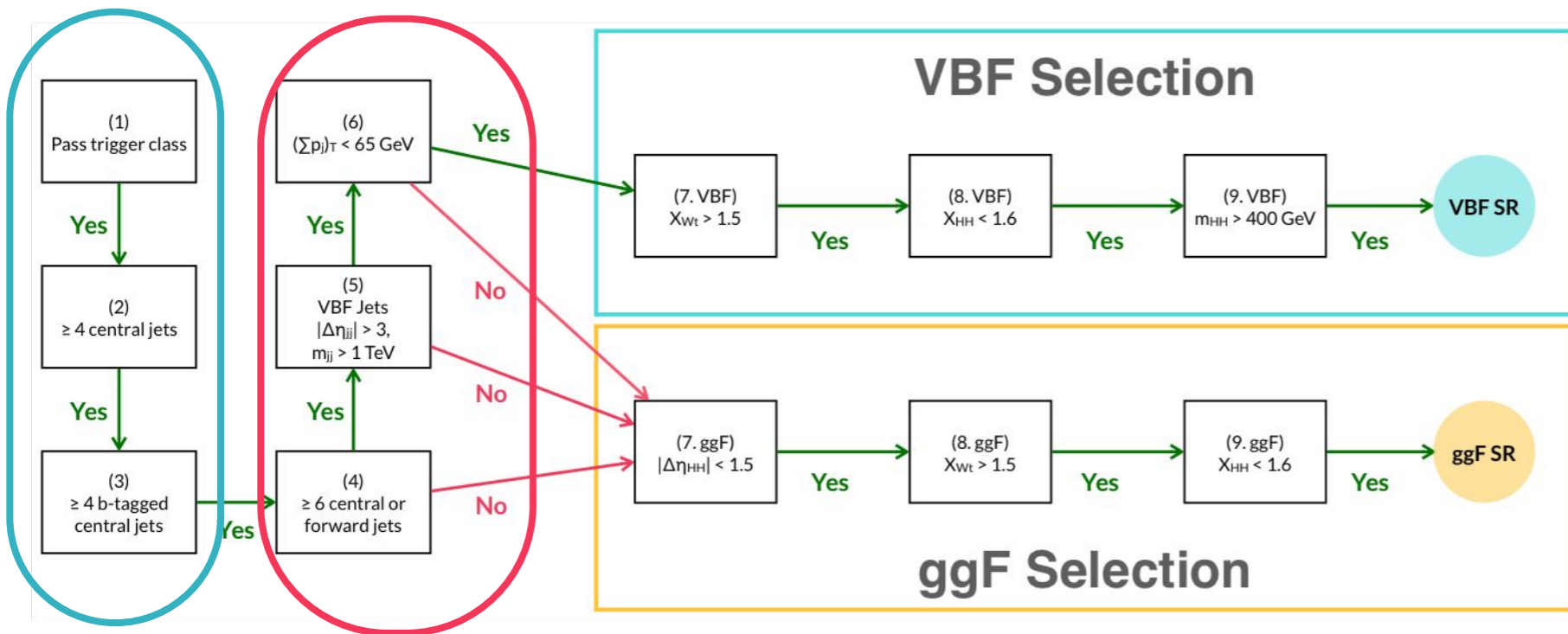
■ Requirements on:

- HT, jet p_T , Trimmed mass & double b-tagging

■ Fully efficient for Jet $p_T > 500$ GeV

Event Categorization (ATLAS resolved)

Same analysis selection is performed in 2016, 2017 & 2018 datasets, $L_{\text{total}}=126 \text{ fb}^{-1}$



Common selection

ggF vs VBF separation

Top veto cut

Signal region cut

$$X_{Wt} = \sqrt{\left(\frac{m_W - 80.4 \text{ GeV}}{0.1 m_W}\right)^2 + \left(\frac{m_t - 172.5 \text{ GeV}}{0.1 m_t}\right)^2}$$

$$X_{HH} = \sqrt{\left(\frac{m_{H1} - 124 \text{ GeV}}{0.1 m_{H1}}\right)^2 + \left(\frac{m_{H2} - 117 \text{ GeV}}{0.1 m_{H2}}\right)^2}$$

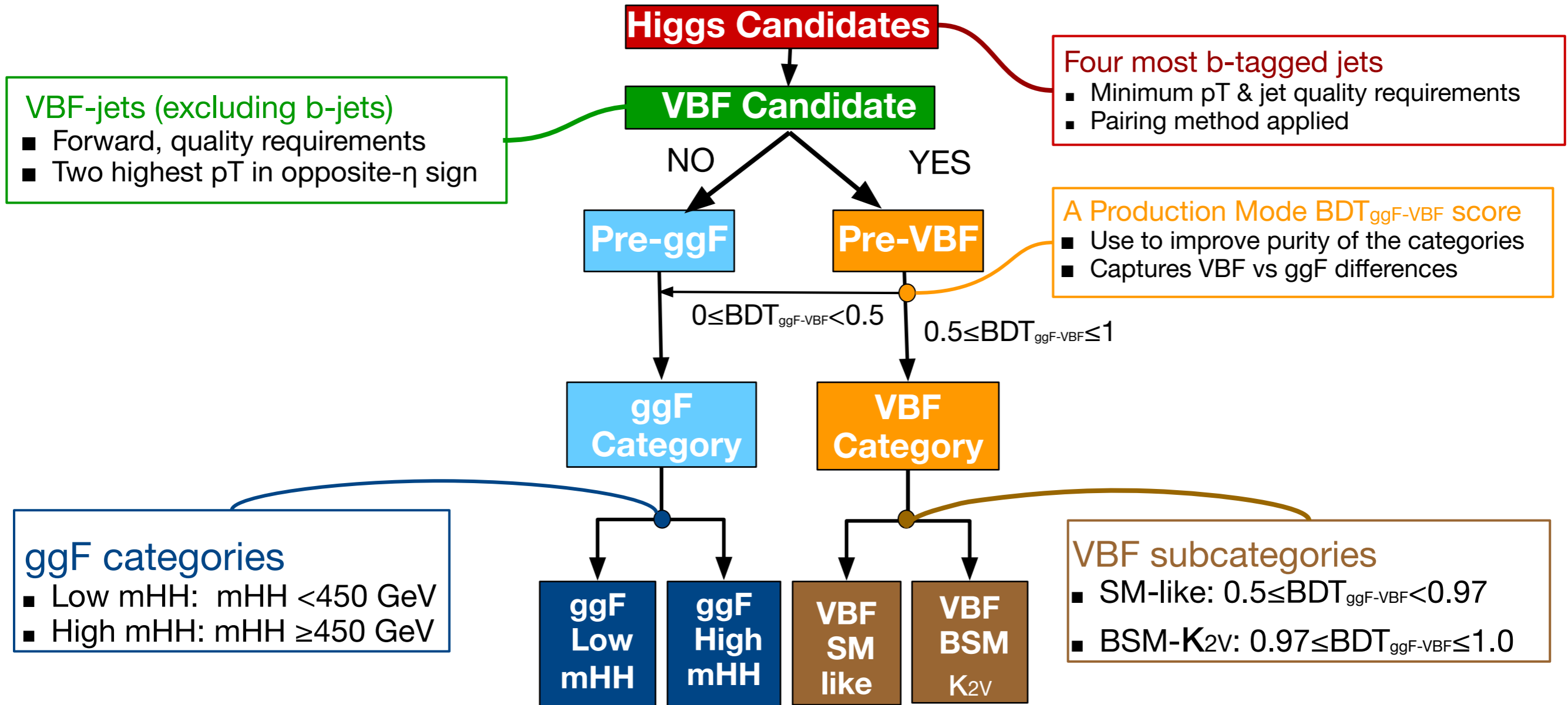
Categories

$ \Delta\eta_{HH} $	$ \Delta\eta_{HH} $
< 1.5	> 1.5
VBF Cat 1	VBF Cat 2

ggF Cats.	$ \Delta\eta_{HH} $	$0.5 > \Delta\eta_{HH} $	$1.0 > \Delta\eta_{HH} $
	< 0.5	$ \Delta\eta_{HH} < 1.0$	$ \Delta\eta_{HH} < 1.5$
$X_{HH} < 0.95$	1	2	3
$0.95 > X_{HH}$ $X_{HH} < 1.6$	4	5	6

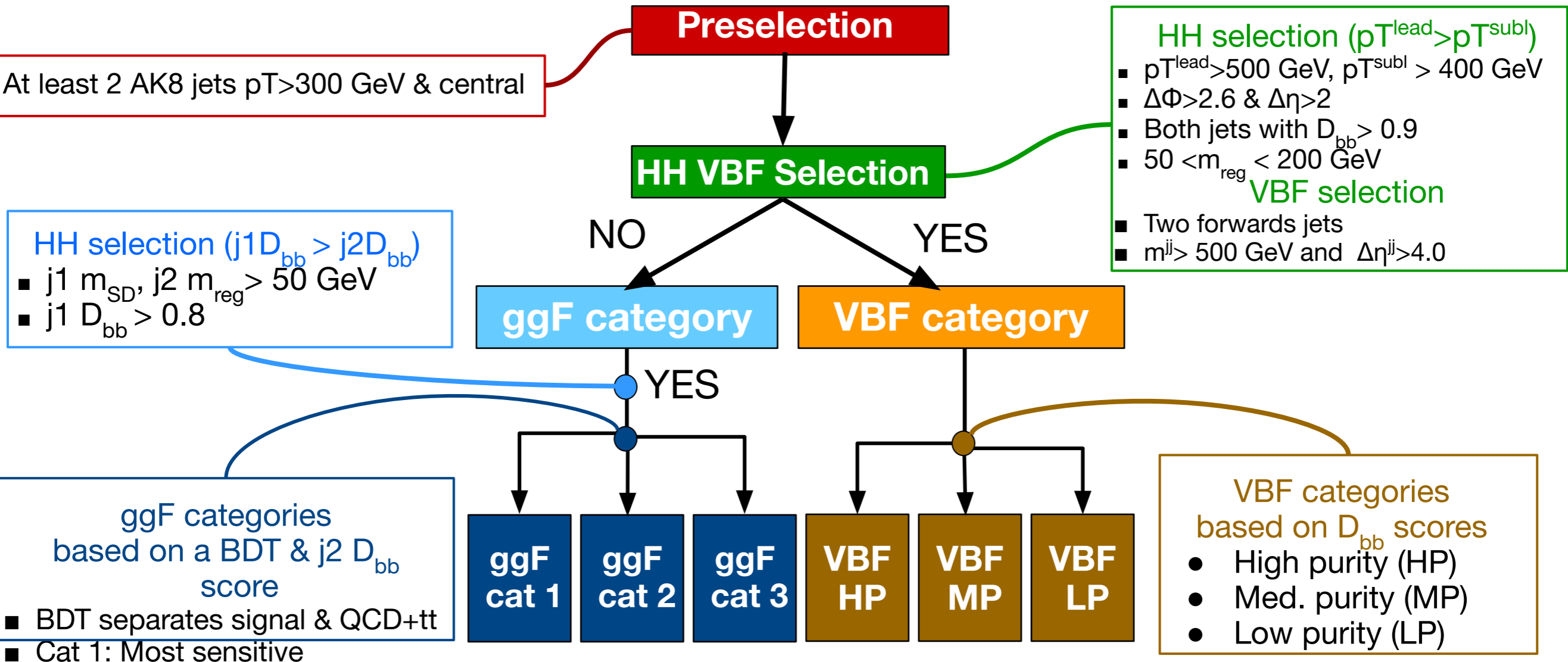
Event Categorization (CMS resolved)

Same analysis selection is performed in 2016 & 2017-2018 datasets, $L_{\text{total}}=138 \text{ fb}^{-1}$



Event Categorization (CMS boosted)

Same analysis selection is performed in 2016, 2017 & 2018 datasets, $L_{\text{total}} = 138 \text{ fb}^{-1}$



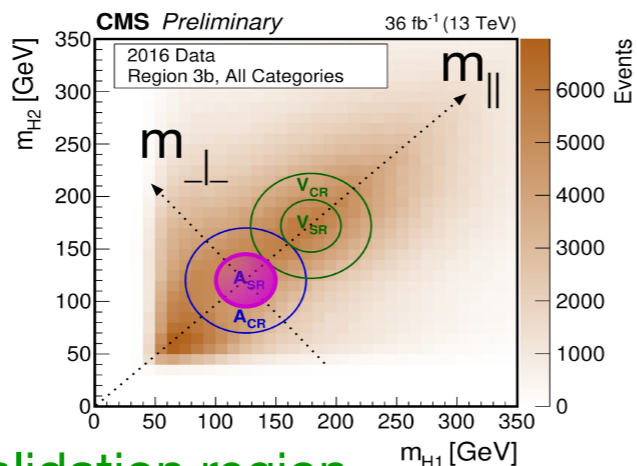
Background estimation (resolved)

Data-driven multijet bkg model using 'low-btag' data to derive '4-btag' background distributions

CMS

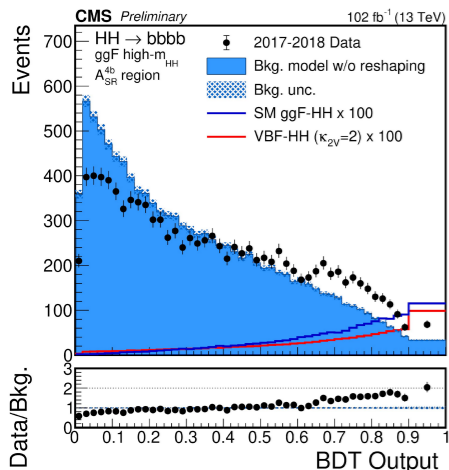
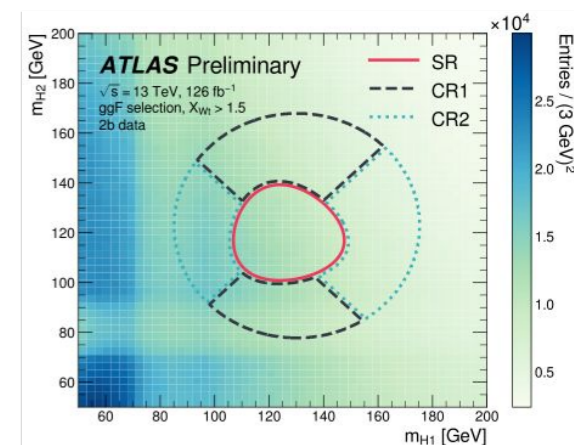
ATLAS

- 3b data → 4b full bkg
- Derived in CR data (ring)
 - Transfer factor (4b/3b)
 - Residual differences corrected via weights from [BDT re-weighting](#)
- Applied to SR (circle)
- Method is fully verified in validation region

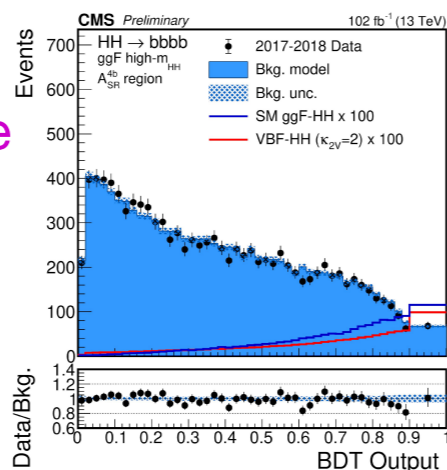


Similar to CMS method, but three main differences:

- Use 2b data instead of 3b (use 3b as validation)
- Split ring in two: nominal and alternative models
- Use DNN instead of BDT

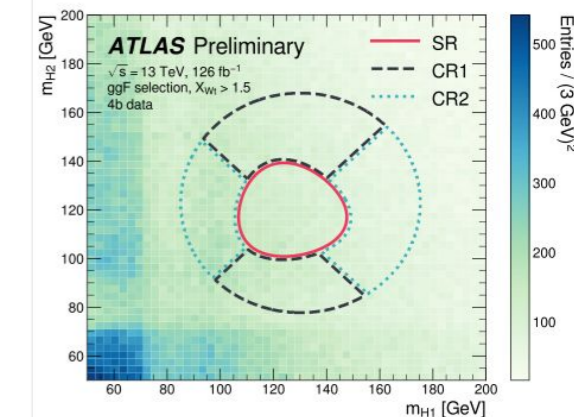


Performance in SR(4b) region



Many validation regions:

- Shifted region (like CMS)
- 3b+1f region
- Reversed $|\eta_{HH}|$ selection



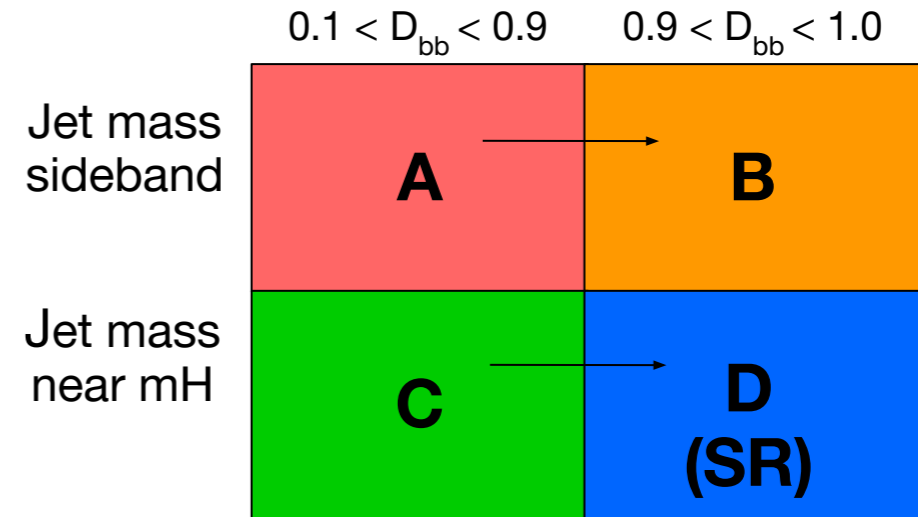
Background estimation (CMS boosted)

ggF Categories

- QCD bkg. is derived using parametric alphabet method:
 - Fail region: $j_2 D_{bb}$ score < 0.95 & $BDT > 0.03$
 - Model in pass region = fail region shape x t.f. ($j_2 m_{reg}$)
- $t\bar{t}$ bkg. is modeled from MC:
 - Correction from $t\bar{t}$ (had) & $t\bar{t}$ (semilep.) in CRs
 - $t\bar{t}$ recoil & D_{bb} shape corrections
- Other minor bkg (Single H, VV, etc) from MC

VBF Categories

- QCD bkg derived using ABCD method:
 - Model D = Shape from C region x tf



- $t\bar{t}$ bkg is modeled from MC
 - Corrections from semileptonic tt events
 - Normalization & bb-mistagging SFs

More details on
[Data-driven QCD background methods talk](#)
 by Matej tomorrow!

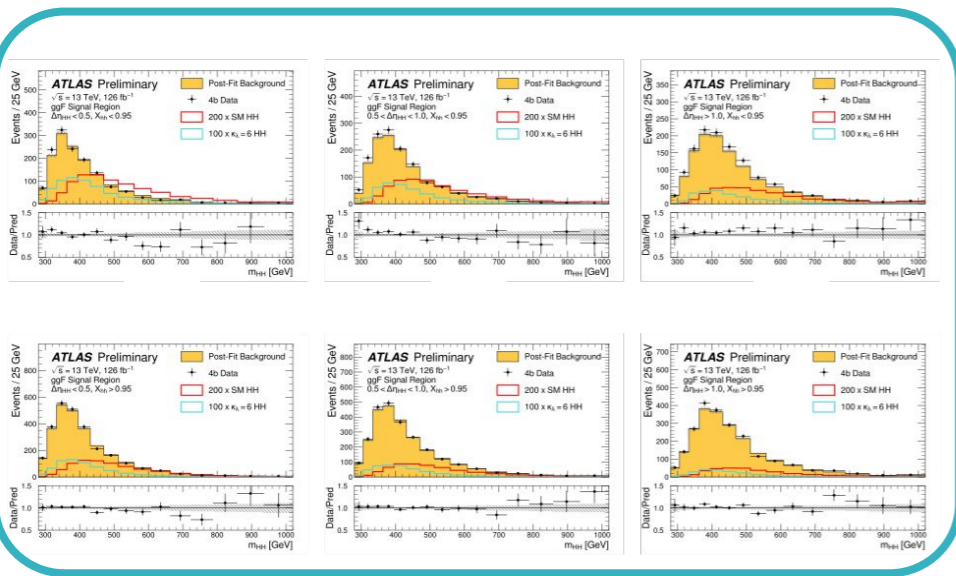
Discriminant observables (resolved)

The selected observables maximize the analysis sensitivity

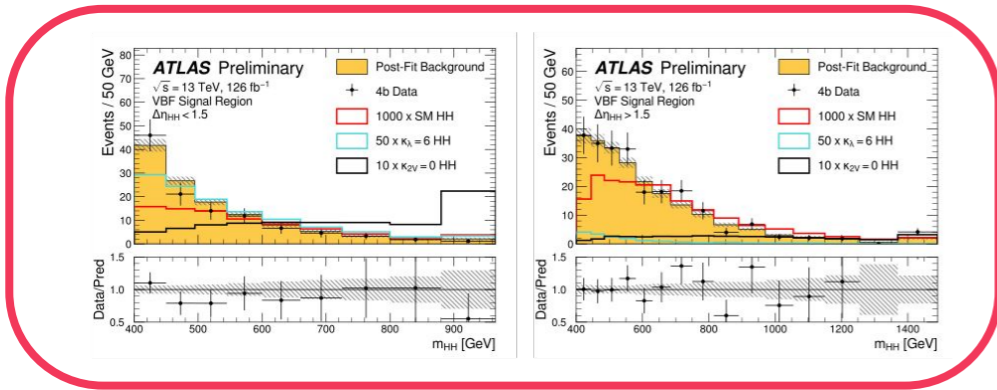
ATLAS

CMS

Gluon-gluon fusion



VBF

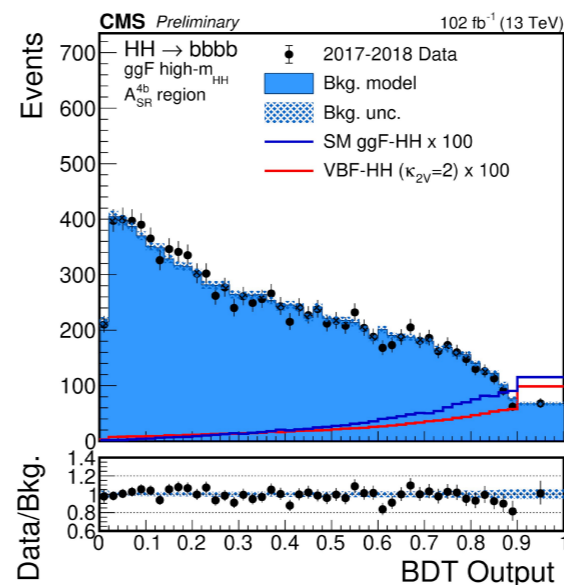


ggF categories: BDT distribution

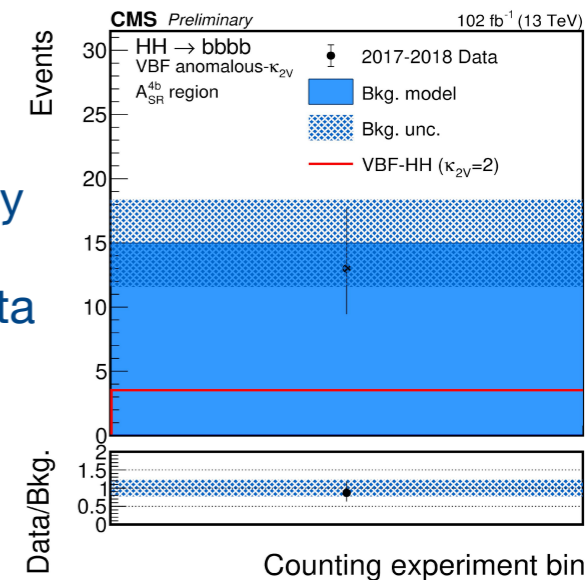
- Bkg model stats. enables the model of a BDT discriminant
- Discriminant is trained by category (16 variables)

VBF SM category: mHH distribution

VBF BSM category: Counting experiment

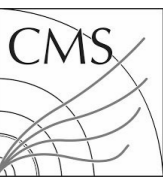


B-only
Fit
to data



Signal extraction performed by fitting signal region distributions of all categories/datasets

Discriminant observables (CMS boosted)

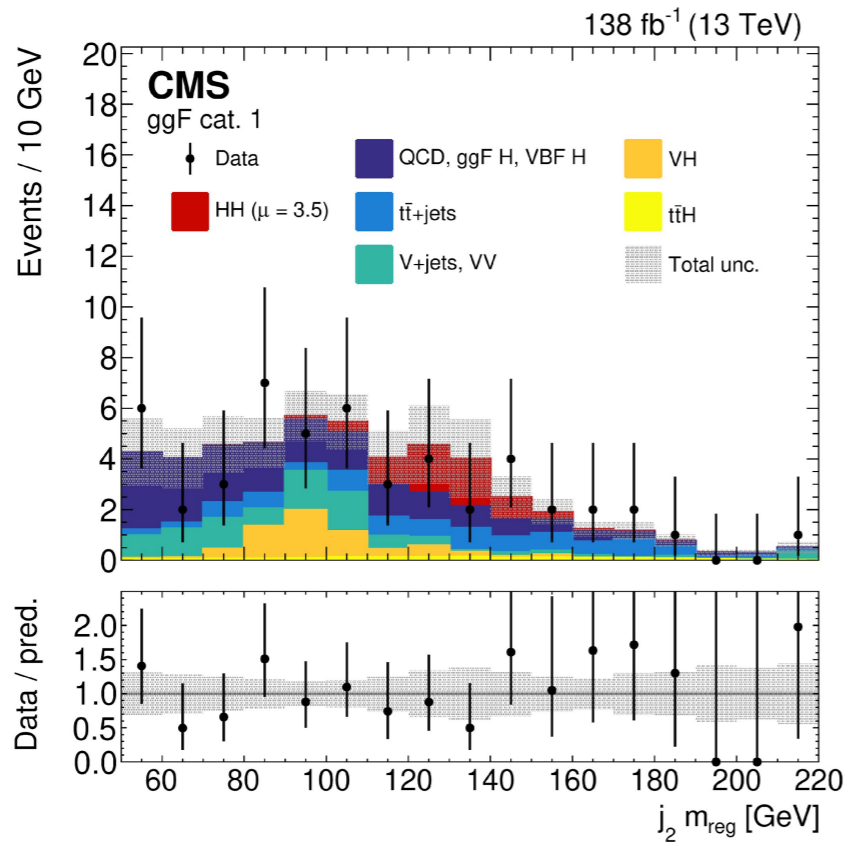


The selected observables maximize the analysis sensitivity

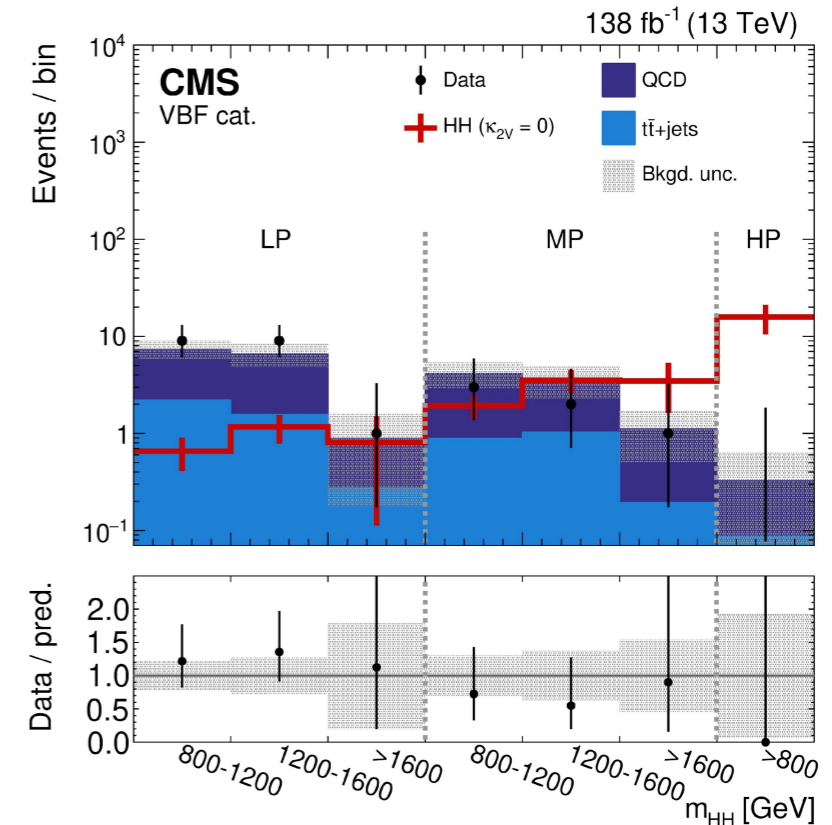
ggF categories: j_2 regressed mass

VBF categories: m_{HH} distribution (3 or 1 bin)

S+B fit to data
on most
sensitive
ggF category



B-only
fit to data



Signal extraction performed fitting all categories in signal and control regions

The bbbb resolved analyses are dominated by background estimate uncertainties!

ATLAS

Background uncertainties:

- Statistical: 2b statistics + DNN variation under bootstrapped deep ensembles (100 trainings)
- Alternative vs nominal estimate (CR1 vs CR2)
- 3b1f region non-closure
- Normalization uncertainty from 2b/4b CR



CMS

Background uncertainties:

- Statistical uncertainty in SR(3b) region (dominant)
- Transfer factor from CR 4b/3b regions
- Shape variation (Alternative CR definition vs nominal)
- Validation yields non-closure
- Validation limited statistical precision

ATLAS & CMS signal uncertainties

- Luminosity , PU, trigger
- b-tagging, jet scale and resolution,
- PS, PDF, factorization scales
- PS dipole recoil in VBF signals (CMS-only)
- Cross section & branching ratio

Systematics (CMS boosted)

Breakdown of the impact
of systematics
on measured SM signal strength ($\mu=3.5$)
grouped by their statistical, systematic
and theoretical nature

Uncertainty source	$\Delta\mu$	
Statistical	+2.55	-2.30
Signal extraction	+2.32	-2.06
QCD multijet modeling	+1.12	-1.01
$t\bar{t}$ modeling	+0.28	-0.19
Systematic	+2.09	-0.89
Simulated sample size	+0.55	-0.55
$D_{b\bar{b}}$ selection	+0.72	-0.32
Jet energy and mass scale and resolution	+0.54	-0.39
Trigger selection	+0.26	-0.03
Luminosity measurement	+0.13	-0.04
Pileup modeling	+0.05	-0.06
Other experimental uncertainties	+0.05	-0.03
Theoretical	+0.63	-0.63
Total	+3.30	-2.47

Results: SM

No excess of events is observed relative to the background-only hypothesis
95% CL upper limits are set on the SM HH (ggF+VBF) production

ATLAS

Observed (expected) limit (resolved):

5.4 (8.1) x SM prediction

(~3 x more sensitive than before)

CMS

Observed (expected) limit (resolved):

3.9 (7.8) x SM prediction

(~5 x more sensitive than before)

Best observed constraint on bbbb channel!

Observed (expected) limit (boosted):

9.9 (5.1) x SM prediction

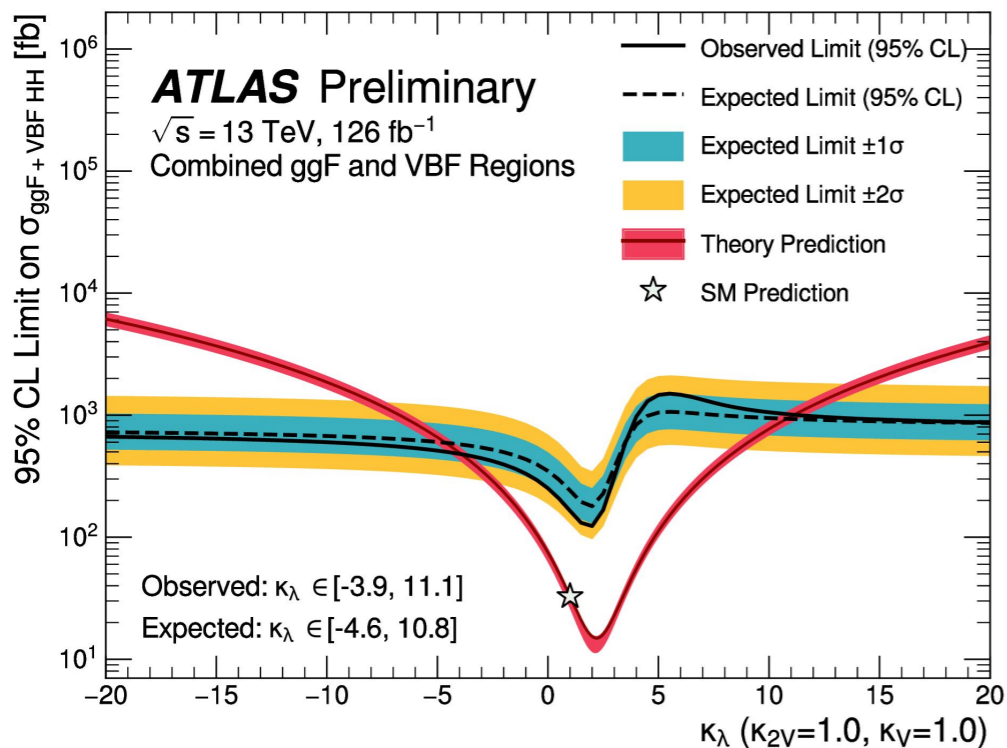
(~30 x more sensitive than before)

Significant improvements using full Run-2 data!

Results: Anomalous κ_λ

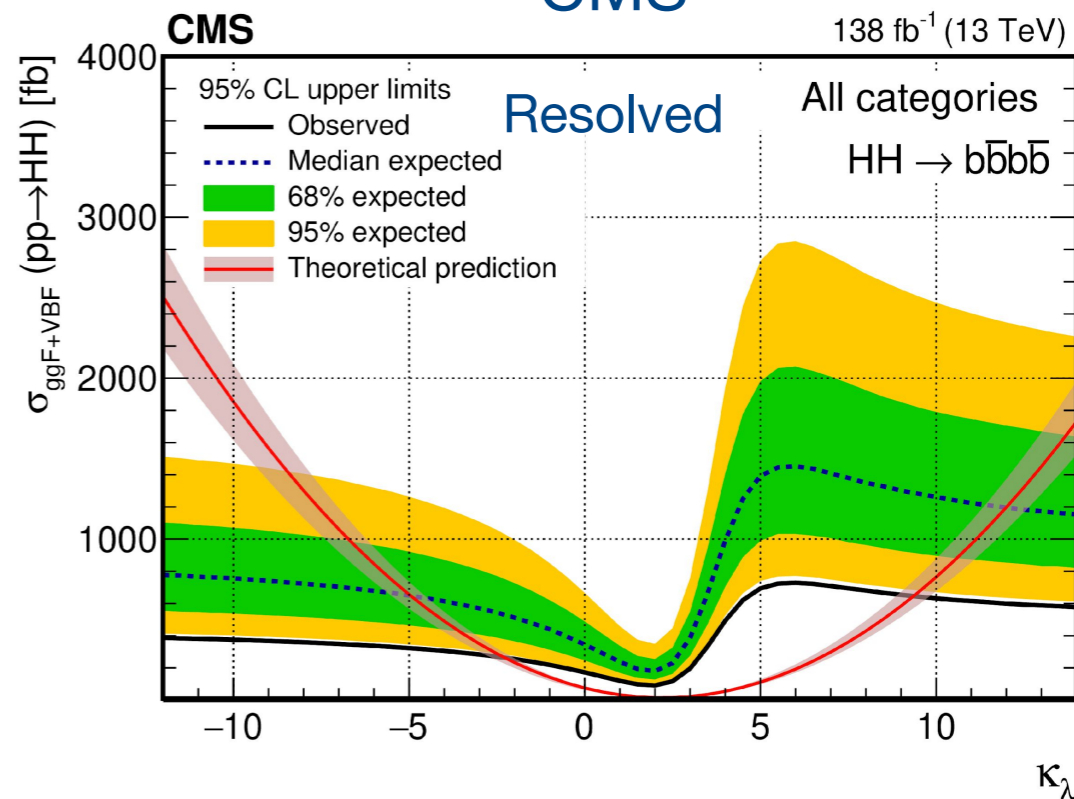
95% CL upper limits are set on the HH production cross section vs κ_λ values

ATLAS



Obs. (exp.) constraint (resolved)
 $[-3.5, 11.3]$ ($[-5.4, 11.4]$)

CMS



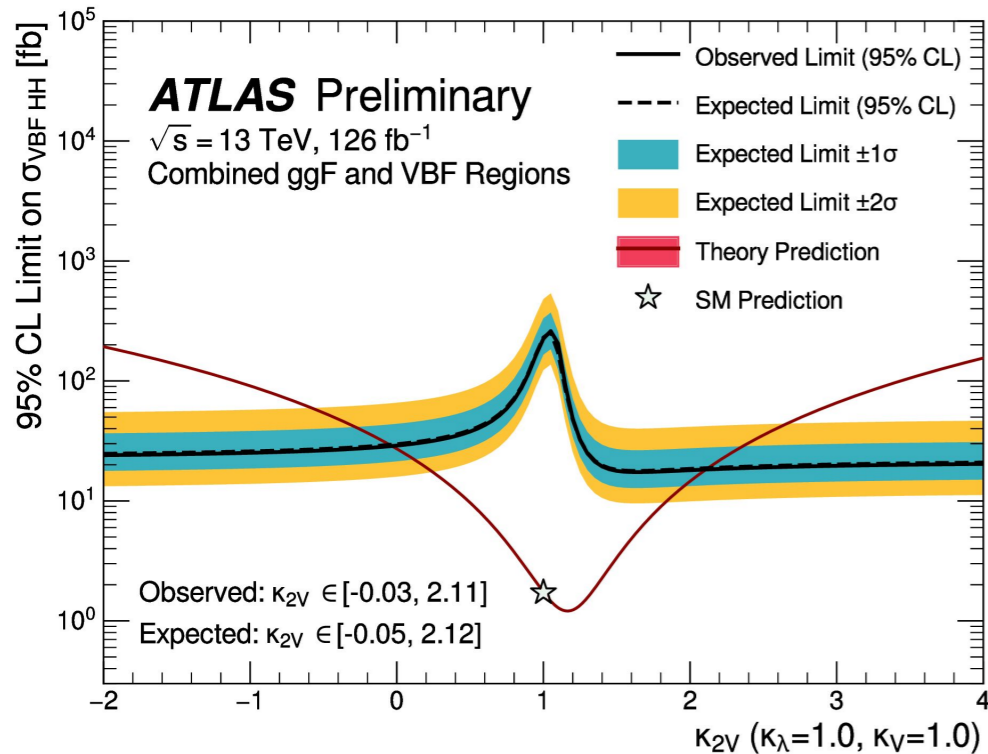
Obs. (exp.) constraint (resolved)
 $[-2.3, 9.4]$ ($[-5.0, 12.0]$)

Obs. (exp.) constraint (boosted)
 $[-9.9, 16.9]$ ($[-5.1, 12.2]$)

Results: Anomalous κ_{2V}

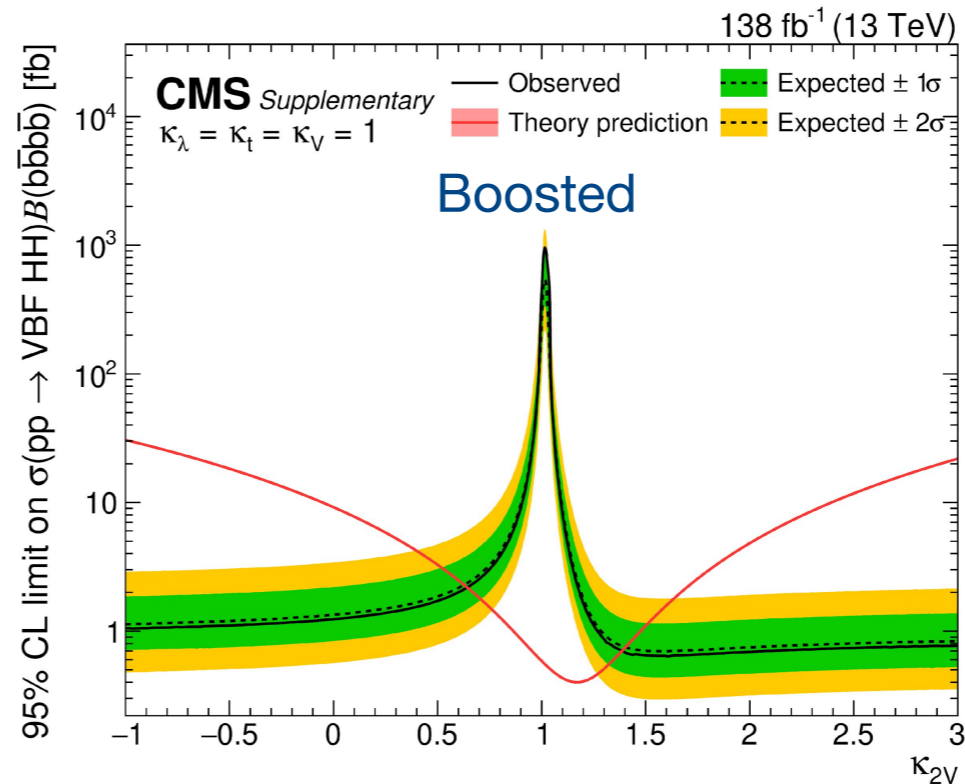
95% CL upper limits are set on the HH production cross section vs κ_{2V} values

ATLAS



Obs. (exp.) constraint (resolved)
 $[0, 2.1] \text{ } ([-0.1, 2.1])$

CMS



Obs. (exp.) constraint (resolved)
 $[-0.1, 2.2] \text{ } ([-0.4, 2.5])$
 Obs. (exp.) constraint (boosted)
 $[0.6, 1.4] \text{ } ([0.7, 1.4])$

The $\kappa_{2V}=0$ hypothesis is excluded for the first time, with a 6.3 s.d. significance ($\kappa_\lambda=\kappa_V=\kappa_t=1$)

The bbbb channel is one of the main players in the HH game

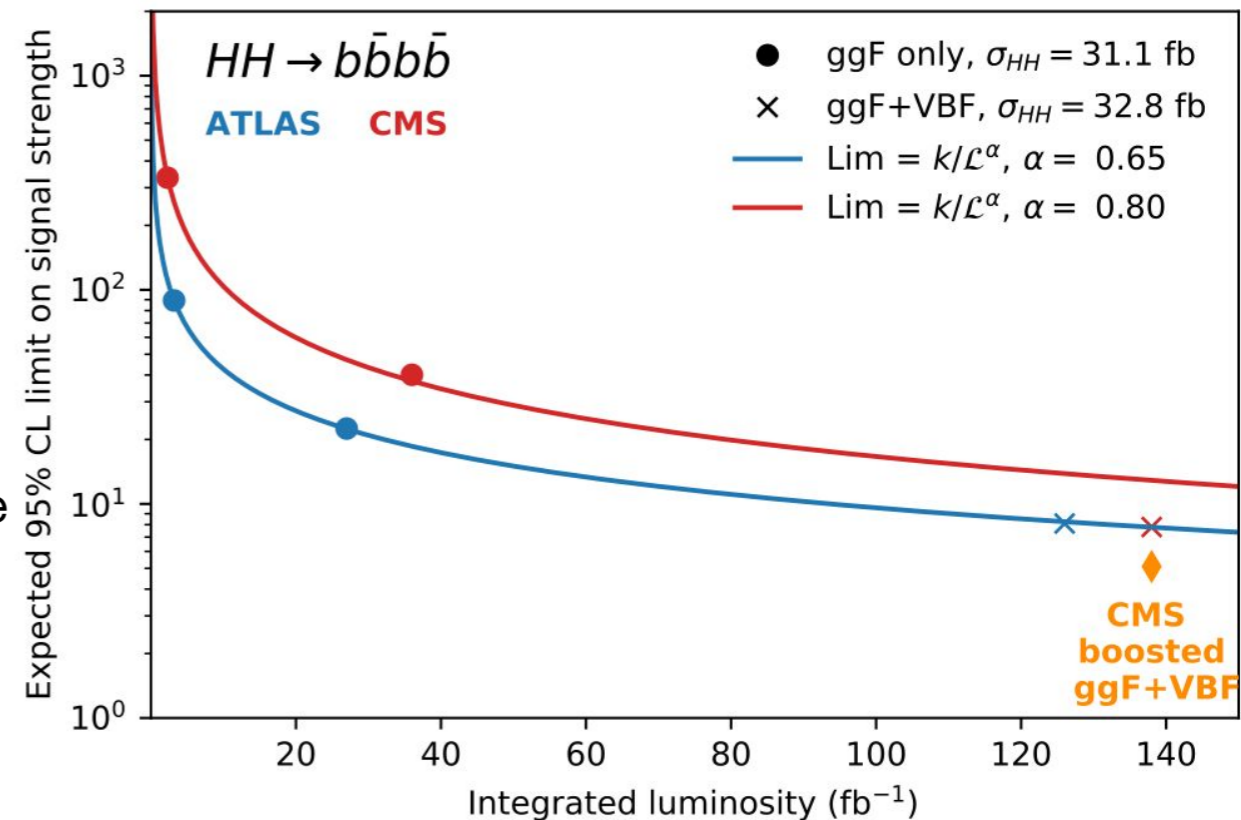
Significant improvements in both ATLAS and CMS Run-2 analyses!

- Sophisticated b-tagging & Higgs-tagging using AI/ML
- New jet pairing methods
- Dedicated ggF and VBF mode categorizations
- Novel background estimation using ML-based reweighting

Looking forward to the future bbbb analyses!

- How can we improve our background estimate?
- How can we make it more sensitive in the low mass region?
- How can we cope with the increasing trigger thresholds in the (HL) LHC's future?

Stay tuned!





CMS Experiment at the LHC, CERN

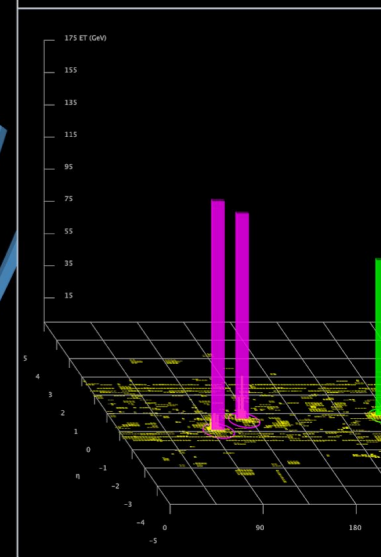
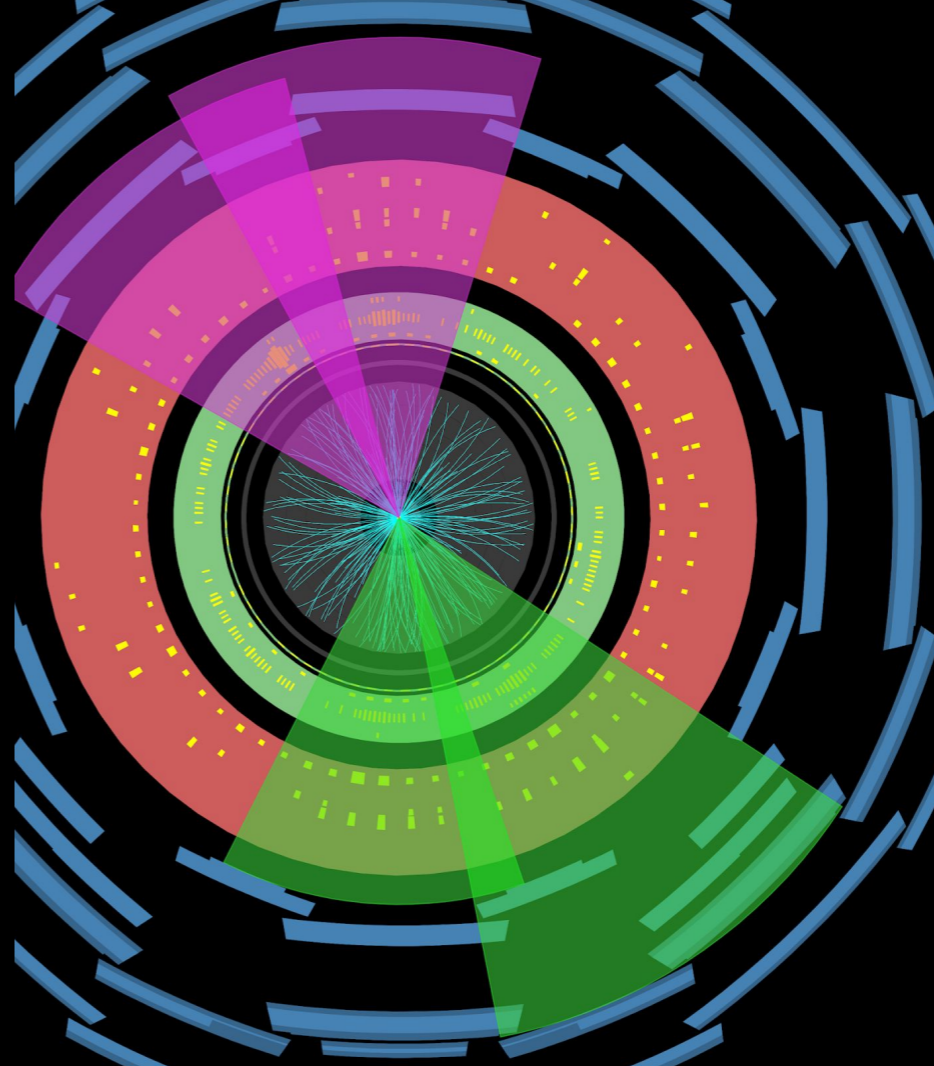
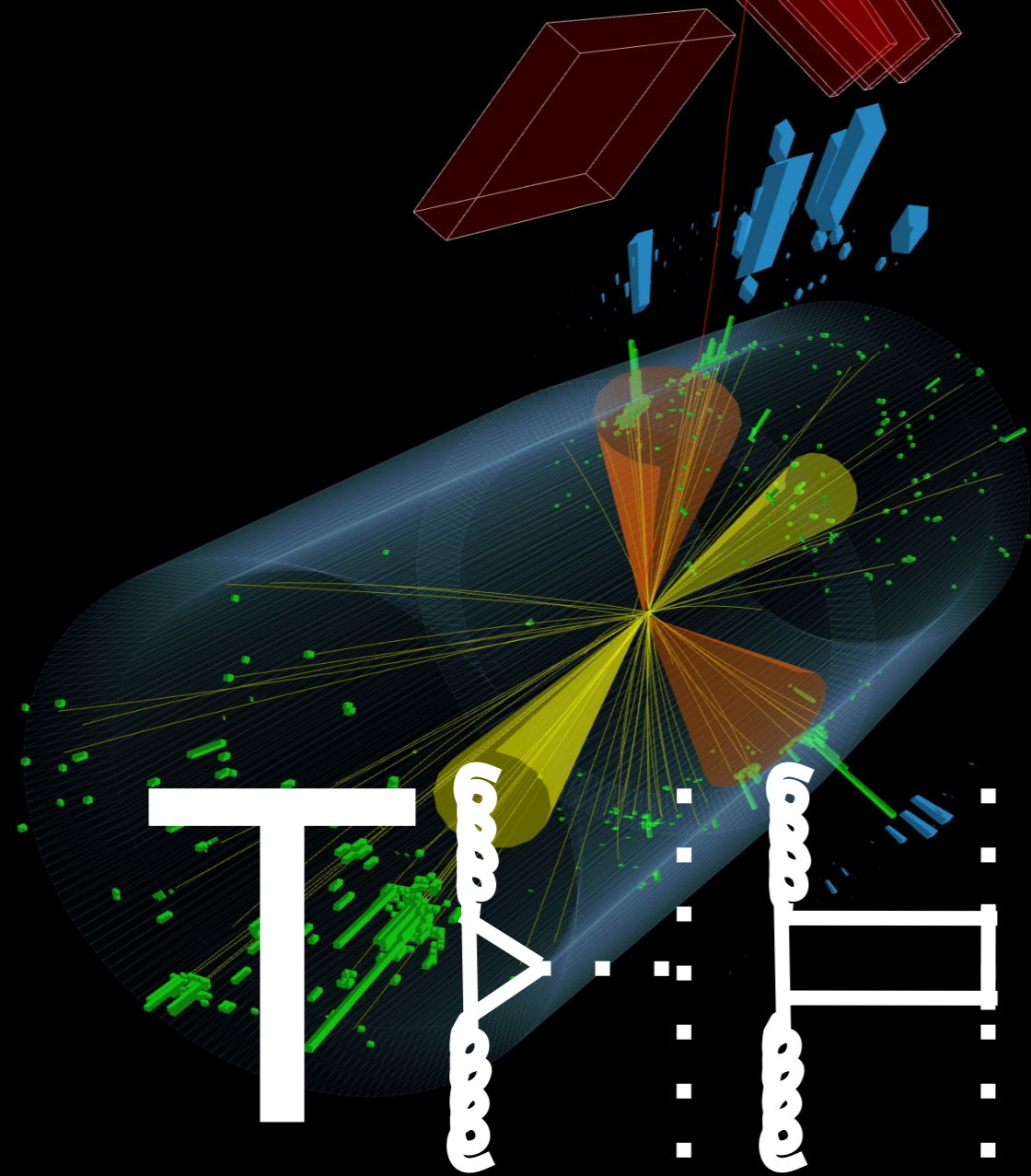
Data recorded: 2017-Aug-07 19:13:22.727552 GMT

Run / Event / LS: 300633 / 525384863 / 347



Run Number: 362619, Event Num

Date: 2018-10-03 17:06



THANKS!

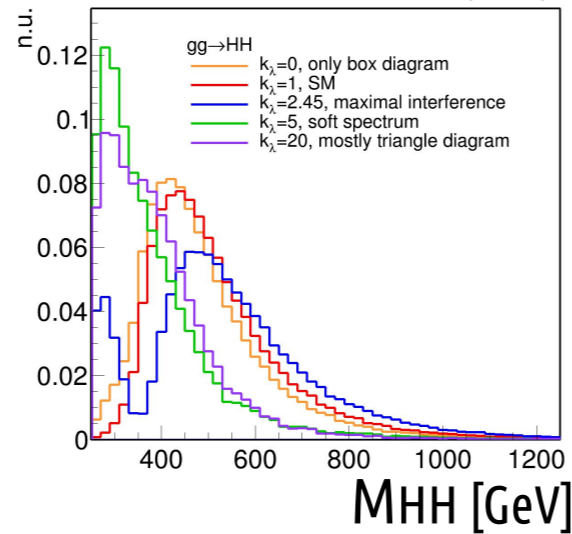
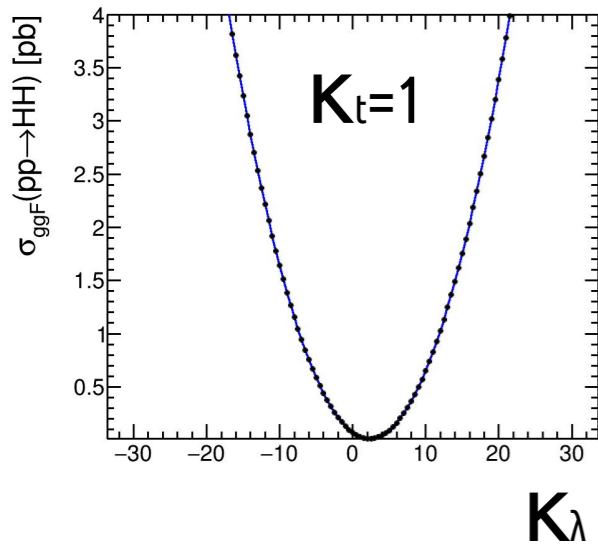
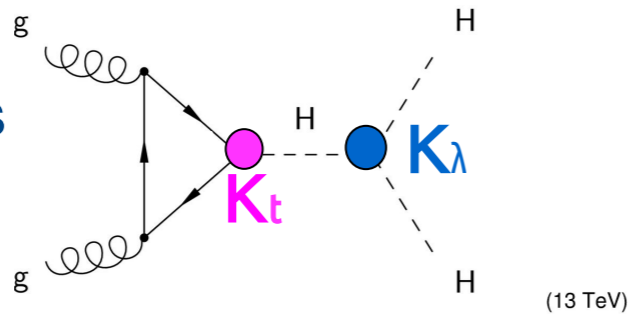
Non-resonant HH production at the LHC

SM production gives access direct access to self-coupling (λ) \rightarrow Reconstruct the Higgs potential
 It's hard to measure it at current LHC datasets (let's see how far we can go!)

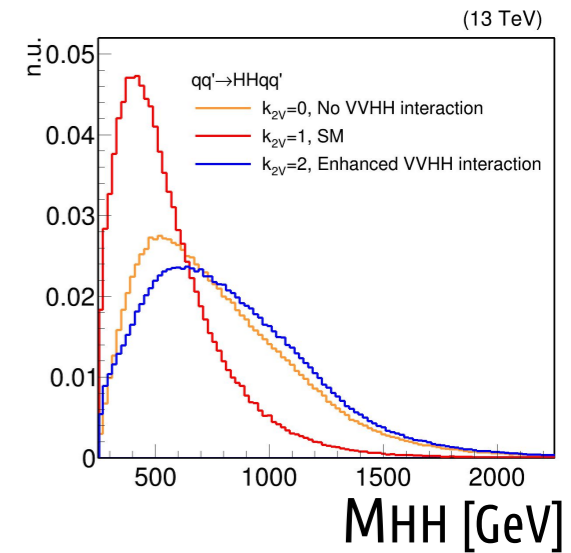
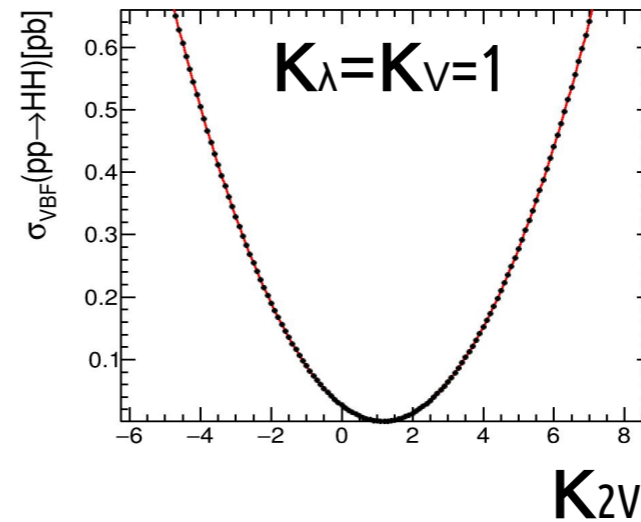
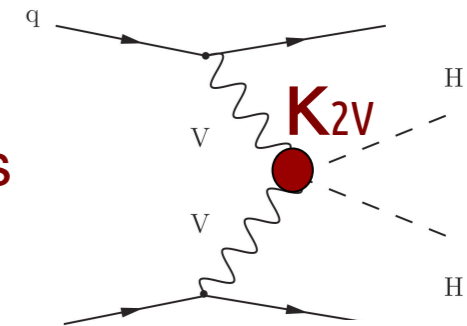
BSM physics may modify Higgs couplings or activate vertices \rightarrow Potential for discovery at the LHC!

- Anomalous Higgs couplings are studied w.r.t. the SM using a K-framework, e.g. $K_\lambda = \lambda / \lambda^{\text{SM}}$
- Large changes in cross sections & kinematics predicted

BSM self-couplings in ggF mode

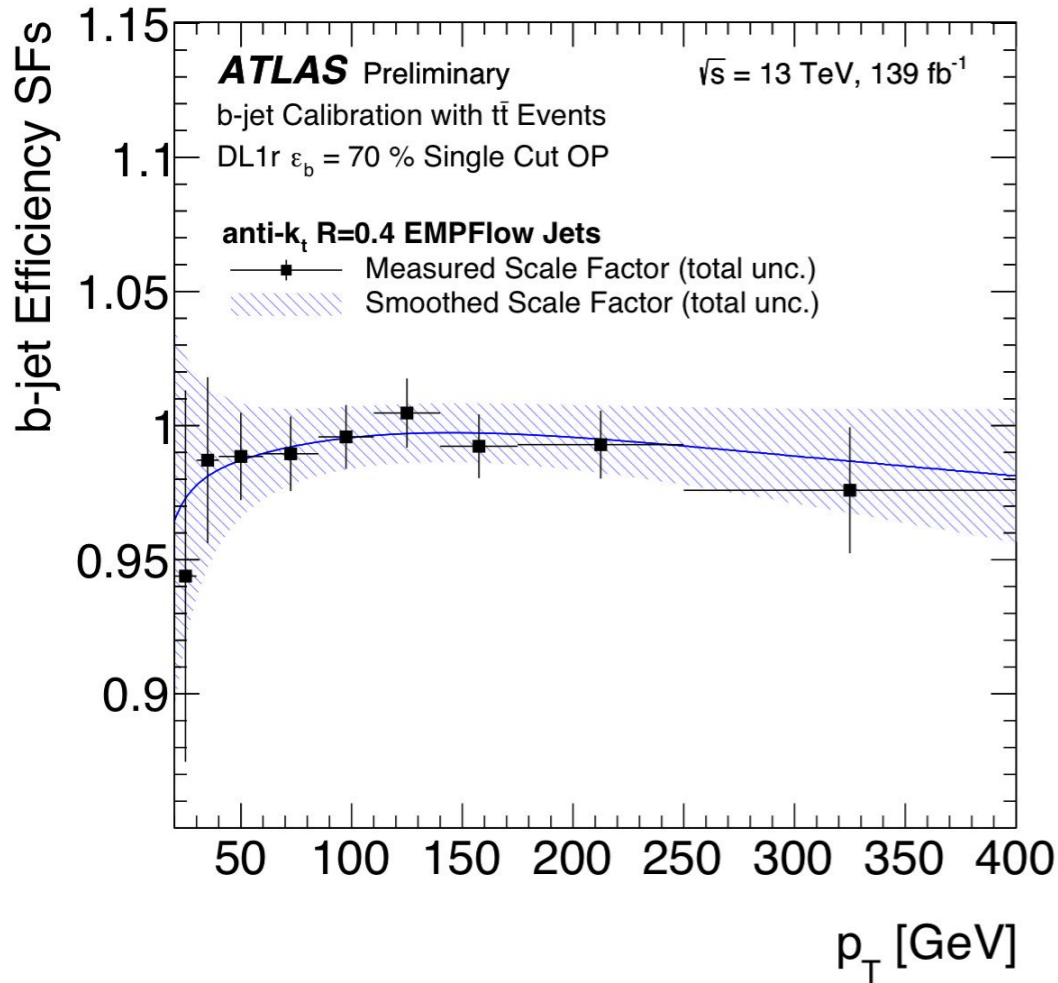


BSM VVHH couplings in VBF mode



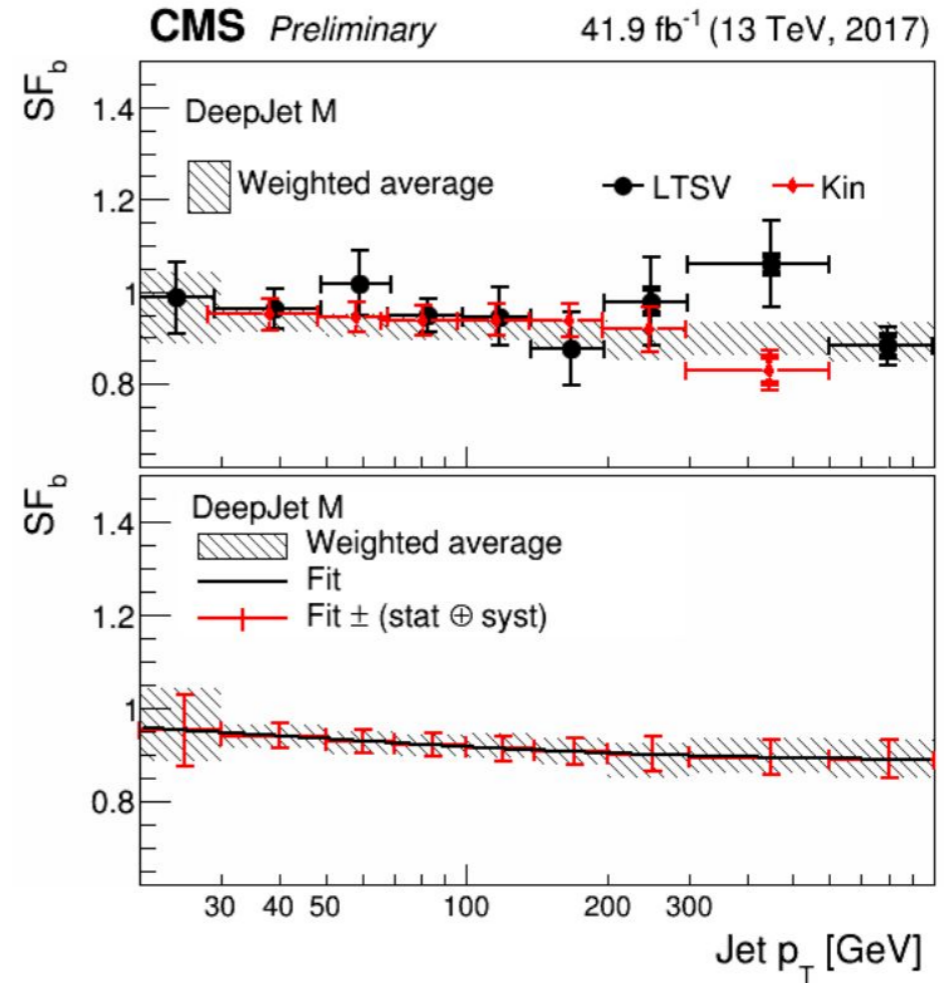
Small-R b-tagging calibration

ATLAS



[DL1r calibration plots](#)

CMS



[DeepJet calibration plots](#)

Production Mode BDT (CMS resolved)

Target: Pre-VBF events

Signal (S): VBF-HH ($k_2 v = 2$)

- Signature with strongest contribution from longitudinal scattering amplitude $V(L)V(L) \rightarrow HH$
- VBF-HH ($k_2 v = 0$) has similar response

Background (B): NLO SM ggF-HH

Variable	Meaning
$p_T(H_1)$ ($p_T(H_2)$)	Transverse momentum of the H_1 (H_2) candidate
$p_T(j_1)$ ($p_T(j_2)$)	Transverse momentum of the j_1 (j_2) candidate
$ \eta(jj) $	VBF-jet pair pseudorapidity
$M(jj)$	VBF-jet pair invariant mass
$\Delta R(H_1, H_2)$	ΔR distance between two Higgs bosons
$\Delta R(H_1, j_1)$	ΔR distance between H_1 and j_1
$\Delta R(H_1, j_2)$	ΔR distance between H_1 and j_2
$\Delta R(H_2, j_1)$	ΔR distance between H_2 and j_1
$\Delta R(H_2, j_2)$	ΔR distance between H_2 and j_2
$ \cos(\theta)^*(j_1) $	$ \cos(\theta) $ of j_1 in the six-jet center of mass frame
$ \cos(\theta)^*(j_2) $	$ \cos(\theta) $ of j_2 in the six-jet center of mass frame
H1-centrality · H2-centrality	Product of the Higgs boson centralities

where: H1-centrality · H2-centrality: $\exp\left[-\left(\frac{\eta(H_1) - \eta_{avg}}{\Delta\eta}\right)^2 - \left(\frac{\eta(H_2) - \eta_{avg}}{\Delta\eta}\right)^2\right]$,

$$\Delta\eta = \eta(j_1) - \eta(j_2) \quad \eta_{avg} = \frac{\eta(j_1) + \eta(j_2)}{2}$$

BDT-reweighting variables (CMS resolved)

ggF categories

BDT Reweigher Input variables

Regressed p_T of the leading- p_T b jet of the H_1 candidate
Regressed p_T of the trailing- p_T b jet of the H_1 candidate
Regressed p_T of the leading- p_T b jet of the H_2 candidate
Regressed p_T of the trailing- p_T b jet of the H_2 candidate
Mass of the H_1 candidate, $M(H_1)$
Mass of the H_2 candidate, $M(H_2)$
Mass of the Higgs pair system, m_{HH}
Transverse momentum of the H_1 candidate, $P_T(H_1)$
Transverse momentum of the H_2 candidate, $P_T(H_2)$
Pseudorapidity separation between the two Higgs candidates, $\Delta\eta(H_1, H_2)$
 ΔR distance between two b jets of the H_1 candidate, $\Delta R(H_1(bb))$
 ΔR distance between two b jets of the H_2 candidate, $\Delta R(H_2(bb))$
 $|\cos(\theta)^*(H)|$ in HH frame
 $|\cos(\theta)^*(b)|$ in H_1 frame
Sum of four b jets' regressed p_T
Transverse momentum of the HH system, $p_T(HH)$
Number of tight b-tags in 3 highest b-tags
Sum of 3b's resolution scores
Minimal ΔR distance between two b jets, $\text{Min}|\Delta R(bb)|$
Maximum pseudorapidity separation between two b jets, $\text{Max}|\Delta\eta(bb)|$

VBF categories

BDT Reweigher Input variables

Regressed p_T of the leading- p_T b jet of the H_1 candidate
Regressed p_T of the trailing- p_T b jet of the H_1 candidate
Regressed p_T of the leading- p_T b jet of the H_2 candidate
Regressed p_T of the trailing- p_T b jet of the H_2 candidate
Mass of the H_1 candidate, $M(H_1)$
Mass of the H_2 candidate, $M(H_2)$
Mass of the Higgs pair system, m_{HH}
Transverse momentum of the H_1 candidate, $P_T(H_1)$
Transverse momentum of the H_2 candidate, $P_T(H_2)$
Pseudorapidity separation between the two Higgs candidates, $\Delta\eta(H_1, H_2)$
Azimuthal angle separation between the two Higgs candidates, $\Delta\phi(H_1, H_2)$
Mass of the VBF-jet pair system, $M(jj)$
Pseudorapidity separation between the two VBF jets, $\Delta\eta(j1, j2)$
PMMVA score

NN-reweighting variables (ATLAS resolved)

ggF	VBF
1. $\log(p_T)$ of the 2 nd leading Higgs boson candidate jet	1. Maximum di-jet mass out of the possible pairings of the four Higgs boson candidate jets
2. $\log(p_T)$ of the 4 th leading Higgs boson candidate jet	2. Minimum di-jet mass out of the possible pairings of the four Higgs boson candidate jets
3. $\log(\Delta R)$ between the closest two Higgs boson candidate jets	3. Energy of the leading Higgs boson candidate
4. $\log(\Delta R)$ between the other two Higgs boson candidate jets	4. Energy of the subleading Higgs boson candidate
5. Average absolute η value of the Higgs boson candidate jets	5. Second smallest ΔR between the jets in the leading Higgs boson candidate (out of the three possible pairings for the leading Higgs candidate)
6. $\log(p_T)$ of the di-Higgs system	6. Average absolute η value of Higgs boson candidate jets
7. ΔR between the two Higgs boson candidates	7. $\log(X_{Wt})$
8. $\Delta\phi$ between jets in the leading Higgs boson candidate	8. Trigger class index as one-hot encoder
9. $\Delta\phi$ between jets in the subleading Higgs boson candidate	9. Year index as one-hot encoder (for years inclusive training)
10. $\log(X_{Wt})$	
11. Number of jets in the event	
12. Trigger class index as one-hot encoder	

Discriminant observables (CMS resolved)

Chosen observables maximize the analysis sensitivity
 Their definition is constrained by available statistics of the bkg model

ggF categories: BDT distribution

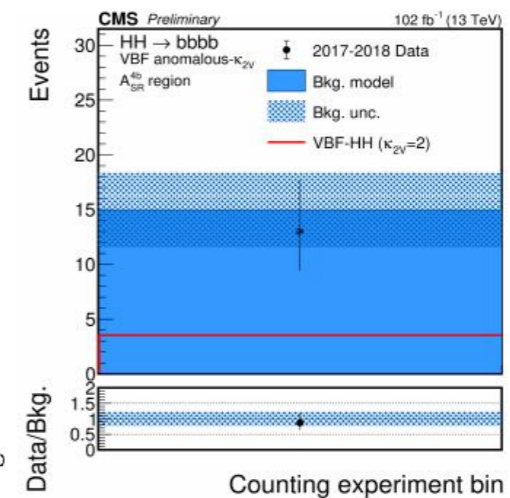
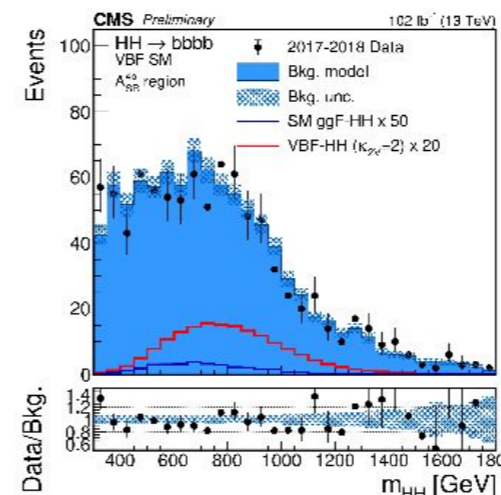
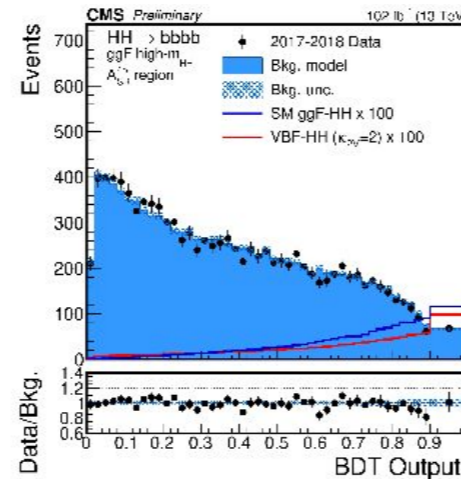
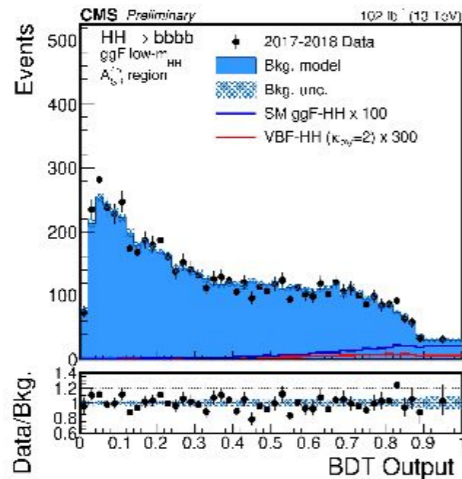
- Bkg model enables to model a BDT discriminant
- Discriminant is trained by category
 - Signal (SM ggF) vs background (bkg model)
 - 16 variables

VBF category 1,2:

- mHH distribution, counting experiment

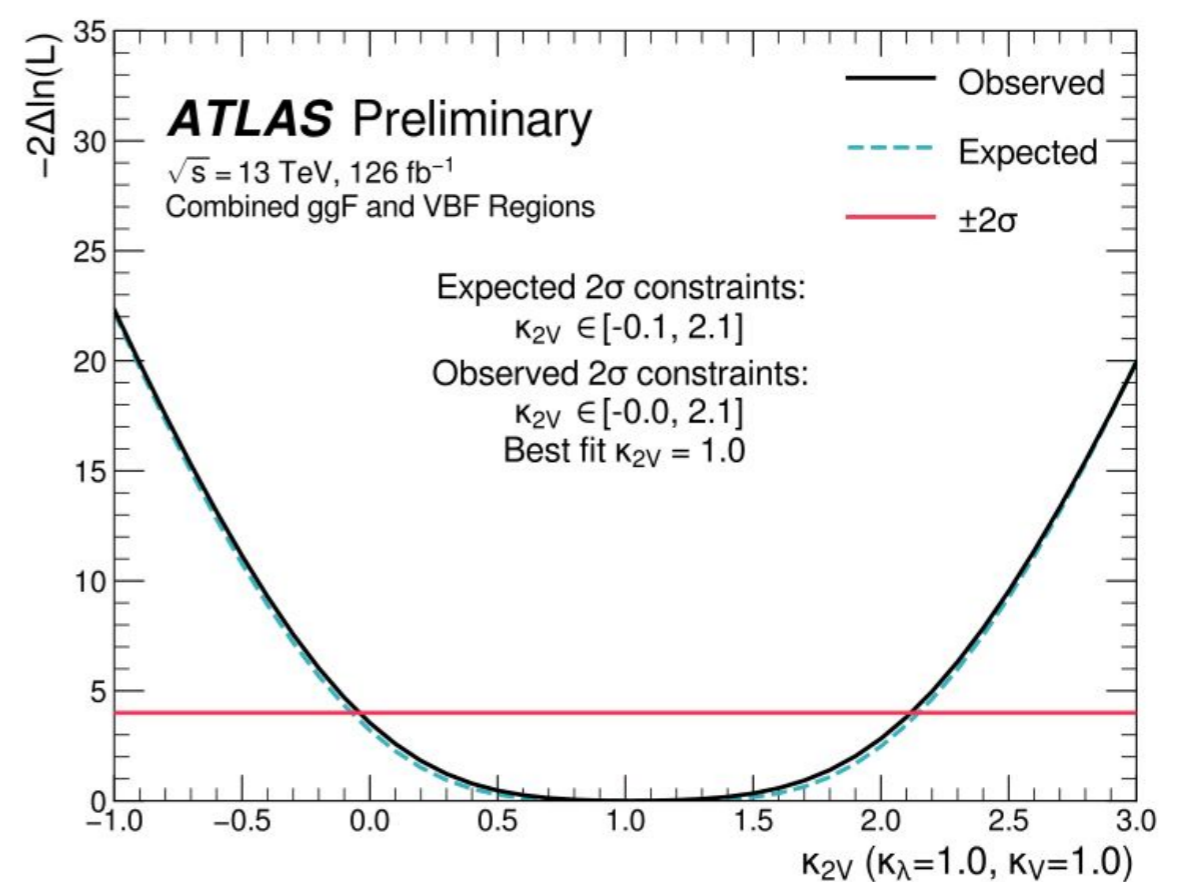
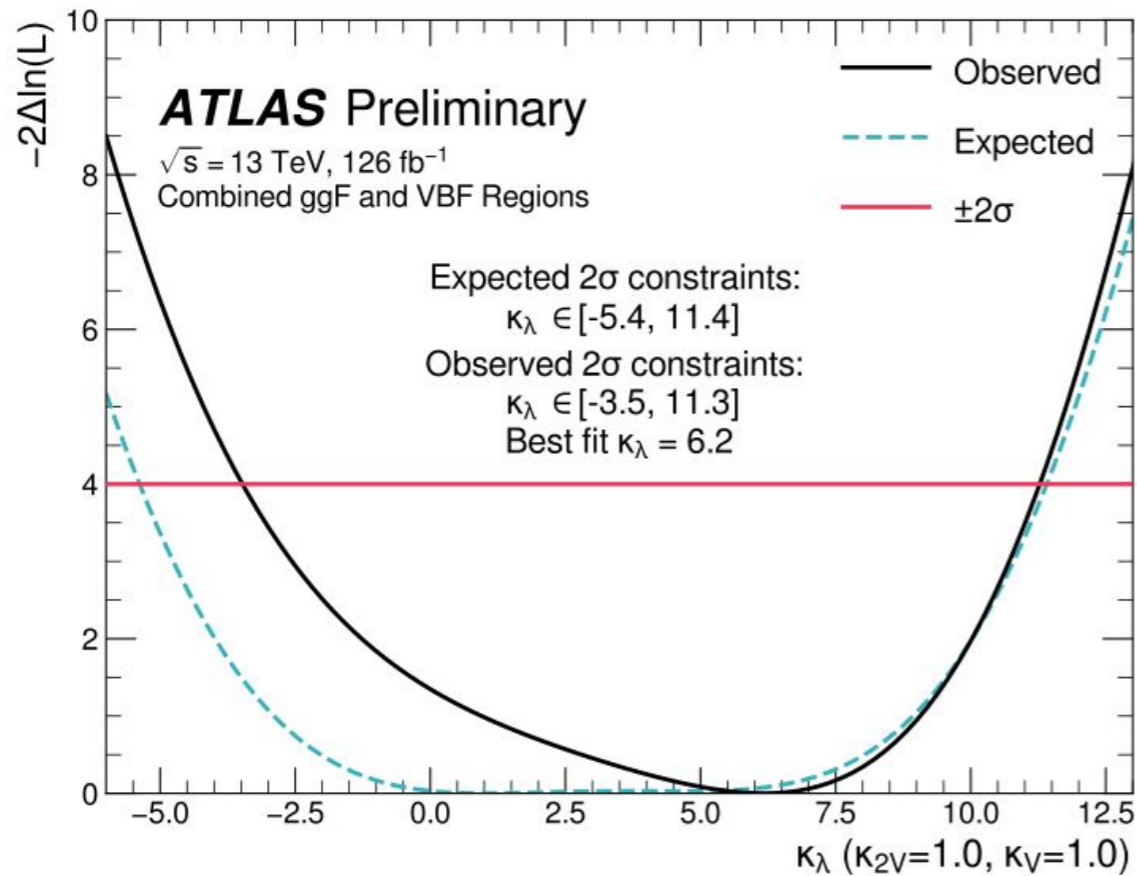
BDT input variables	
M(H1), M(H2), M(HH)	$ \cos^*\theta(b) $ in H1-frame
$p_T(H1), p_T(H2), p_T(HH)$	Max $ \Delta\eta_{bb} $
Scalar sum of 4b's p_T	Min $ \Delta R_{bb} $
$\Delta\eta(H1,H2), \Delta R_{bb}(H1), \Delta R_{bb}(H2)$	Sum of 3b's resolution scores
$ \cos^*\theta(H) $ in HH-frame	N. tight b-tags in 3 highest b-tags

2017-2018
 Observables
 (post-fit)

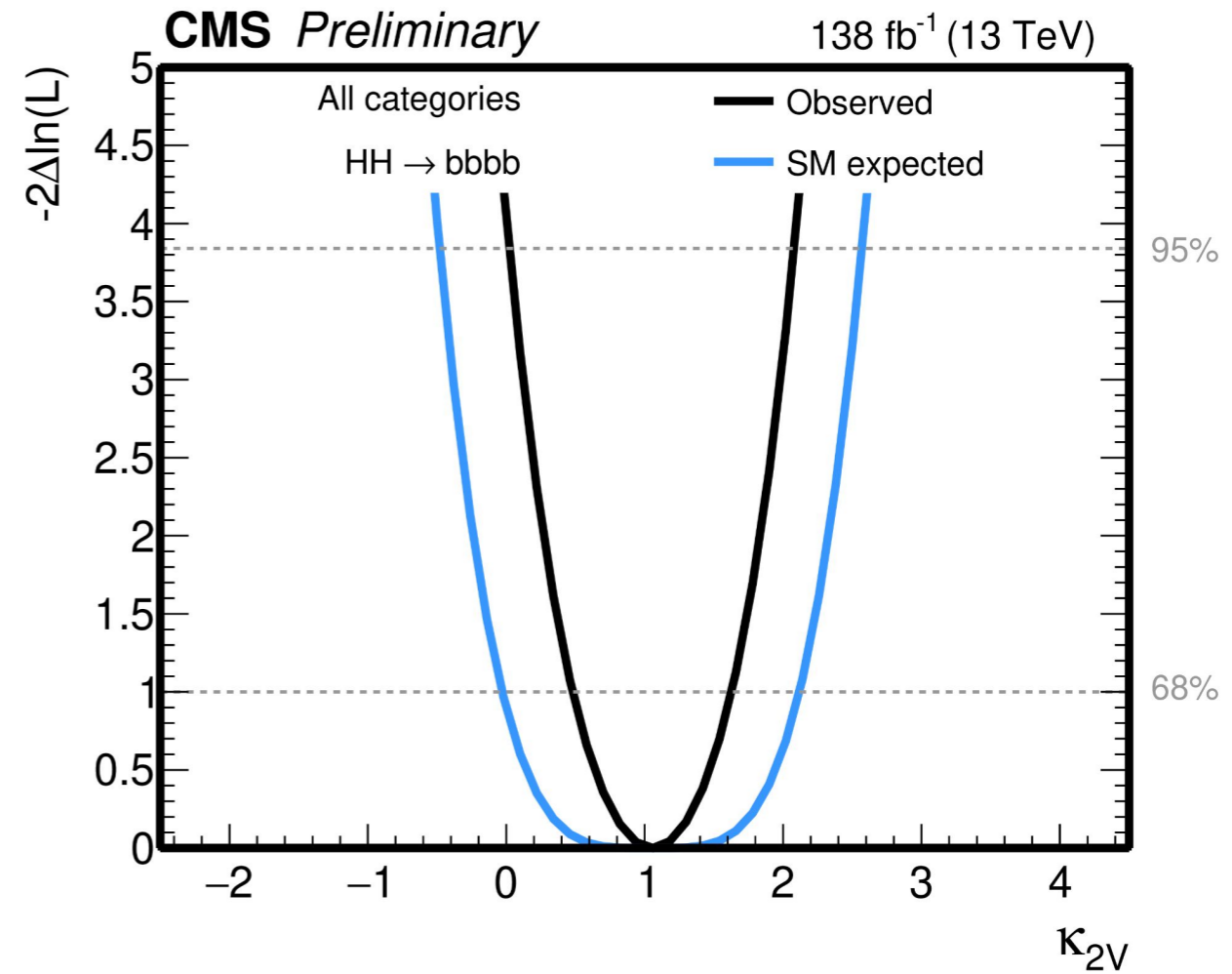
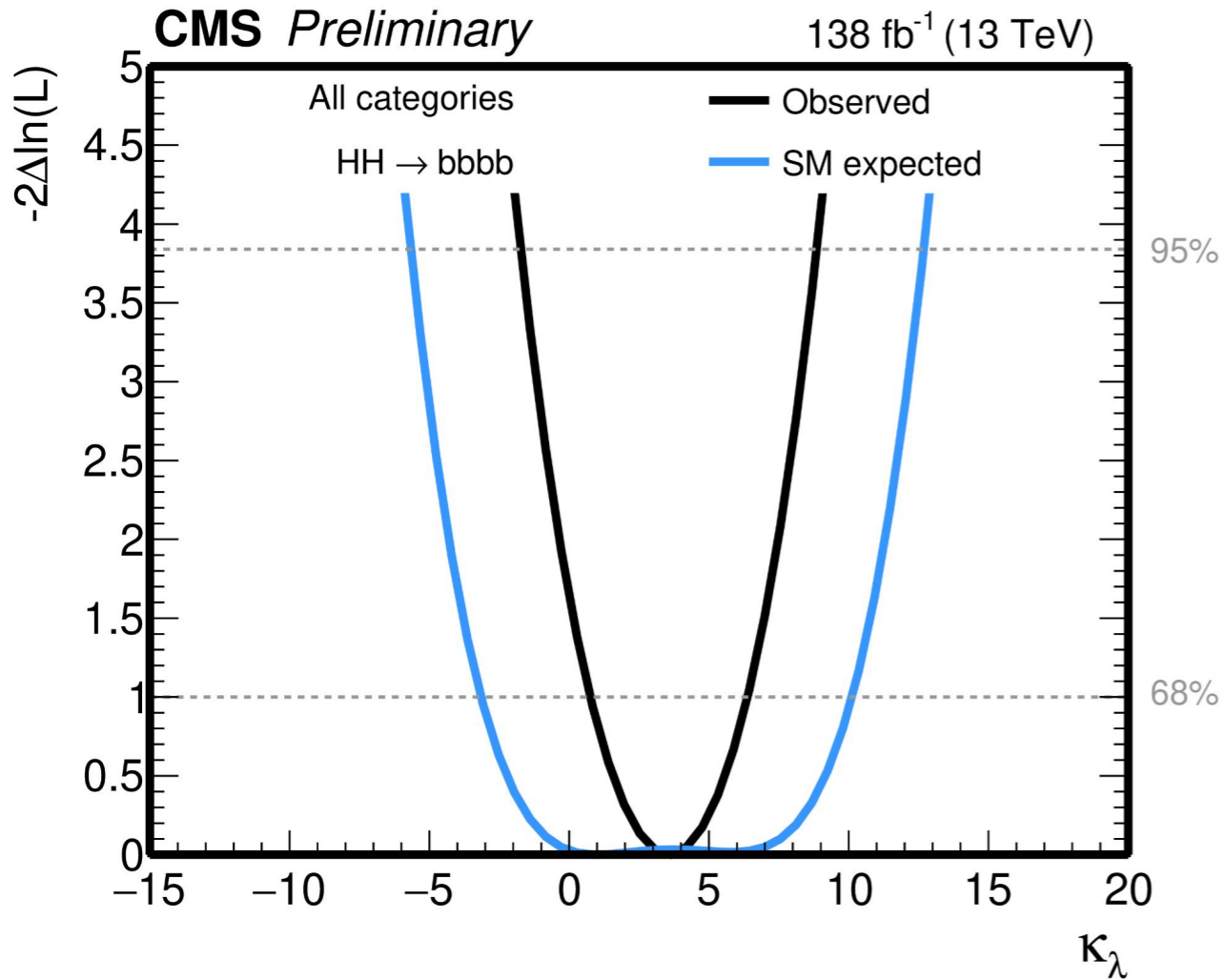
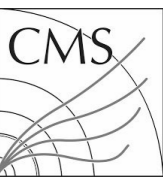


Goodness of fit tests showed that data & bkg model are compatible in all eight observables ($p > 23\%$)

Likelihood scans (ATLAS resolved)



Likelihood scans (CMS resolved)



Likelihood scans (CMS boosted)

