

Search for Higgs boson decay to a charm quark-antiquark pair in proton-proton collisions at $\sqrt{s} = 13$ TeV at CMS

Based on [arXiv:2205.05550](https://arxiv.org/abs/2205.05550), [CMS-PAS-HIG-21-008](https://cds.cern.ch/record/2940313)

Congqiao Li (李聰喬), Peking University

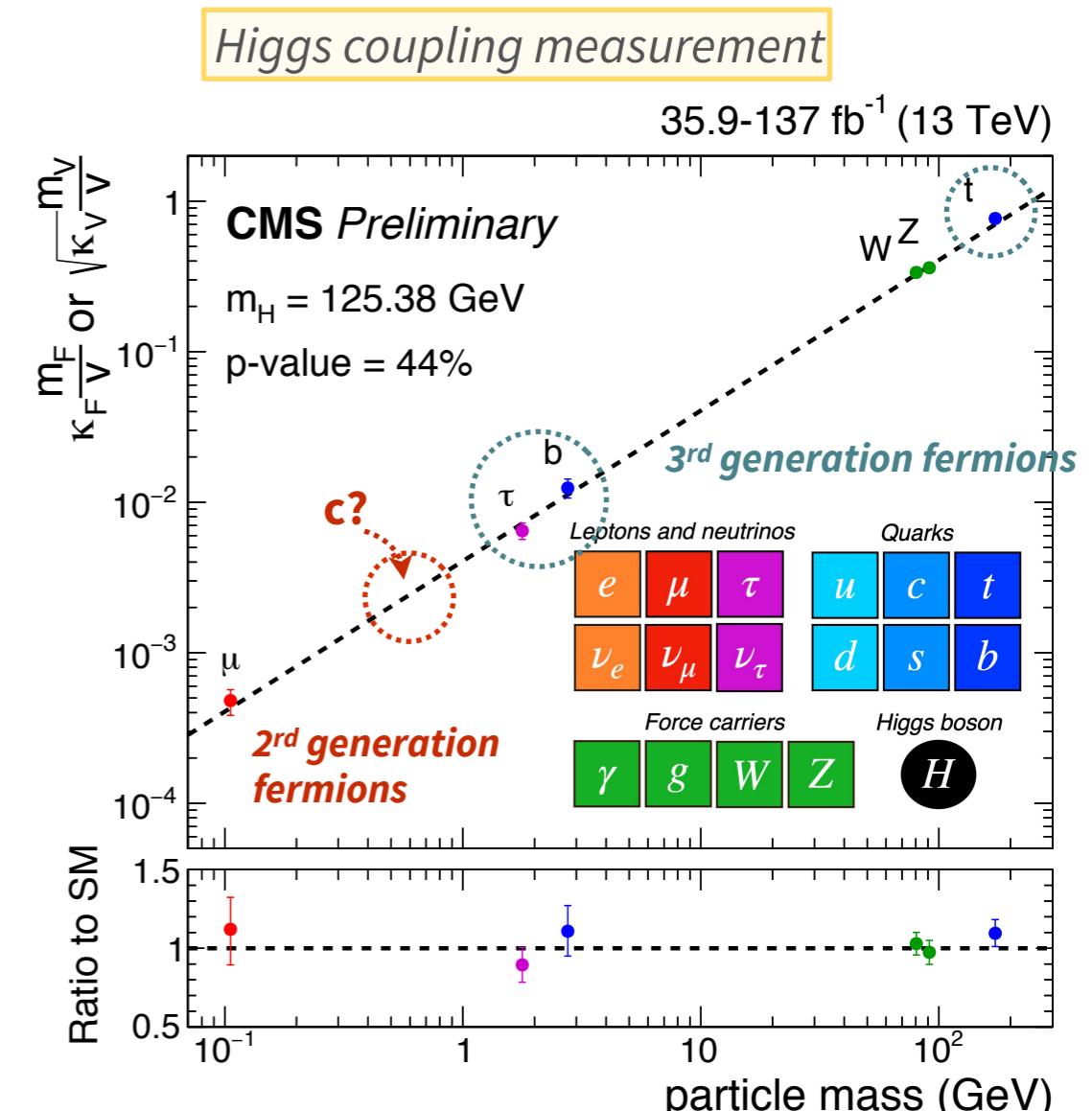
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10 Aug, 2022



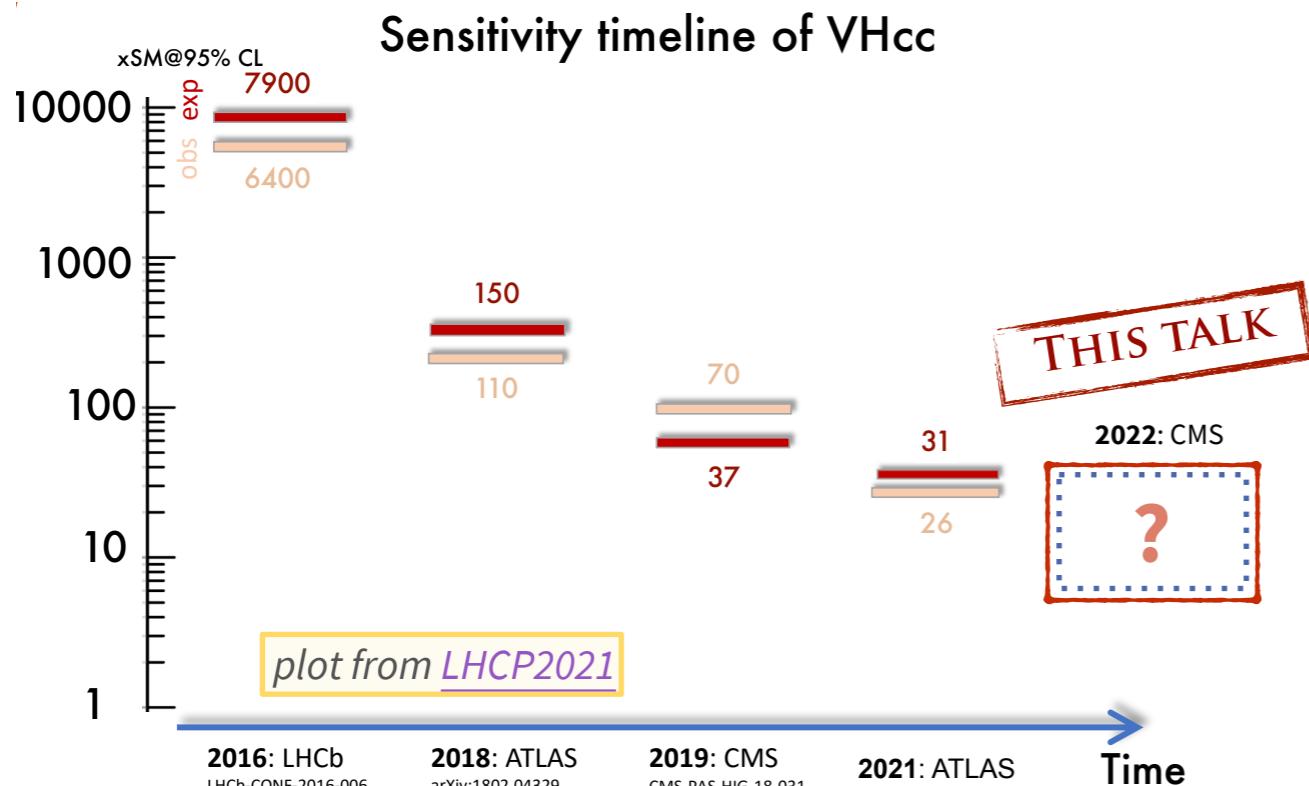
Motivation

- Measurement of the Higgs boson properties is of vital importance since its discovery
 - ❖ couplings with 3rd generation fermions ($t/b/\tau$) well-measured in high precision
- Next milestone: measure Yukawa couplings with 2nd generation fermions
 - ❖ **Higgs-charm** coupling can be modified in some BSM scenarios → essential to verify SM prediction and find possible hints to BSM
 - ❖ direct approach: **measure Higgs-charm coupling via $H \rightarrow cc$ decay channel**



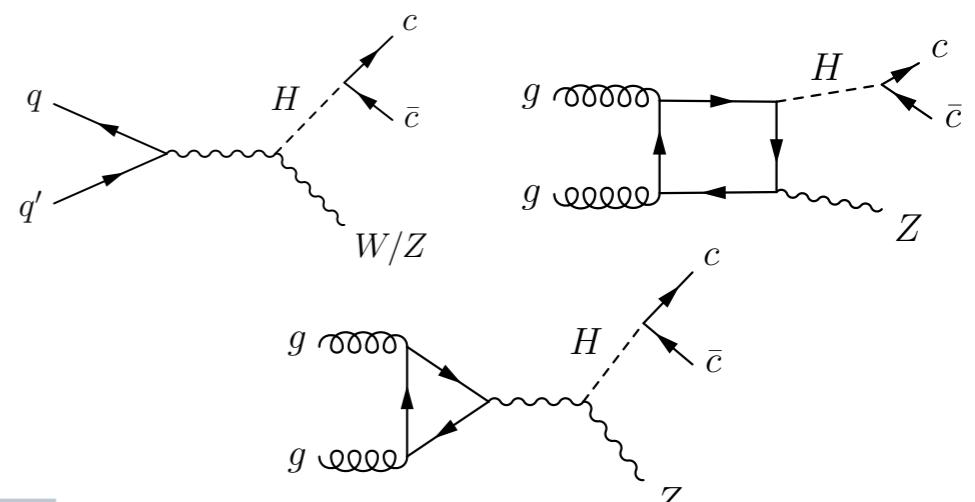


Journey of VHcc search

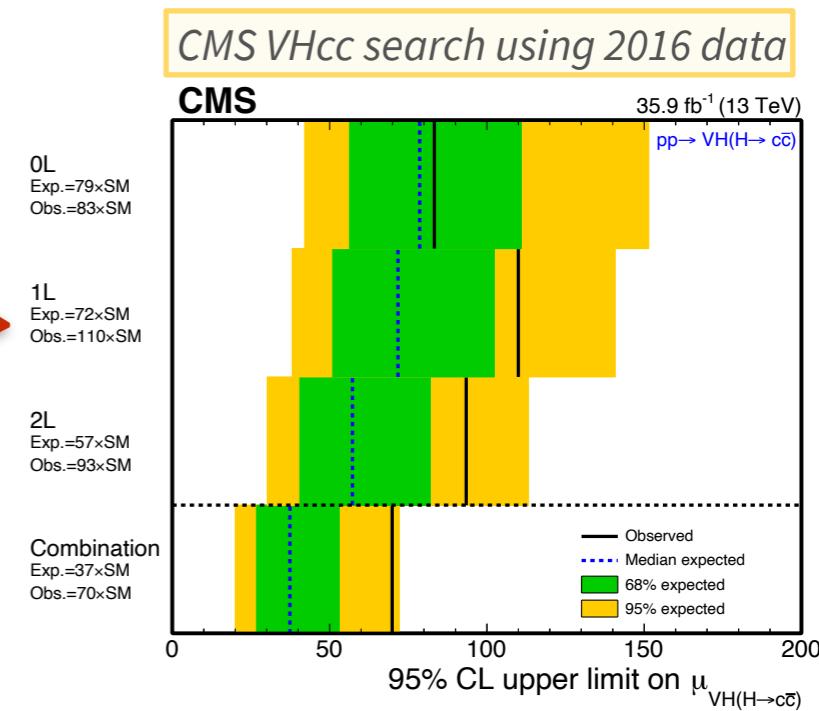


→ **VH($\rightarrow cc$) mode** with $V = W(\rightarrow \ell\nu)$ or $Z(\rightarrow \ell\ell/\nu\nu)$ is most sensitive channel by far

- ❖ leptonic decay of W/Z provides a handle to suppress multijet background

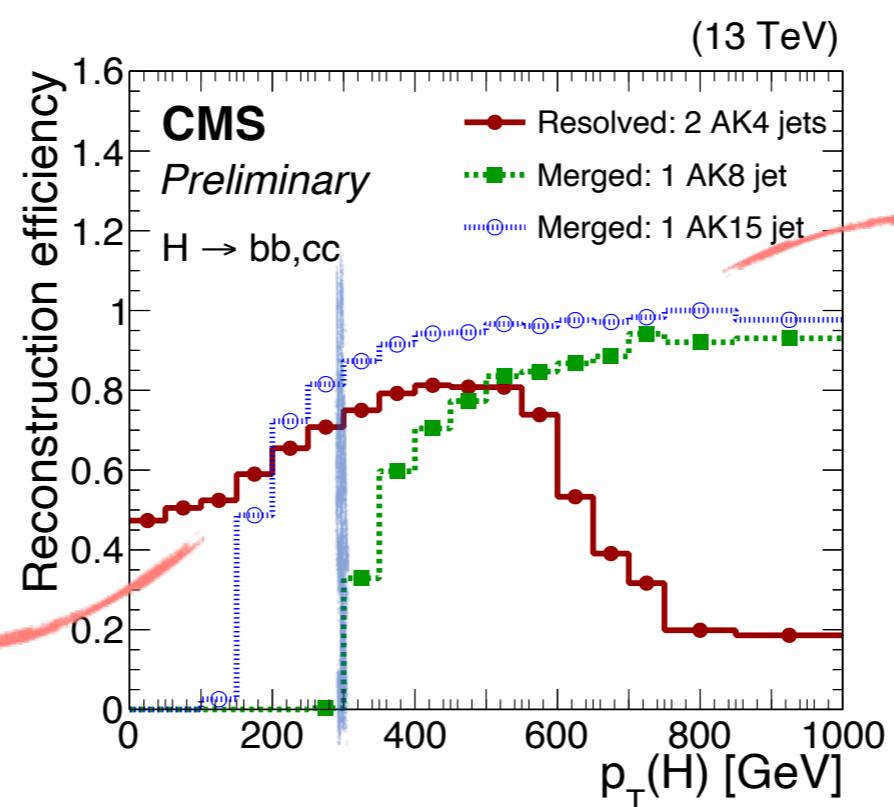
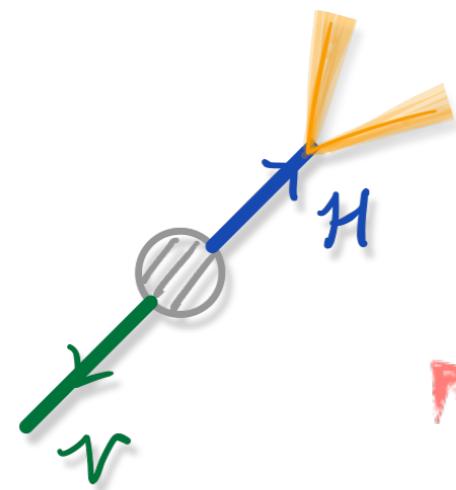


OBS (EXP) UL $\sigma(VH_{cc})$	OBS (EXP) UL on κ_c	$\int L dt$	
$\mu < 110$ (150)	-	36.1 fb^{-1} (2016)	ATLAS [PRL 120 (2018) 211802] $Z(\rightarrow \ell\ell)H$
$\mu < 70$ (37)	-	35.9 fb^{-1} (2016)	CMS [JHEP 03 (2020) 131] $V(Z \rightarrow \ell\ell, Z \rightarrow \nu\nu, W \rightarrow \ell\nu)H$
$\mu < 26$ (31)	$ \kappa_c < 8.5$ (12.4)	139 fb^{-1} (full Run 2)	ATLAS [HIGG-2021-12] $V(Z \rightarrow \ell\ell, Z \rightarrow \nu\nu, W \rightarrow \ell\nu)H$
?	?	138 fb^{-1} (full Run 2)	CMS [arXiv:2205.05550 (accepted by PRL)] [CMS-PAS-HIG-21-008] $V(Z \rightarrow \ell\ell, Z \rightarrow \nu\nu, W \rightarrow \ell\nu)H$

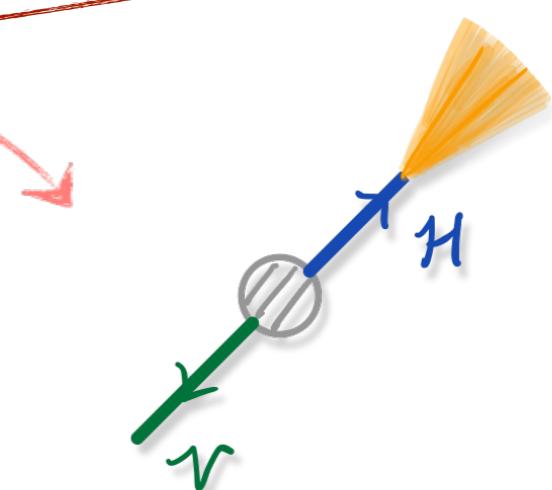




VHcc analysis in a nutshell



NEW PHASE-SPACE
EXPLORED IN CMS



$$\Delta R(c, c) \sim 2m(H)/p_T(H)$$

Resolved-jet topology

- ❖ reconstruct $H \rightarrow cc$ decay with two $R=0.4$ jets
- ❖ apply charm tagging (DeepJet) for each jet
- ❖ able to probe larger fractions of the available phase space ($\sim 95\%$)

Merged-jet topology

- ❖ reconstruct $H \rightarrow cc$ decay with one large- R ($R=1.5$) jets
- ❖ apply di-charm tagging (ParticleNet X $\rightarrow cc$) to the jet; able to exploit the correlations of two charms
- ❖ improved signal purity with large $p_T(H)$



Merged-jet: $H \rightarrow cc$ identification with ParticleNet

- Merged jet topology: Higgs candidate reconstructed by a single large- R jet
 - ❖ using advanced deep learning approach to tag the large- R jet with a $H \rightarrow cc$ signature
 - ❖ major improvement has been made in improving the tagger design

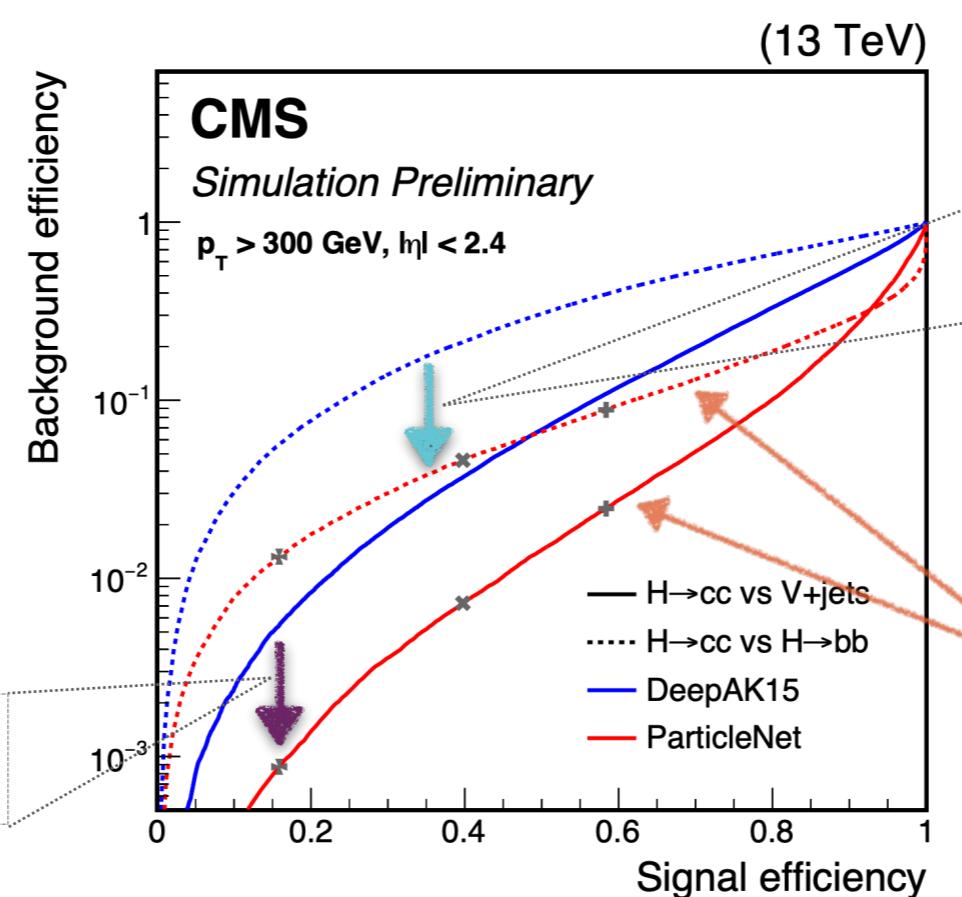
[JHEP 03 \(2020\) 131](#) (2016 analysis)

[DeepAK8 \(DeepAK15\)](#) [JINST 15 (2020)]

P06005]

- ❖ multi-class DNN classifier with 1D convolutional neural network
- ❖ directly uses jet constituents (particle-flow candidates / secondary vertices)
- ❖ mass decorrelation via adversarial training

~5x better
V+jet rejection



~5x better
 $H \rightarrow bb$ rejection

[CMS-PAS-HIG-21-008](#) (Run 2 analysis)

[ParticleNet](#) [CMS-DP-2020-002]

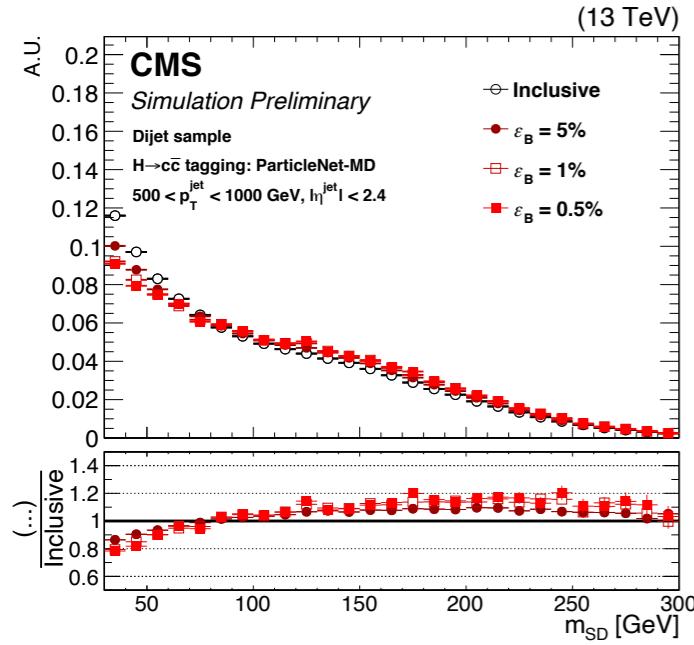
- ❖ same spirit as DeepAK8, but substantially improved:
- ❖ **graph neural network architecture**
- ❖ **novel mass decorrelation technique**
- details in the next slide



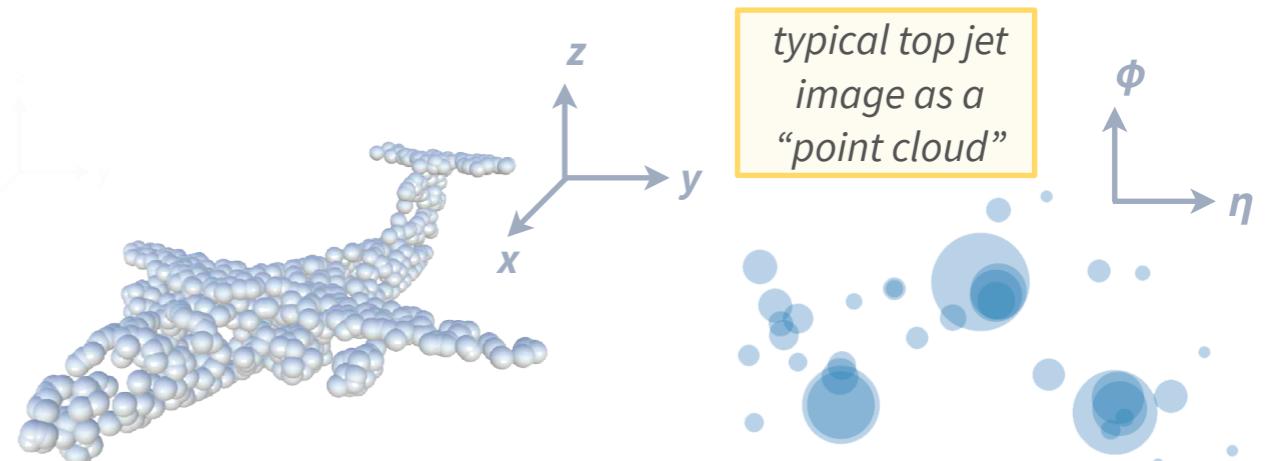
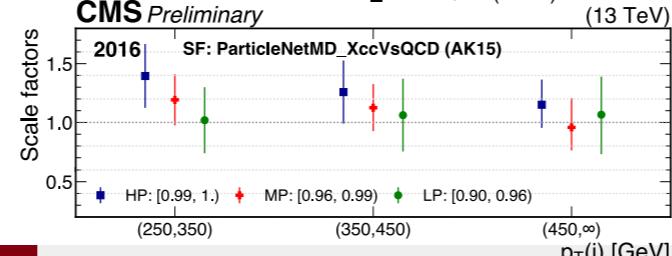
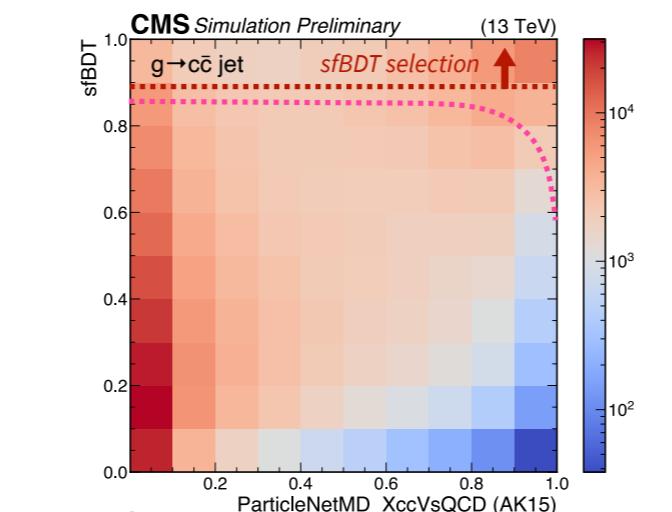
Merged-jet: ParticleNet

- **ParticleNet**: an advanced multi-class jet classifier with a graph neural networks (**GNN**) structure, processing input features on a “**point cloud**” [[CMS-DP-2020-002](#)]
- ❖ **point cloud**: an unordered set of particles
- ❖ **GNN**: apply the EdgeConv operation on the point cloud
- ❖ able to use **low-level jet features** as inputs (particle-flow candidates & secondary vertices)

ParticleNet mass-decorrelation effect



tagger calibration: use BDT to select a phase-space from $g \rightarrow cc$



→ Details and techniques:

- ❖ **categorisation**: resonance decaying to different flavour contents & QCD:
 - ▶ $X \rightarrow bb, X \rightarrow cc, X \rightarrow qq, QCD$ ($bb, cc, b, c, others$)
- ❖ **mass decorrelation** (decorrelate with Higgs mass): signal & QCD reweighted on a flat (p_T , mass) distribution before training
- ❖ **novel calibration method**: select a phase-space from $g \rightarrow cc$ that resembles $H \rightarrow cc$ jets, by vetoing jets with large gluon contamination using a BDT [[CMS-DP-2022-005](#)]



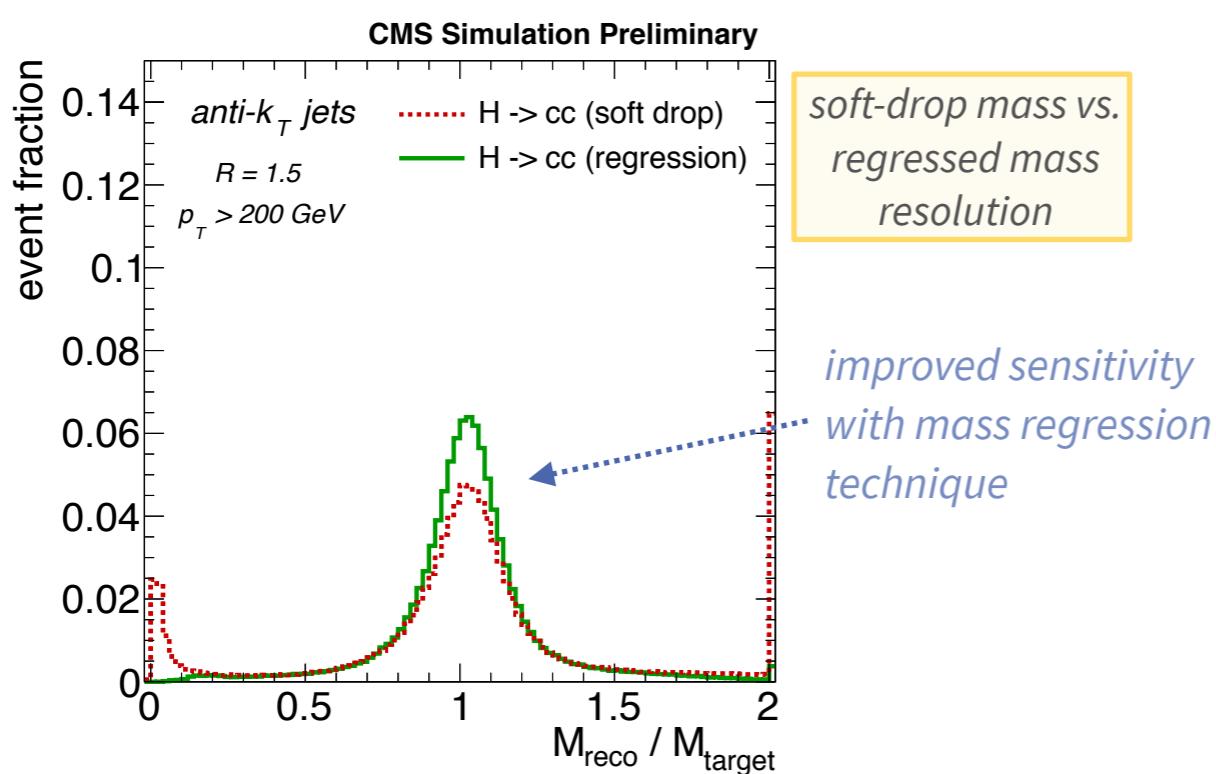
Merged-jet: analysis techniques & strategy

→ ParticleNet for cc-tagging on large- R ($R=1.5$) jets

- ✓ **huge improvement:** ~5 times increase in BKG rejection w.r.t. DeepAK15 tagger (2016) [[CMS-DP-2020-002](#)]
- ✓ tagger calibration with novel BDT approach [[CMS-DP-2022-005](#)]

→ Di-charm jet mass regression

- ❖ use the ParticleNet architecture with a regression layer [[CMS-DP-2021-017](#)]
- ❖ ~20–25% improvement in exp. upper limit, w.r.t. using soft-drop mass



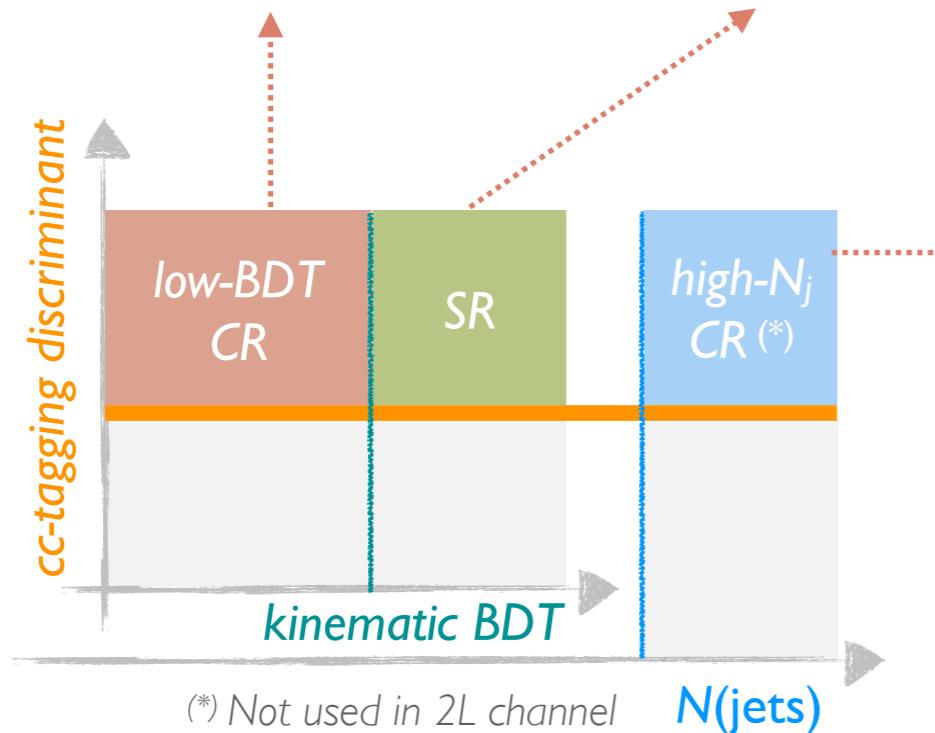
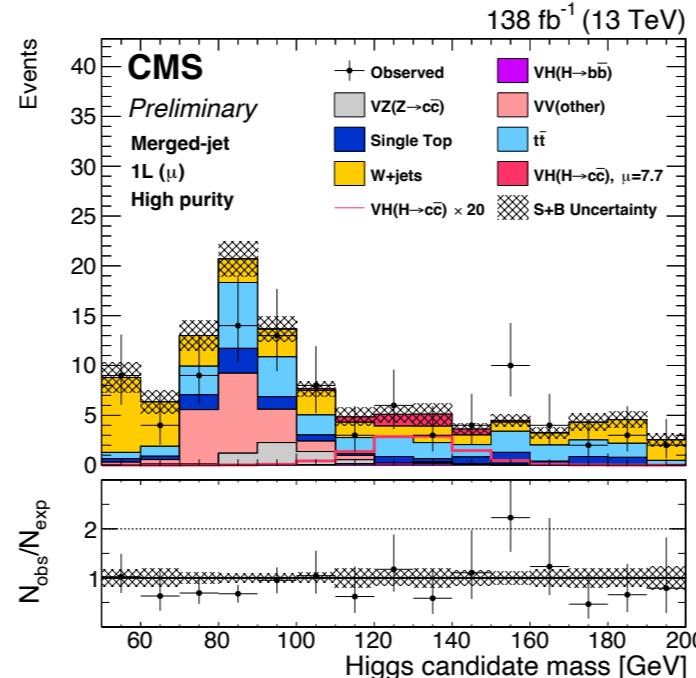
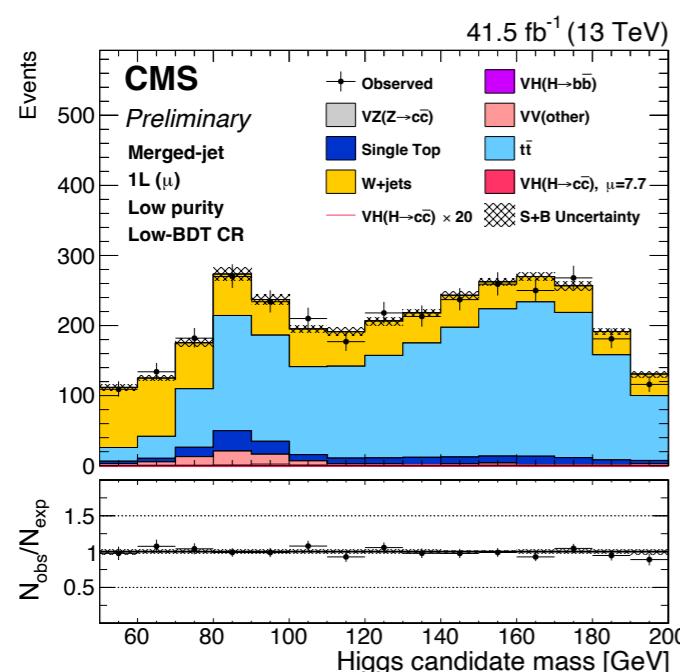
Analysis strategy

- train **kinematics BDTs** to separate VHcc signal vs. BKG
 - ❖ using only kinematics properties - no mass/flavour information
 - ❖ goal: **decorrelated BDT with H_{cand} mass and ParticleNet cc-tagging discriminant**
- CR defined by inverting the BDT cut
 - ❖ also define high N_j region for t \bar{t}
- apply **cc-tagging cut on three WPs**
- finally extract VHcc by **fitting on the regressed mass** variable



Merged-jet: CRs and SR

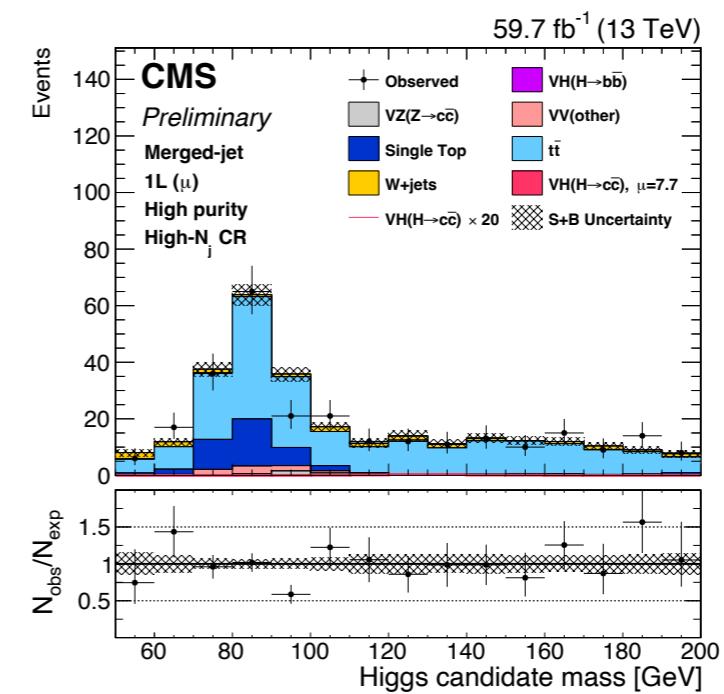
example post-fit plots for CR and SR, for 1L $W(\mu\nu)$



SR and CRs definition

→ Simultaneous fit of all SR and CRs

- ❖ extract BKG normalisation factors assigned to $V+jets$ and $t\bar{t}$ components

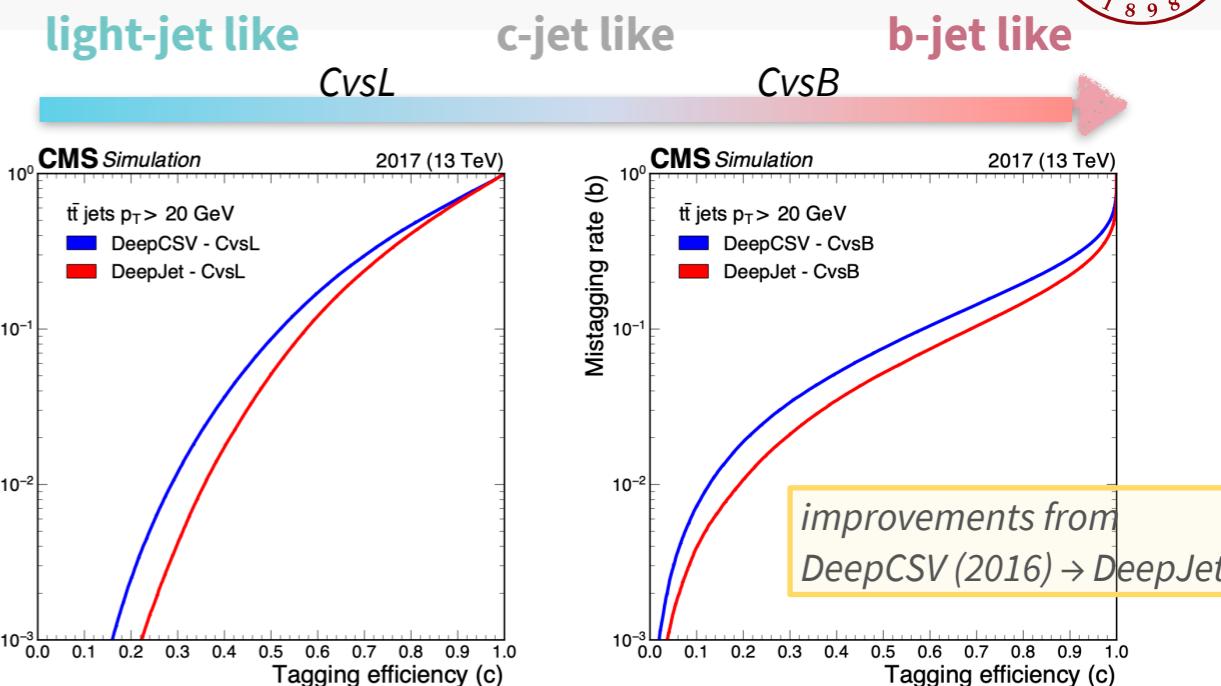
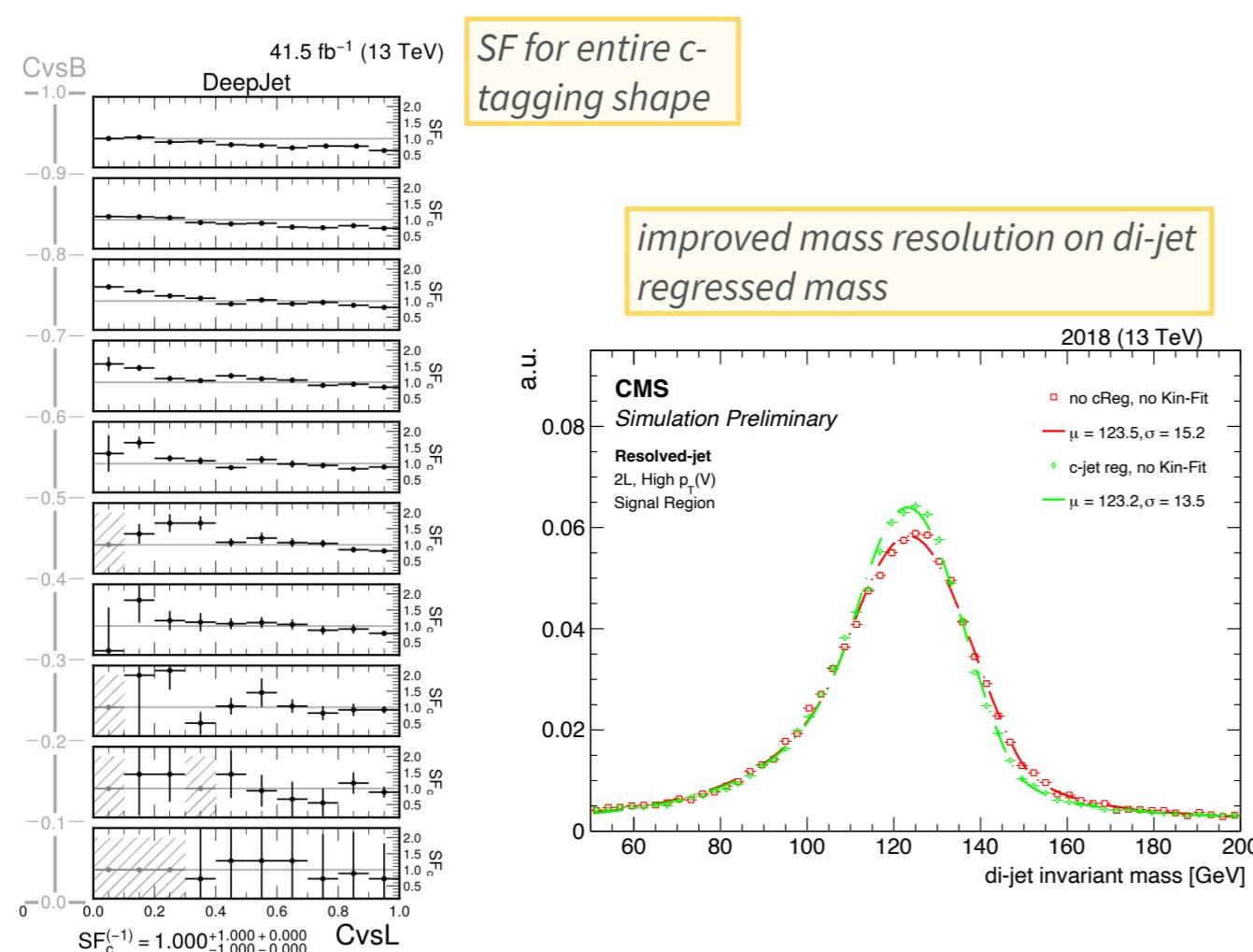




Resolved-jet: analysis techniques

→ Charm jet identification with DeepJet algorithm [JINST 15 (2020) P12012]

- ❖ deep neural network in CNN+RNN architecture
- ❖ two discriminants defined for c-jet vs. b-jet and c-jet vs. light-jet separation
- ❖ ~2x (40%) improvements for light (b)jet rejection at 40% c jet efficiency w.r.t. DeepCSV (2016)



→ Novel charm tagger shape calibration

[JINST 17 (2022) P03014]

- ❖ correct the entire distribution of the c-tagging shape
- ❖ SF derived with an iterative approach using three samples enriched in light/b/c jet

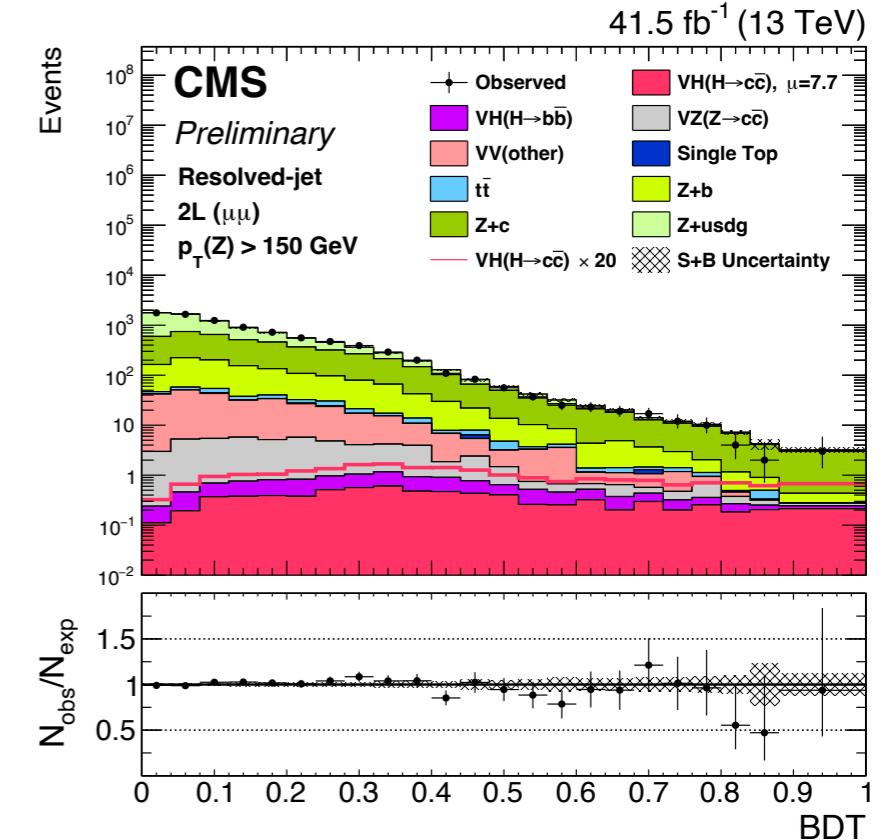
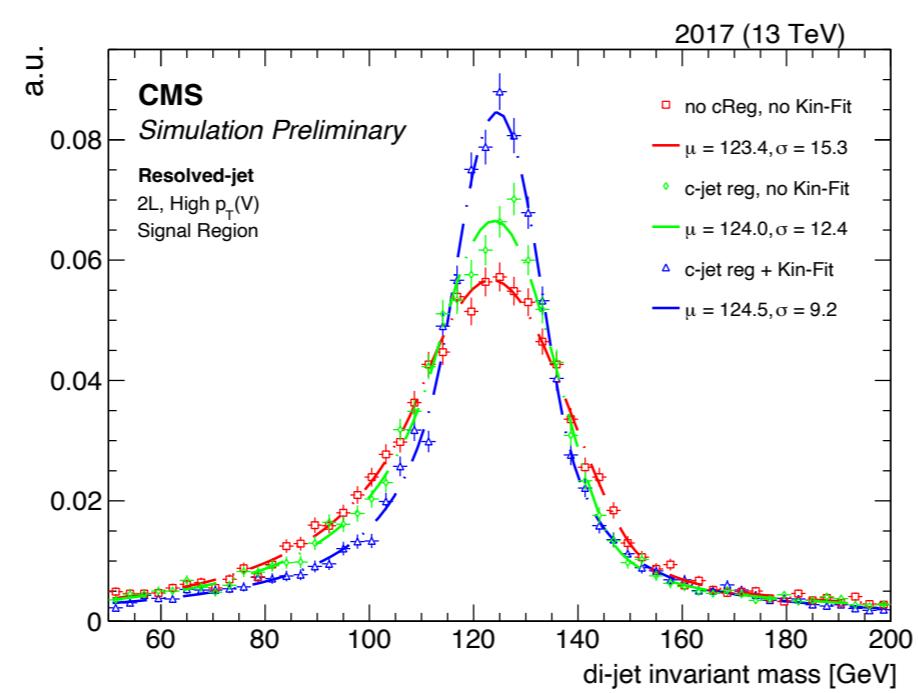
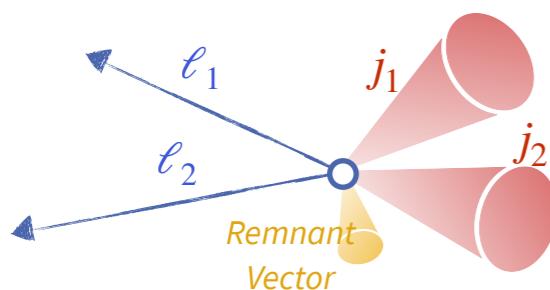
→ Charm jet energy regression

- ❖ regress RECO-to-GEN energy correction using DNN
- ❖ ~15% improvement in di-jet invariant mass resolution

Resolved-jet: analysis strategy

- define SR: two jets with **high c-tagging score**
- train **BDT** to separate VHcc signal vs. backgrounds
 - ❖ **kinematic fit** used in 2L: post-fit $Z(\ell\ell)H(cc)$ system kinematics used as BDT input
- CRs defined for V+b/c/light jets, and $t\bar{t}$
 - ❖ by inverting c-tagger selection or vetoing Z/H_{cand} mass / requiring additional ℓ/jet
- finally extract VHcc signal by fitting on BDT
 - ❖ **simultaneous fit** on SR (on BDT) and all CRs (on c-tagging discriminant)

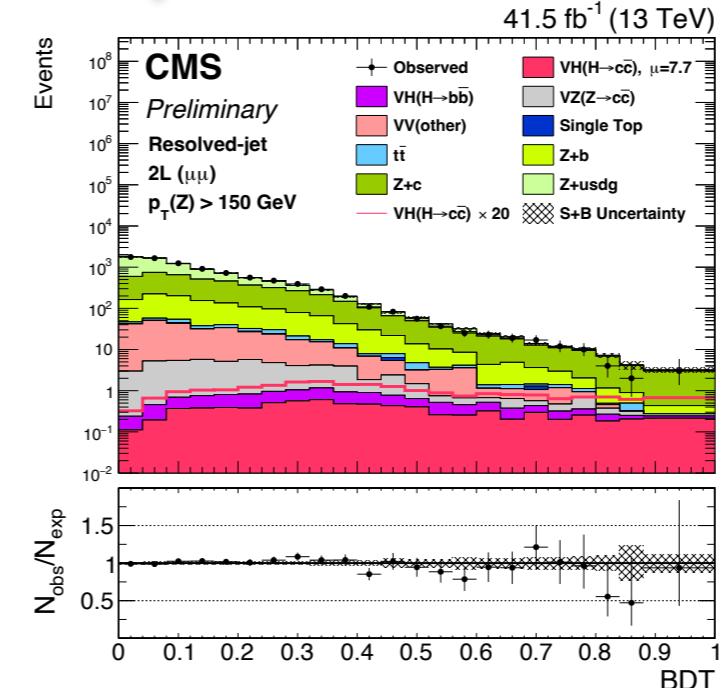
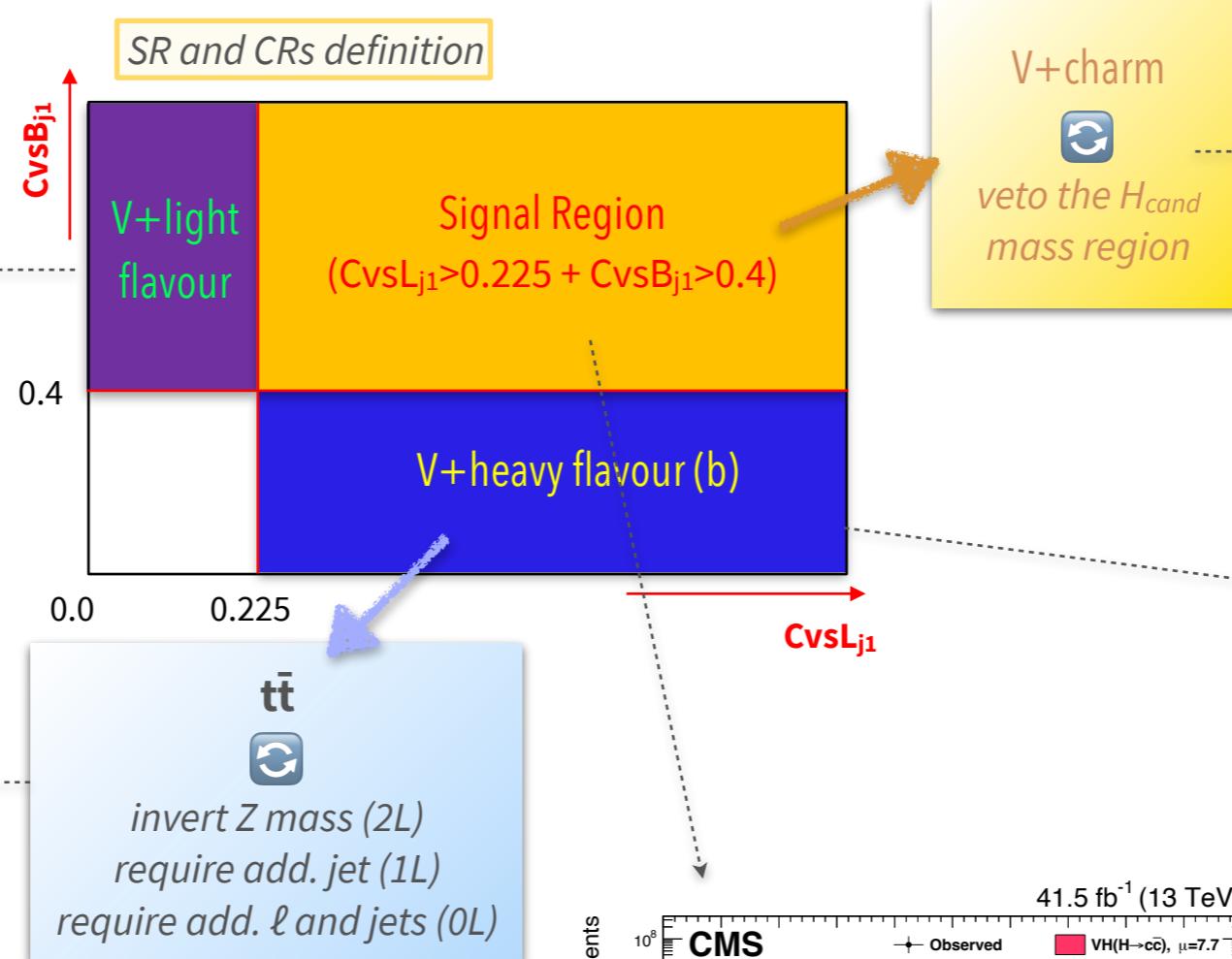
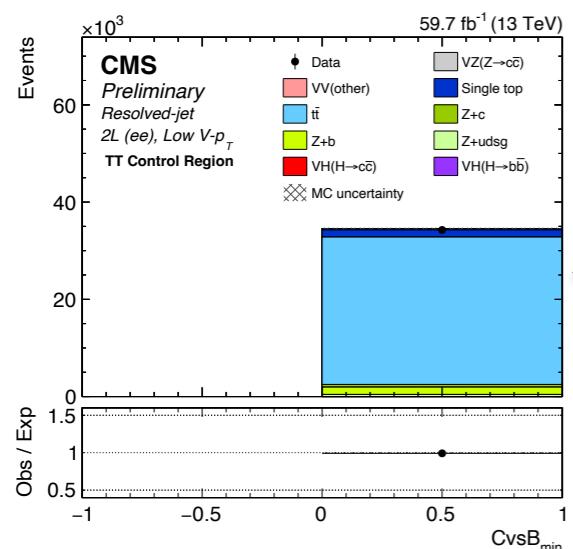
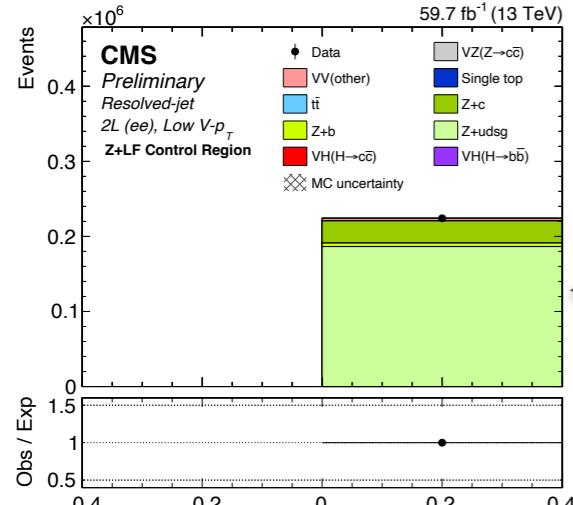
BDT to distinguish VHcc from backgrounds





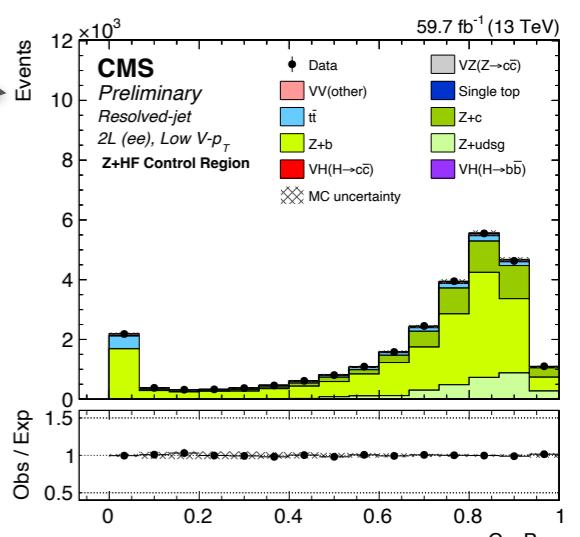
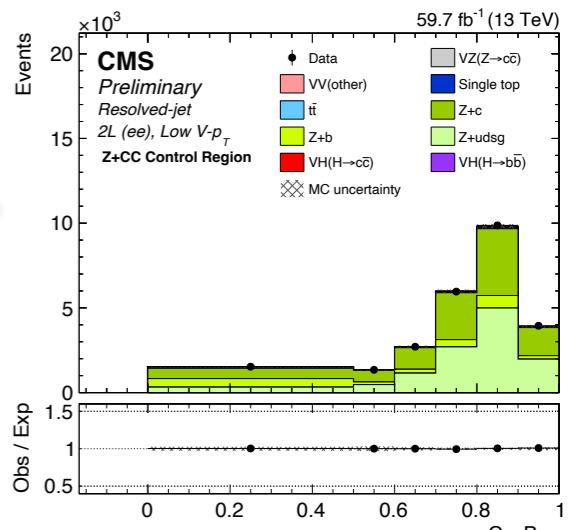
Resolved-jet: CRs and post-fit SR

example CR plots for 2L high- $p_T(V)$ channel, year 2017



→ Simultaneous fit in all SR and CRs

- ❖ **BKG normalisation factors** assigned to multiple components and extracted from the fit
- ❖ combined with merged-jet topology for the final signal extraction fit



SR plot (BKG impose norm factors extracted from CR-only fit)

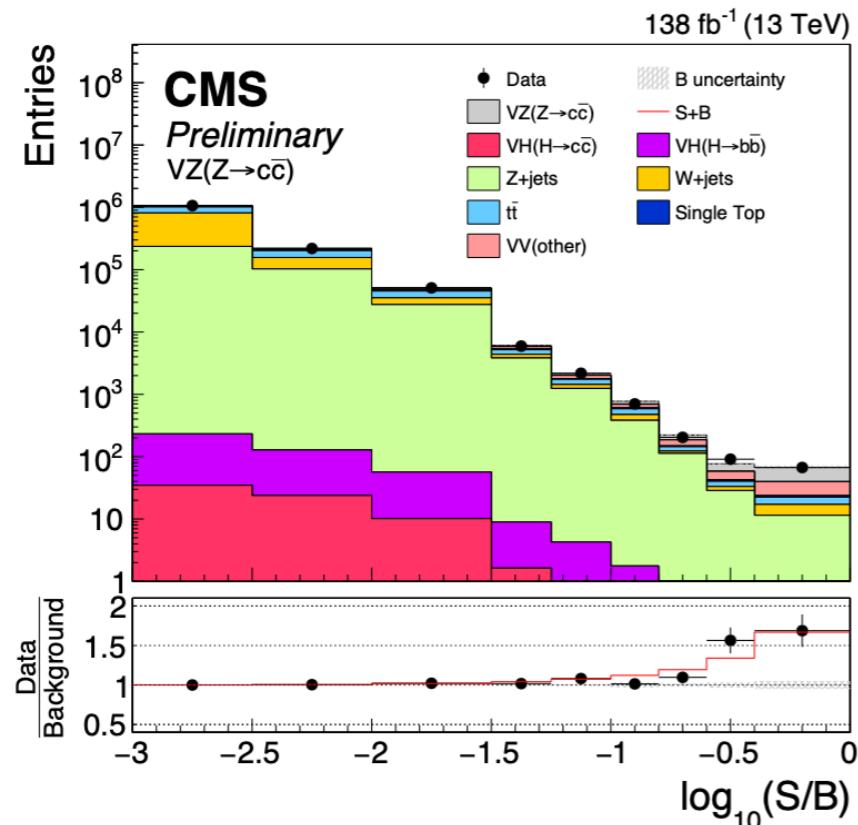


Results from VZcc validation

RESULT

→ Use VZcc analysis as a validation

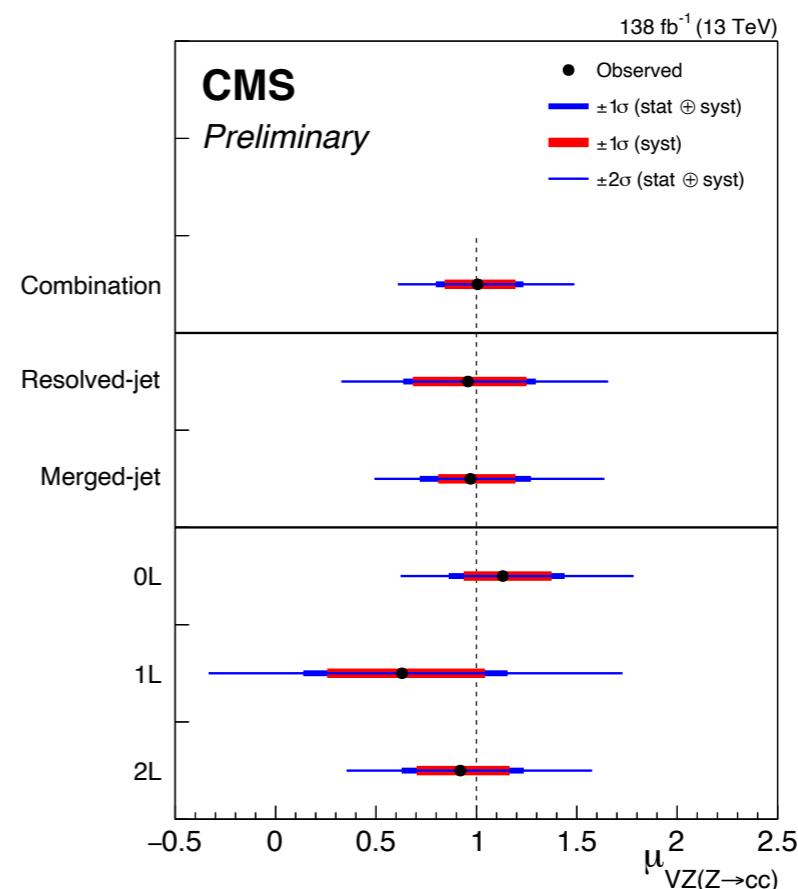
- ❖ consider **VZcc process as the signal** and VHcc as the background
- ❖ apply with the same analysis and fit strategy (for resolved-jet analysis: re-train the BDT using VZcc as the signal)



Observed significance for $VZ(Z \rightarrow cc)$: **5.7 σ**

- expected significance: 5.9σ

First observation of $Z \rightarrow cc$ at a hadron collider!



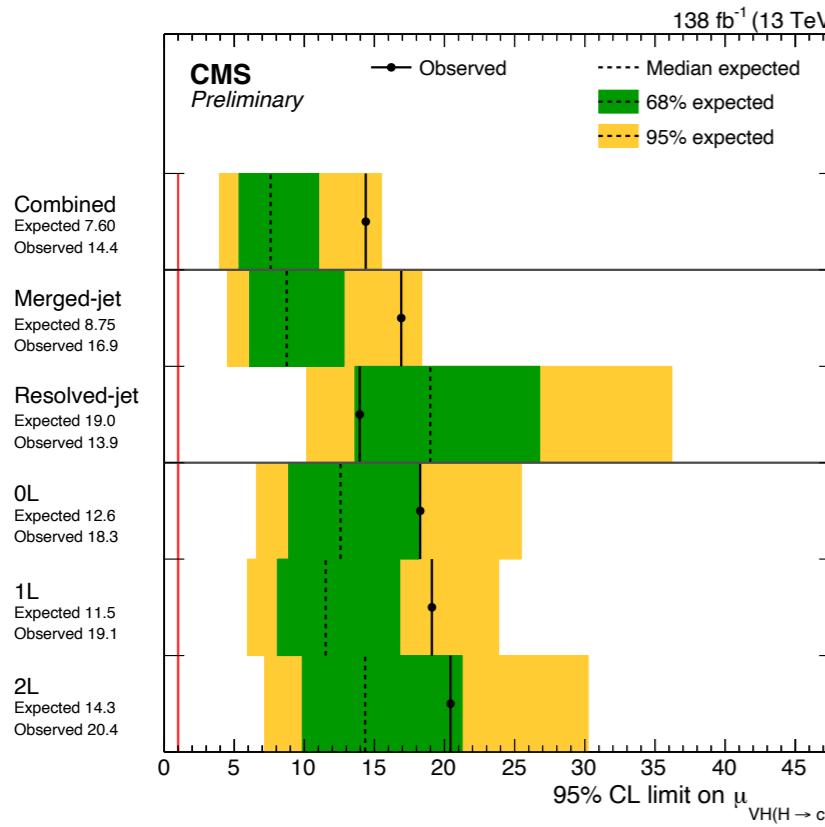
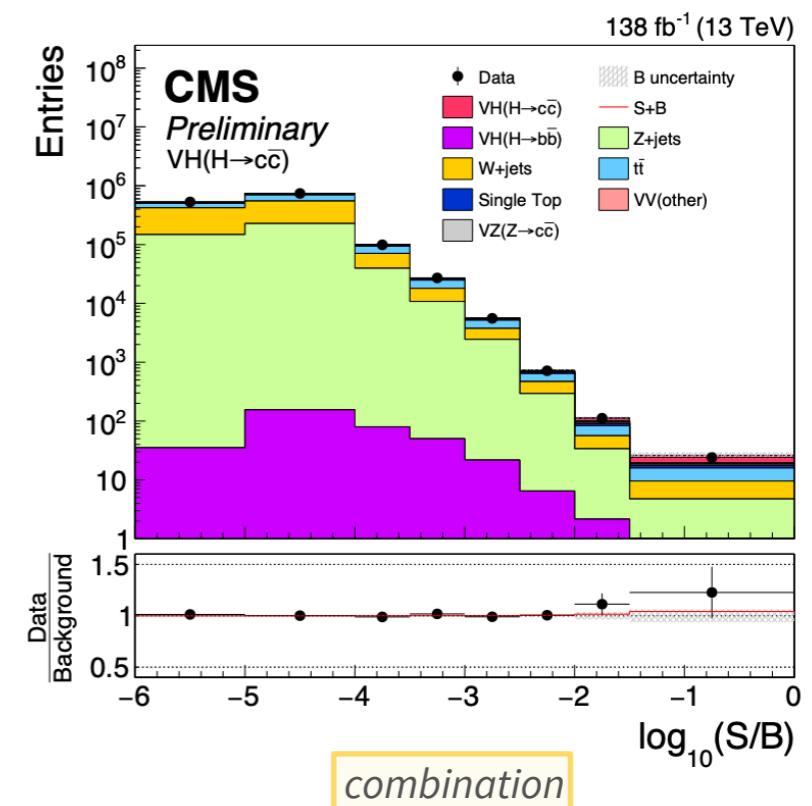
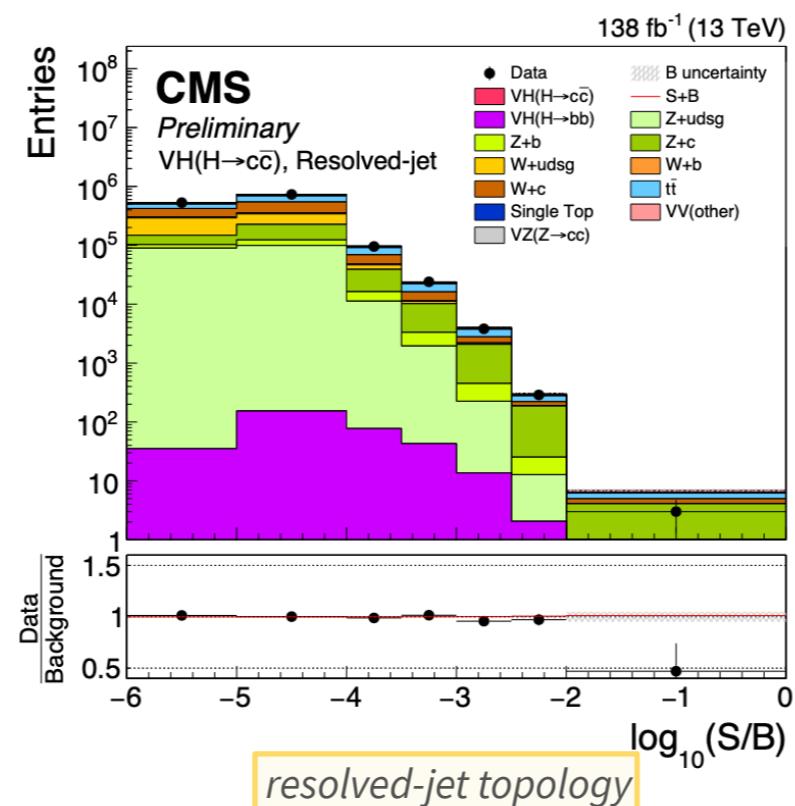
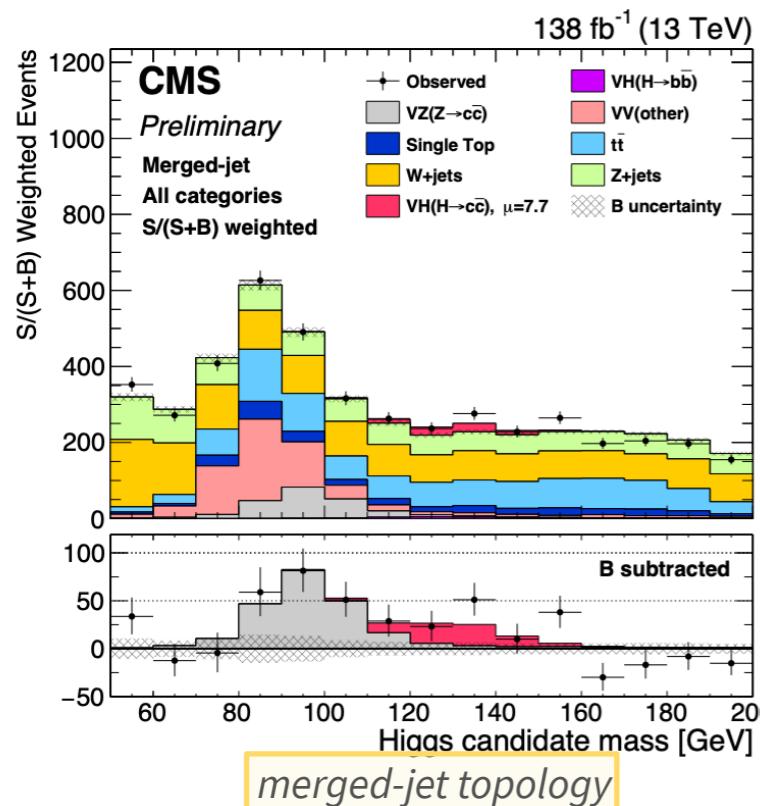
Best-fit signal strength: $\mu_{VZ(Z \rightarrow cc)} = 1.01^{+0.23}_{-0.21}$

- very good agreement with SM expectation
- consistent results between topologies/channels



VHcc results

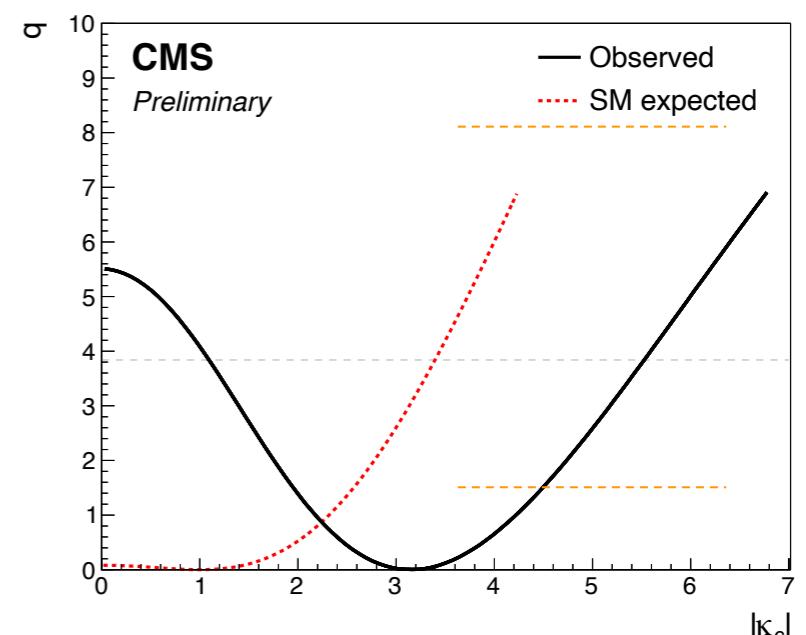
RESULT



Upper limits on the $VH(H \rightarrow cc)$ signal strength at 95% CL

- $\mu_{VH(H \rightarrow cc)} < 14$ (7.6) observed (expected)
- setting stronger limit w.r.t. ATLAS full Run 2 result:
 $\mu_{VH(H \rightarrow cc)} < 26$ (31) obs. (exp.)

[arXiv:2201.11428]



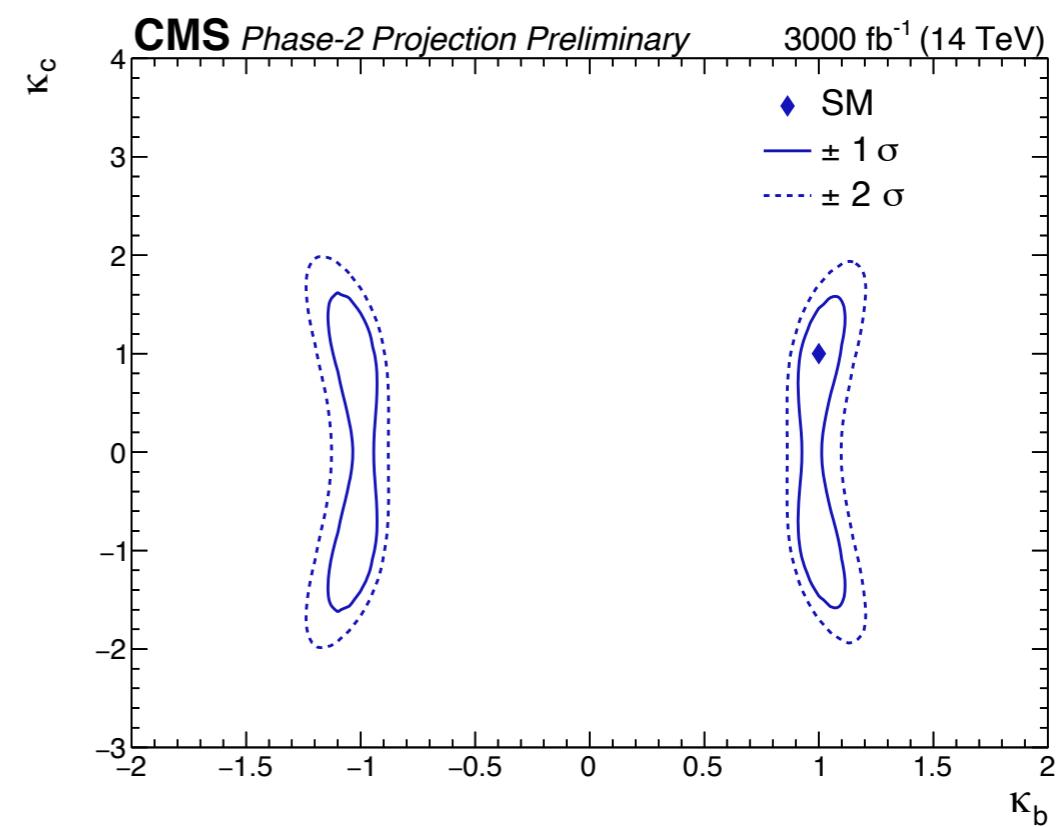
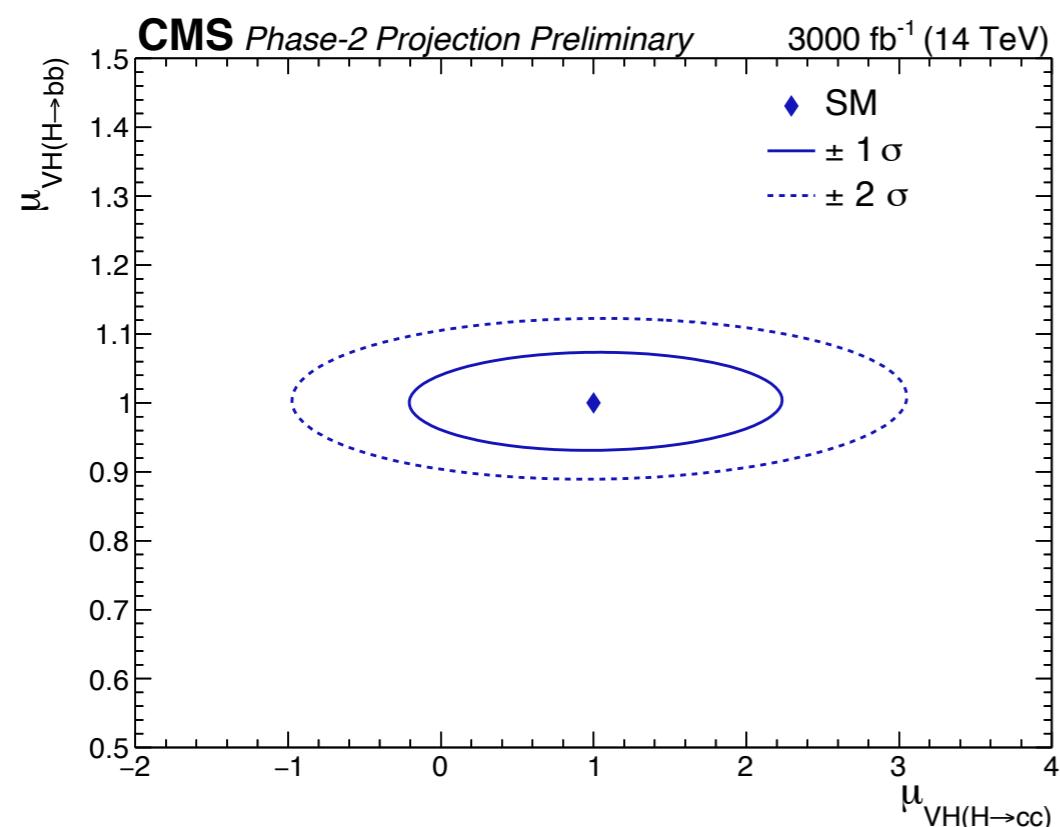
95% CL interval on κ_c :

$1.1 < |\kappa_c| < 5.5$ ($|\kappa_c| < 3.4$)
 observed (expected)



Projection at HL-LHC

- Strategy for **merged-jet topology** extrapolated to HL-LHC (3000 fb^{-1})
 - ❖ **simultaneous extraction of $H \rightarrow bb$ and $H \rightarrow cc$ signal strength**
 - ❖ based on Run 2 analysis and add new categories enriched for $H \rightarrow bb$; lower the pT threshold to 200 GeV → increase the signal acceptance



- $\mu_{VH(H \rightarrow bb)} = 1.00 \pm 0.03 \text{ (stat.)} \pm 0.04 \text{ (syst.)} = 1.00 \pm 0.05 \text{ (total)}$
- $\mu_{VH(H \rightarrow cc)} = 1.0 \pm 0.6 \text{ (stat.)} \pm 0.5 \text{ (syst.)} = 1.0 \pm 0.8 \text{ (total)}$



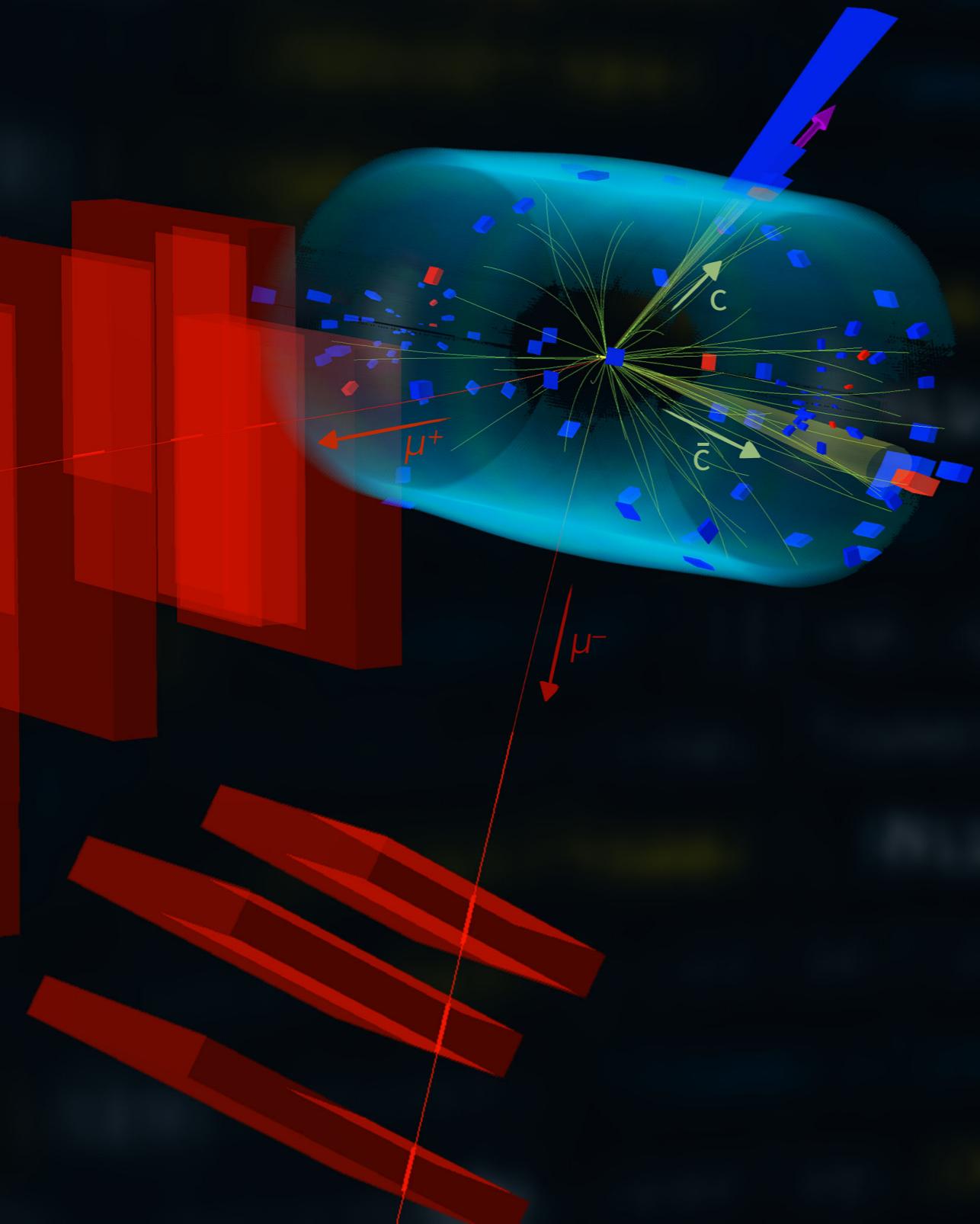
**Expected sensitivity reach $\sim o(0.1-1)$ level
with the use of novel tagger and strategy**



Summary

- The search of H \rightarrow cc in the VH production mode using full Run 2 data is presented
 - ❖ use two complementary approaches: **resolved- & merged-jet topology** to fully explore the VH(H \rightarrow cc) decay phase-space
 - results combined to further improve the sensitivity
 - ❖ exploit advanced **DNN-based charm & di-charm jet taggers**
 - **huge performance gains** in charm tagging with the new ML techniques
 - ❖ explore DNN-based method for **jet mass/energy regression**
- The full analysis procedure validated by measuring VZ(Z \rightarrow cc)
 - ❖ establish the first observation of Z \rightarrow cc on a hadron collider: significance at 5.7σ
 - ❖ set the most stringent limit on VH(H \rightarrow cc)
 - upper limit on 95% CL: $\mu_{VH(H \rightarrow cc)} < 14$ (7.6) observed (expected)
 - pave the way to possibly a real observation of H \rightarrow cc in the future

Backup

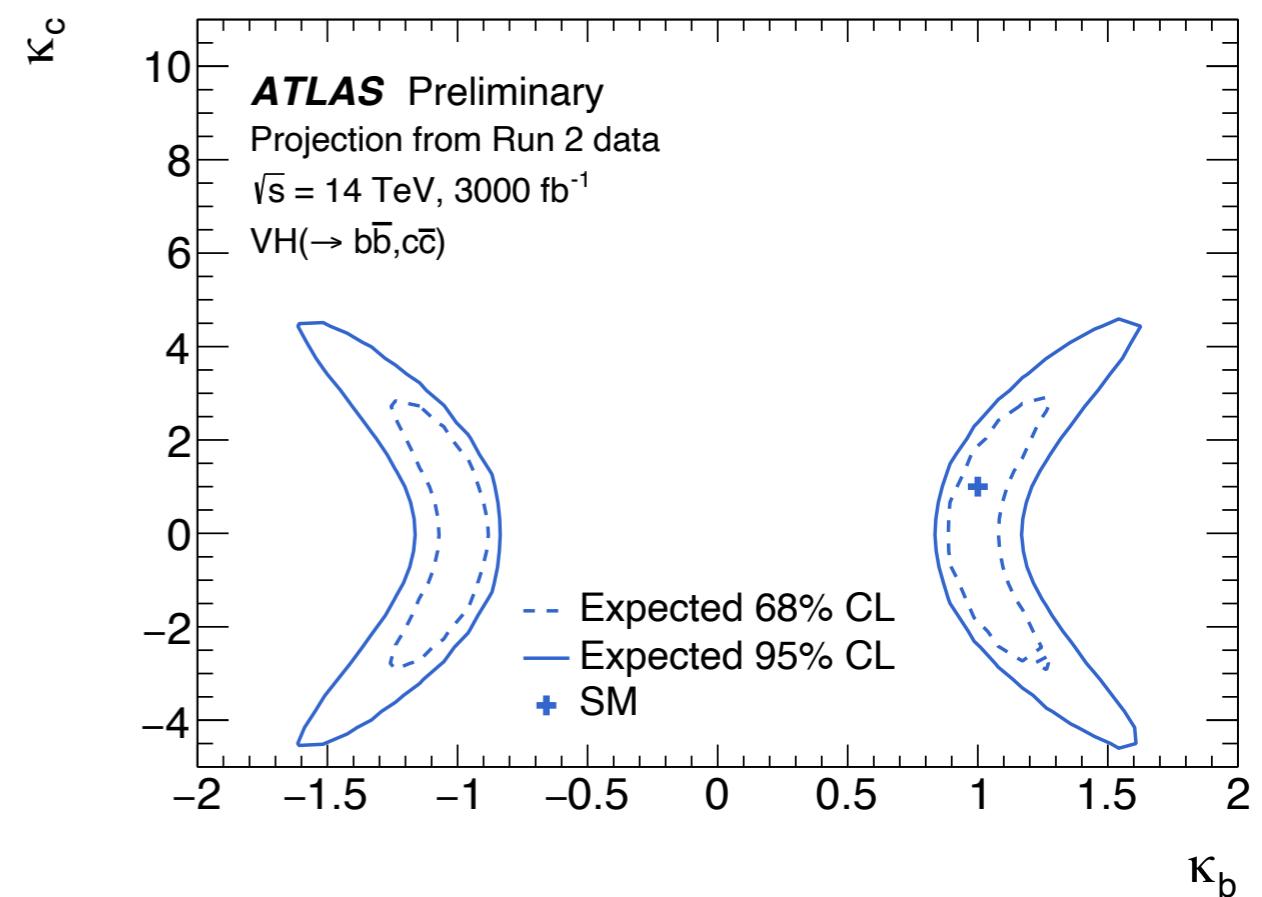
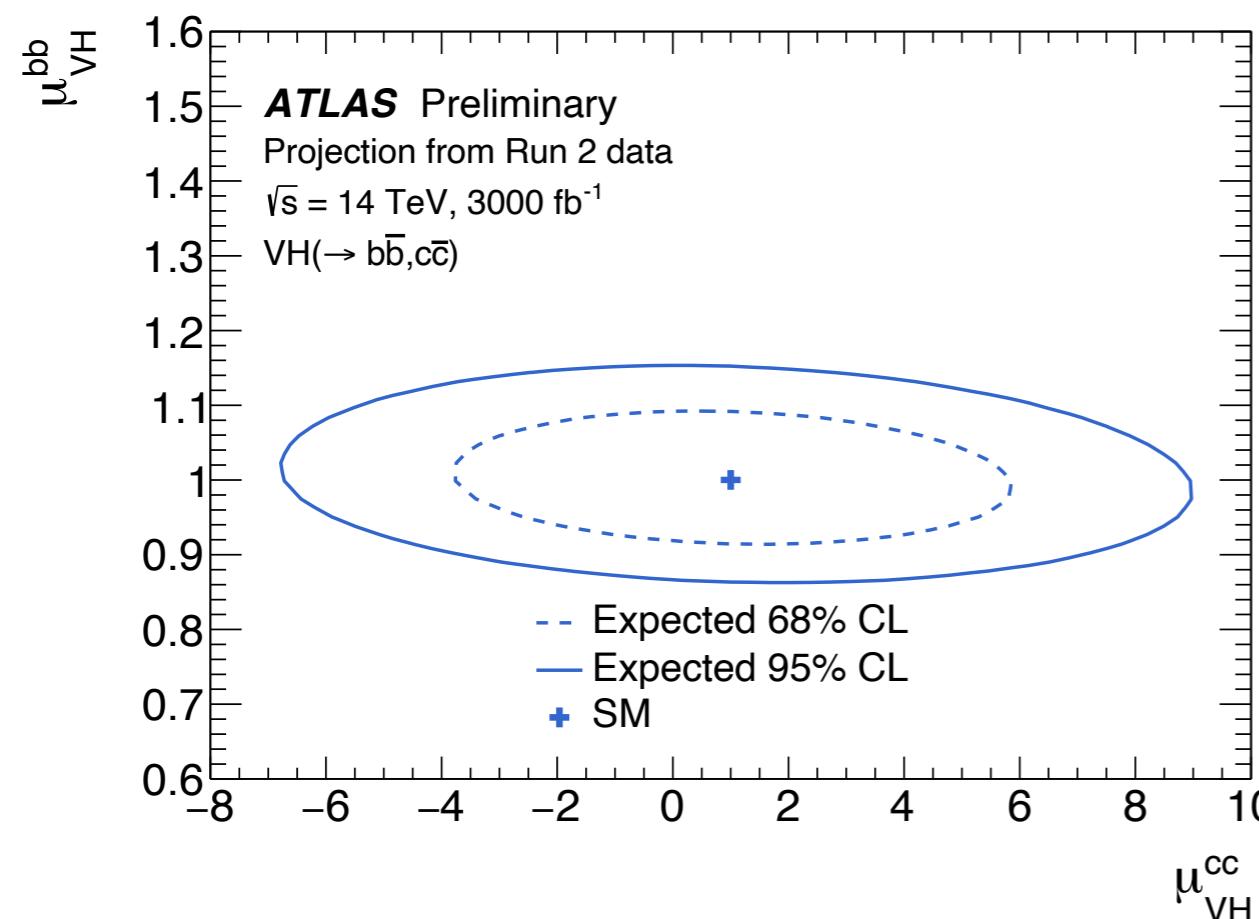


Higgs-charm coupling at HL-LHC

→ Expected sensitivity at HL-LHC [CERN-2019-007]

H \rightarrow cc decay	$ \kappa_c < \sim 1\text{-}2$
cH production	$ \kappa_c < \sim 2\text{-}3$
H \rightarrow J/ $\Psi\gamma$	$ \kappa_c < \sim 80$
p _T (H) distribution	$ \kappa_c < \sim 10$
WH charge asymmetry	$ \kappa_c < \sim 4\text{-}5$

ATLAS HL-LHC projection for $H \rightarrow cc$



Baseline event selection

Merged-jet topology

Variable	0L	1L	2L
p_T^ℓ	—	(>25,>30)	>20
Lepton isolation	—	(<0.06,—)	(<0.25,—)
$N_{a\ell}$	=0	=0	—
$M(\ell\ell)$	—	—	75–105
$N_{\text{small-}R}^{\text{aj}}$	<2	<2	<3
p_T^{miss}	>200	>60	—
$p_T(V)$	>200	>150	>150
$p_T(H_{\text{cand}})$	>300	>300	>300
$m(H_{\text{cand}})$	50–200	50–200	50–200
$\Delta\phi(V, H_{\text{cand}})$	>2.5	>2.5	>2.5
$\Delta\phi(\vec{p}_T^{\text{miss}}, j)$	>0.5	—	—
$\Delta\phi(\vec{p}_T^{\text{miss}}, \ell)$	—	<1.5	—
Kinematic BDT	>0.55	0.55–0.7, >0.7	>0.55
c \bar{c} discriminant			
High purity	>0.99	>0.99	>0.99
Medium purity	0.96–0.99	0.96–0.99	0.96–0.99
Low purity	0.90–0.96	0.90–0.96	0.90–0.96

Resolved-jet topology

Variable	0L	1L	2L low- $p_T(V)$	2L high- $p_T(V)$
p_T^ℓ	—	(>25,>30)	>20	>20
Lepton isolation	—	(<0.06,—)	(<0.25,—)	(<0.25,—)
$N_{a\ell}$	=0	=0	—	—
$M(\ell\ell)$	—	—	75–105	75–105
$p_T(j_1)$	>60	>25	>20	>20
$p_T(j_2)$	>35	>25	>20	>20
$CvsL(j_1)$	>0.225	>0.225	>0.225	>0.225
$CvsB(j_2)$	>0.4	>0.4	>0.4	>0.4
$N_{\text{small-}R}^{\text{aj}}$	—	<2	—	—
p_T^{miss}	>170	—	—	—
p_T^{miss} significance	—	>4	—	—
$p_T(V)$	>170	>100	60–150	>150
$p_T(H_{\text{cand}})$	>120	>100	—	—
$m(H_{\text{cand}})$	<250	<250	<250	<250
$\Delta\phi(V, H_{\text{cand}})$	>2.0	>2.5	>2.5	>2.5
$\Delta\phi(\vec{p}_T^{\text{miss}}, j)$	>0.5	—	—	—
$\Delta\phi(\vec{p}_T^{\text{miss}}, \ell)$	—	<2.0	—	—

Uncertainties

Merged-jet topology

Uncertainty source	$\Delta\mu / (\Delta\mu)_{\text{tot}}$
Statistical	88%
Background normalizations	39%
Experimental	40%
Sizes of the simulated samples	24%
Charm identification efficiencies	26%
Jet energy scale and resolution	15%
Simulation modeling	1%
Luminosity	5%
Lepton identification efficiencies	2%
Theory	25%
Backgrounds	21%
Signal	14%

Resolved-jet topology

Uncertainty source	$\Delta\mu / (\Delta\mu)_{\text{tot}}$
Statistical	66%
Background normalizations	28%
Experimental	72%
Sizes of the simulated samples	59%
Charm identification efficiencies	27%
Jet energy scale and resolution	17%
Simulation modeling	20%
Luminosity	13%
Lepton identification efficiencies	10%
Theory	22%
Backgrounds	21%
Signal	7%

Datasets and MC samples

→ Datasets and triggers

- ❖ 138 fb $^{-1}$ for full Run 2 data
- ❖ **0L**: MET dataset + MET_MHT triggers
- ❖ **1L**: SingleElectron/SingleMuon dataset + single-e/ μ triggers
- ❖ **2L**: DoubleEG/DoubleMuon dataset + double-e/ μ triggers
 - ▶ (merged-jet analysis uses single-e/ μ datasets and triggers)

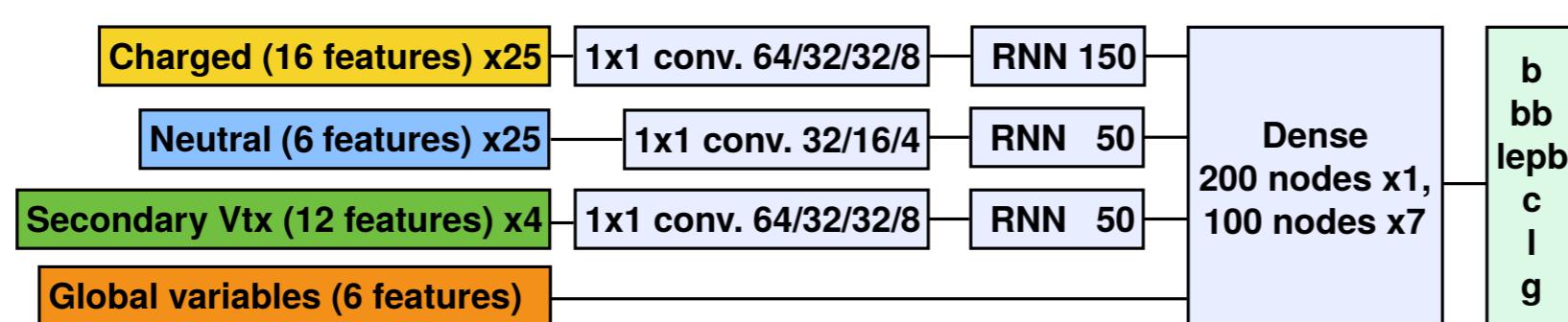
→ Simulated samples

- ❖ **W/Z+jets**: MadGraph5 NLO samples
- ❖ **t \bar{t}** : POWHEG NLO sample
- ❖ **single top**: POWHEG/MadGraph5 NLO samples
- ❖ **VV**: POWHEG/MadGraph5 NLO samples (NNLO QCD & NLO EW correction on $p_T(V_2)$)
- ❖ **VH signals**: POWHEG NLO samples (NLO EWK $p_T(V)$ correction; NNLO cross section)

→ Corrections

- ❖ lepton efficiencies, JEC/JER, pileup reweighting, MET filters, c-tagging, etc.

DeepJet details



DeepJet architecture
[JINST 15 (2020) P12012]

→ Use low-level features (PF candidates, SV) as well as global features

→ Model architecture:

- ❖ separate 1D CNNs to process three low-level feature classes
 - for each class, concatenate multiple CNNs with decreasing dimensions
 - compress the features to lower dimensional space
- ❖ RNNs (LSTM type) applied after CNNs
 - better handles the variable length sequence (PF candidates/SV)
- ❖ fully connected layer to connect all channels

→ Output score:

- ❖ 6 raw scores: bb, b (hadronically decay B-hadron), b_{lep} (leptonically decay B-hadron), c, light (uds), gluon
- ❖ construct CvsB, CvsB used in VHcc analysis

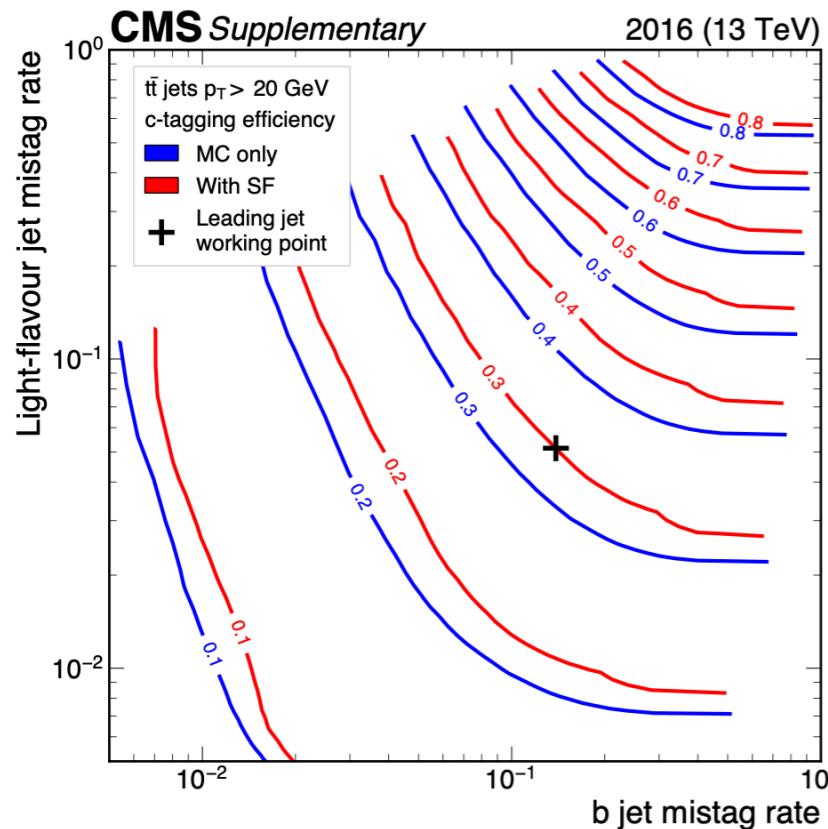
DeepJet discriminant

	BvsC/L	CvsB	CvsL
DeepJet :	$P(b) + P(bb) + P(b_{lep})$	$\frac{P(c)}{P(c) + P(b) + P(bb) + P(b_{lep})}$	$\frac{P(c)}{P(c) + P(uds) + P(g)}$

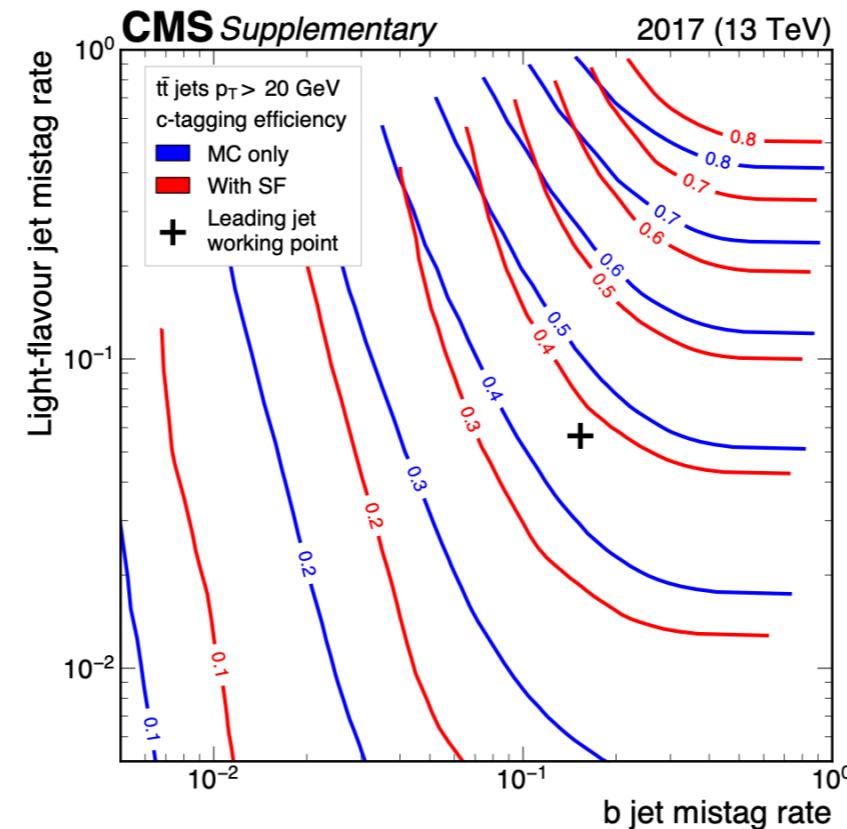
Resolved

DeepJet performance

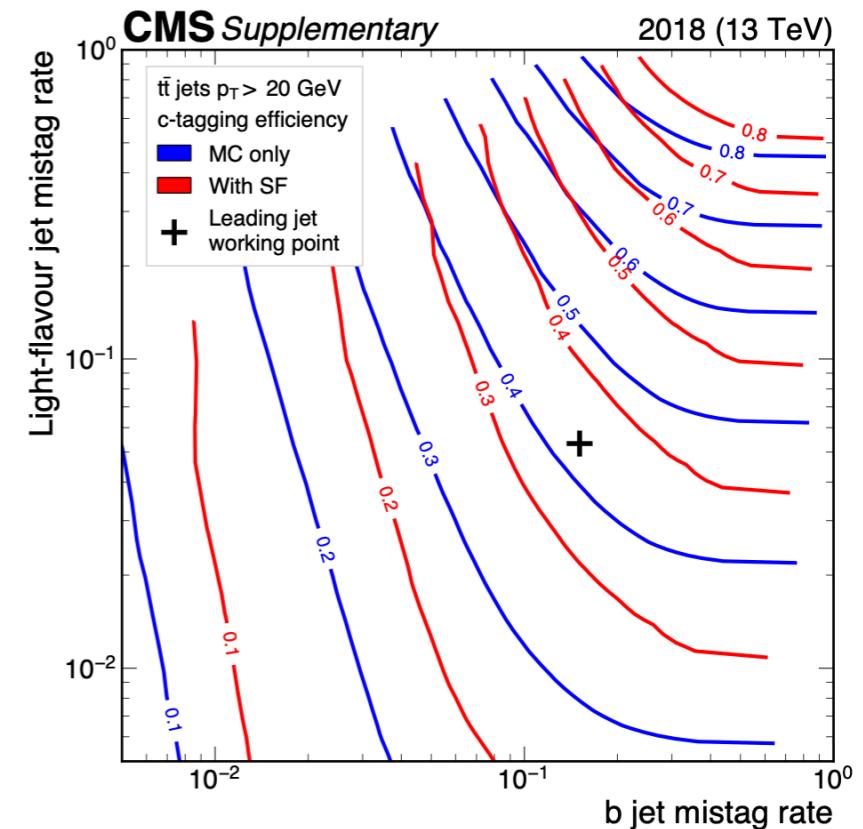
2016



2017



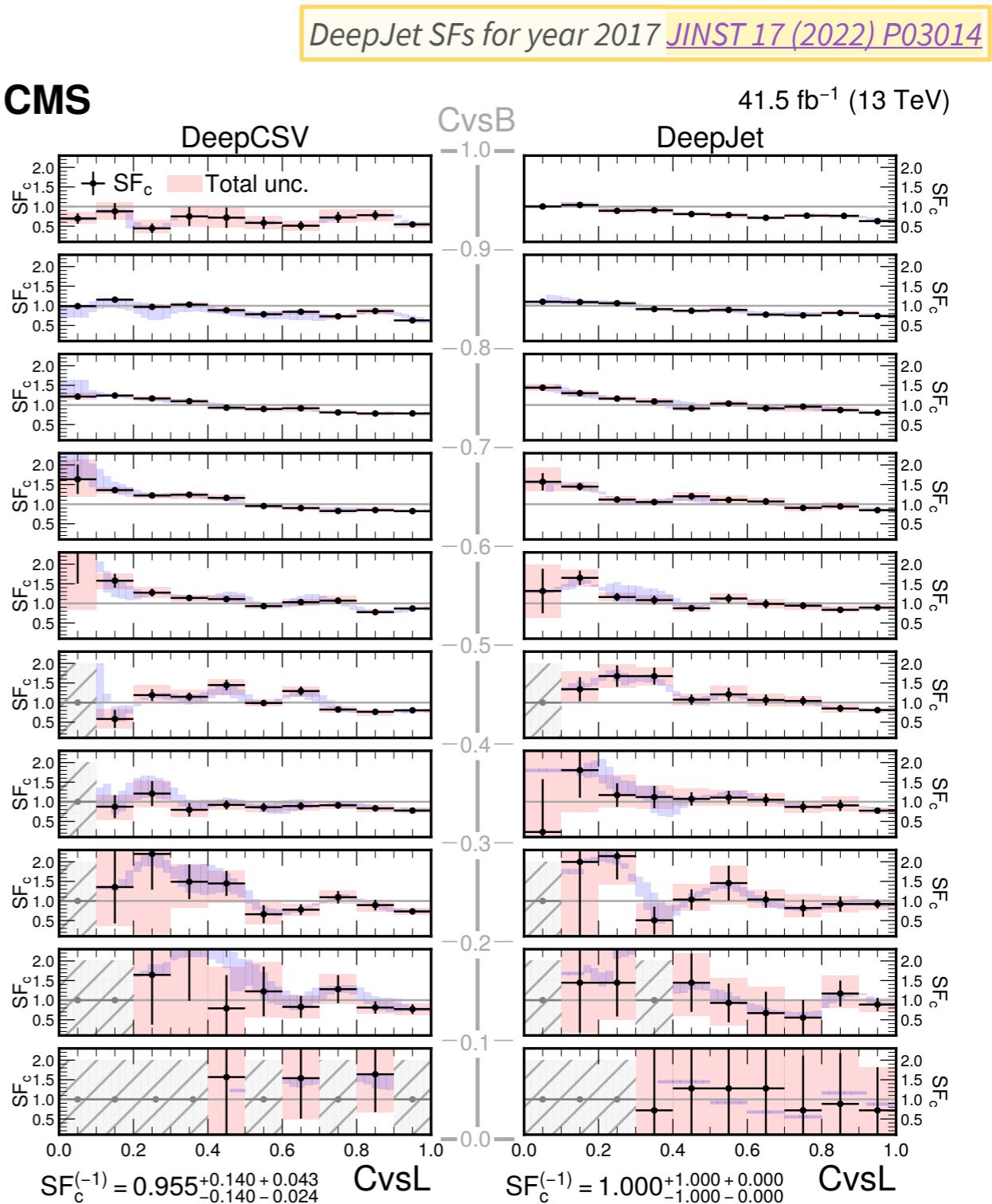
2018



- CMS c-tagging WP: ~40% (c), ~16% (b), ~4% (light)
- ATLAS c-tagging WP: ~27% (c), 8% (b), 1.6% (light) [[arXiv:2201.11428](https://arxiv.org/abs/2201.11428)]

Charm tagging calibration

- Per-jet SFs derived as a 3D function of CvsL score, CvsB score, and 3 true flavours
 - ❖ 3 selections pure in b, c, and light jets
 - ❖ iteratively fit in 3 selected phase-space to find simultaneous corrections for all 3 flavours
 - ▶ $Z(\ell\ell) + \text{jets}$ (light jet enriched)
 - ▶ $W + c$ (c-jet enriched)
 - ▶ $t\bar{t}$ (b-jet enriched)
- Propagate to analysis
 - ❖ per-event weight calculated as ***products*** of SF_i (flavour, CvsL, CvsB) for all jet i

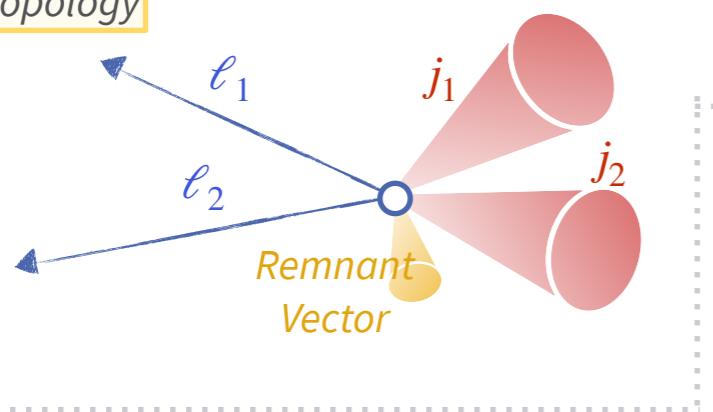


Kinematic fit & signal extraction BDT for resolved-jet

→ Kinematic fit (KinFit) in 2L channel

- ❖ to achieve more precise jet p_T measurement and improve Higgs mass resolution by 30%

KinFit topology



→ BDTs are trained in 4 categories, 3 years separately to discriminate VHcc signal vs. backgrounds

- ❖ to maximize the sensitivity
- ❖ **BDT used as the fit variable** in SR to extract the VHcc signal

❖ apply the fit by:

- constrain di-lepton system to Z mass
- balance the $\ell\ell+cc$ system in (p_x, p_y) plane
- allow $p_{T,j1,2}$ to adjust within the uncertainties
- add recoil jets in the p_T -constraint if they are present

		0L	1L	2L
<i>Higgs and vector boson properties</i>	$m(H)$	✓	✓	✓
	$p_T(H)$	—	✓	✓
	$p_T(V)$	—	✓	✓
	$m_T(V)$	—	✓	—
	p_T^{miss}	✓	✓	—
<i>c-tagging score</i>	$p_T(V)/p_T(H)$	✓	✓	✓
	$CvsL_{\max}$	✓	✓	✓
	$CvsB_{\max}$	✓	✓	✓
	$CvsL_{\min}$	✓	✓	✓
	$CvsB_{\min}$	✓	✓	✓
<i>event kinematics</i>	$p_{T\max}$	✓	✓	✓
	$p_{T\min}$	✓	✓	✓
	$\Delta\phi(V, H)$	✓	✓	✓
	$\Delta R(j_1, j_2)$	—	✓	✓
	$\Delta\phi(j_1, j_2)$	✓	✓	—
	$\Delta\eta(j_1, j_2)$	✓	✓	✓
	$\Delta\phi(\ell_1, \ell_2)$	—	✓	✓
	$\Delta\eta(\ell_1, \ell_2)$	—	—	✓
	$\Delta\phi(\ell_1, j_1)$	—	✓	—
	$\Delta\phi(\ell_2, j_1)$	—	—	✓
	$\Delta\phi(\ell_2, j_2)$	—	—	✓
	$\Delta\phi(\ell_1, p_T^{\text{miss}})$	—	✓	—
	$\sigma_{cReg}(j_1)$	✓	✓	✓
	$\sigma_{cReg}(j_2)$	✓	✓	✓
<i>Variables from KinFit (for 2L only)</i>	$\Delta\eta(V, H) \ _{\text{kinfit}}$	—	—	✓
	$\Delta\phi(V, H) \ _{\text{kinfit}}$	—	—	✓
	$m(H) \ _{\text{kinfit}}$	—	—	✓
	$p_T(H) \ _{\text{kinfit}}$	—	—	✓
	$p_{T\max} \ _{\text{kinfit}}$	—	—	✓
	$p_{T\min} \ _{\text{kinfit}}$	—	—	✓
	$p_T(V)/p_T(H) \ _{\text{kinfit}}$	—	—	✓
	$\sigma(H) \ _{\text{kinfit}}$	—	—	✓
<i>BDT training variables</i>				

ParticleNet architecture and inputs

→ ParticleNet harness the Dynamic Graph CNN (DGCNN) [[arXiv:1801.07829](https://arxiv.org/abs/1801.07829)]

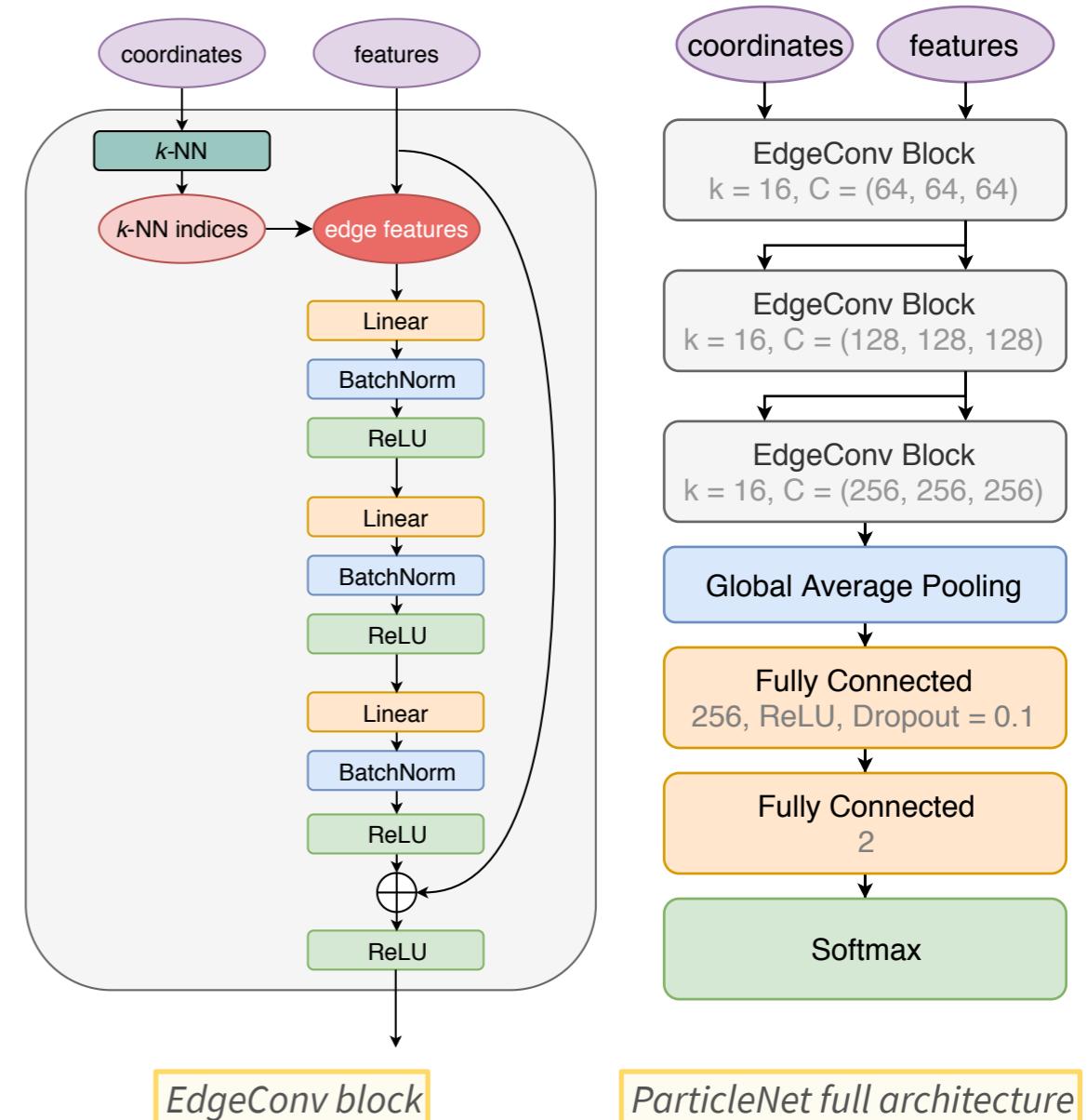
- ❖ harness EdgeConv operation to pass on learned features for each point which is interpreted as “coordinates” in high-dim latent space
- ❖ increasing layer size for EdgeConv
- ❖ fully connected layer transfer the feature space to N outputs

→ Input feature

- ❖ up to 100 PF candidates per jet with 20 features + 2 coordinates (η_{rel} , ϕ_{rel})
- ❖ up to 10 SVs per jet with 11 features + 2 coordinates (η_{rel} , ϕ_{rel})

→ Training performed by the framework [weaver](#)

→ More details in [[CMS-DP-2020-002](#)]

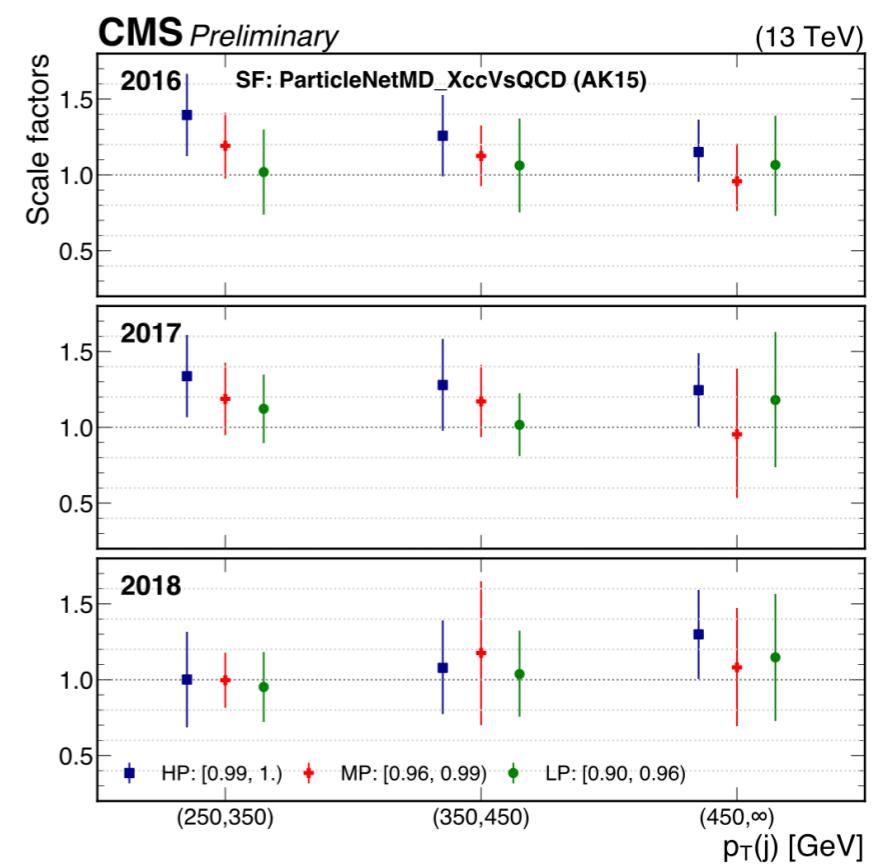
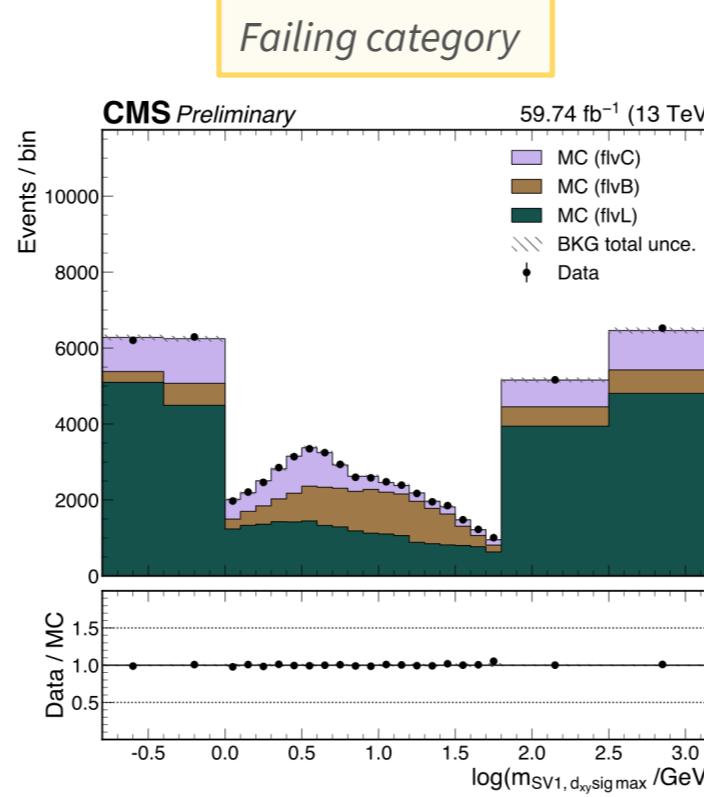
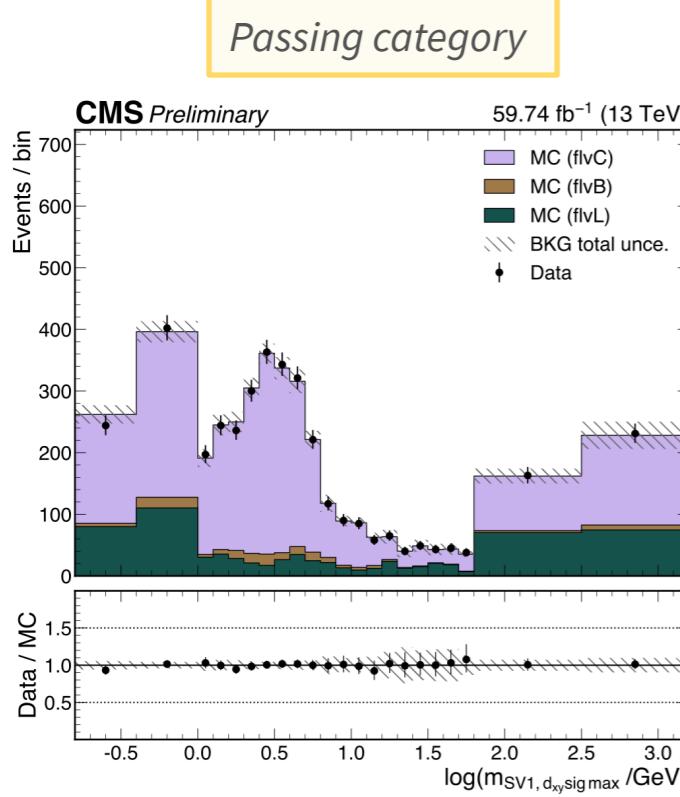


Merged

ParticleNet calibration

→ ParticleNet tagger calibration for H \rightarrow cc signal jet

- ❖ use g \rightarrow cc from QCD multi-jet events as a proxy to H \rightarrow cc jets
- ❖ ***sfBDT***: select a phase-space from g \rightarrow cc that resembles H \rightarrow cc jets
 - provide a handle to adjust the proxy–signal jet similarity
- ❖ simultaneous fit in pass and fail tagger region
 - fit variable: $\log(m_{\text{sv}})$
 - 3 rate parameters for c(cc), b(bb), light
 - improved treatment of the systematics
- ❖ details in [[CMS-DP-2022-005](#)]



Merged

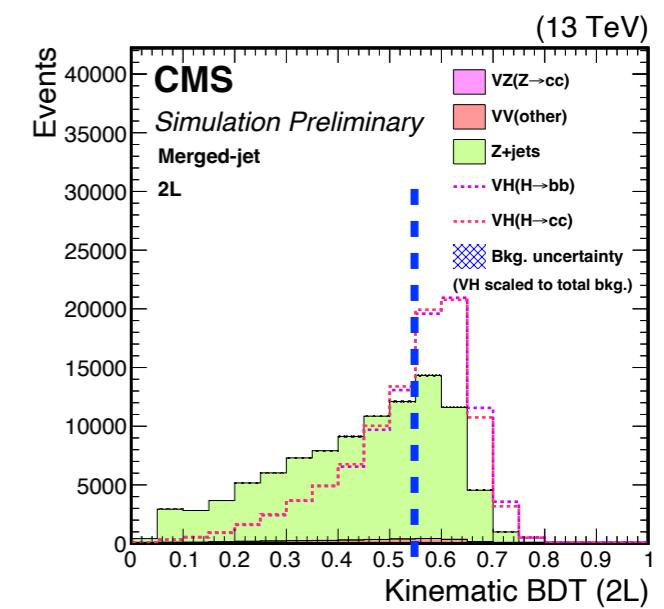
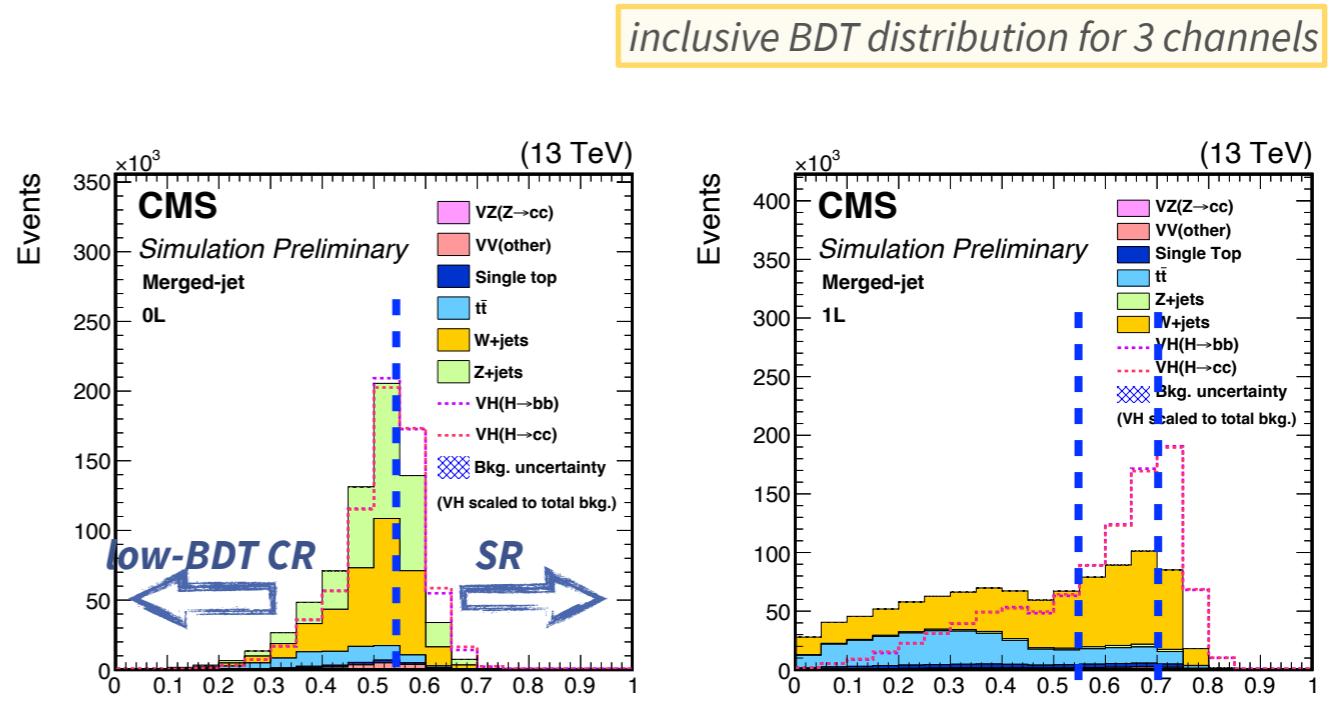
kinematic BDT for merged-jet topology

- Kinematic BDT developed to better separate VH signals from major backgrounds ($V+jets, t\bar{t}$)

- ❖ using only event kinematics, not intrinsic property (e.g. mass/flavour) of the AK15 jets
 - ▶ BDT largely uncorrelated with m_{reg} and ParticleNet cc-tagger
- ❖ low-BDT CR defined based on kinematic BDT

input variables to kinematic BDT

Variable	Description	0-lepton	1-lepton	2-lepton
$p_T(V)$	vector boson transverse momentum		✓	✓
$\Delta R(\ell, \ell)$	angular separation between the two leptons			✓
$p_T(H_{\text{cand}})$	H_{cand} transverse momentum	✓	✓	✓
$ \eta(H_{\text{cand}}) $	absolute value of the H_{cand} pseudorapidity	✓		
$\Delta\phi(V, H_{\text{cand}})$	azimuthal angle between vector boson and H_{cand}	✓	✓	✓
p_T^{miss}	missing transverse momentum		✓	
$\Delta\eta(H_{\text{cand}}, \ell)$	difference in pseudorapidity between H_{cand} and the lepton		✓	
$\Delta\eta(H_{\text{cand}}, V)$	difference in pseudorapidity between H_{cand} and vector boson			✓
$\Delta\eta(H_{\text{cand}}, j)$	min. difference in pseudorapidity between H_{cand} and small- R jets	✓	✓	✓
$\Delta\eta(\ell, j)$	min. difference in pseudorapidity between the lepton and small- R jets		✓	
$\Delta\eta(V, j)$	min. difference in pseudorapidity between vector boson and small- R jets			✓
$\Delta\phi(\vec{p}_T^{\text{miss}}, j)$	azimuthal angle between \vec{p}_T^{miss} and closest small- R jet	✓		
$\Delta\phi(\vec{p}_T^{\text{miss}}, \ell)$	azimuthal angle between \vec{p}_T^{miss} and lepton		✓	
m_T	transverse mass of lepton $\vec{p}_T + \vec{p}_T^{\text{miss}}$		✓	
N_j	number of small- R jets	✓	✓	✓



Merged

Merged-jet: analysis strategy

→ Signal region selections and fit

- ❖ kinematic BDT \rightarrow cc-tagging discriminant (ParticleNet score) \rightarrow fit on regressed Higgs candidate mass

