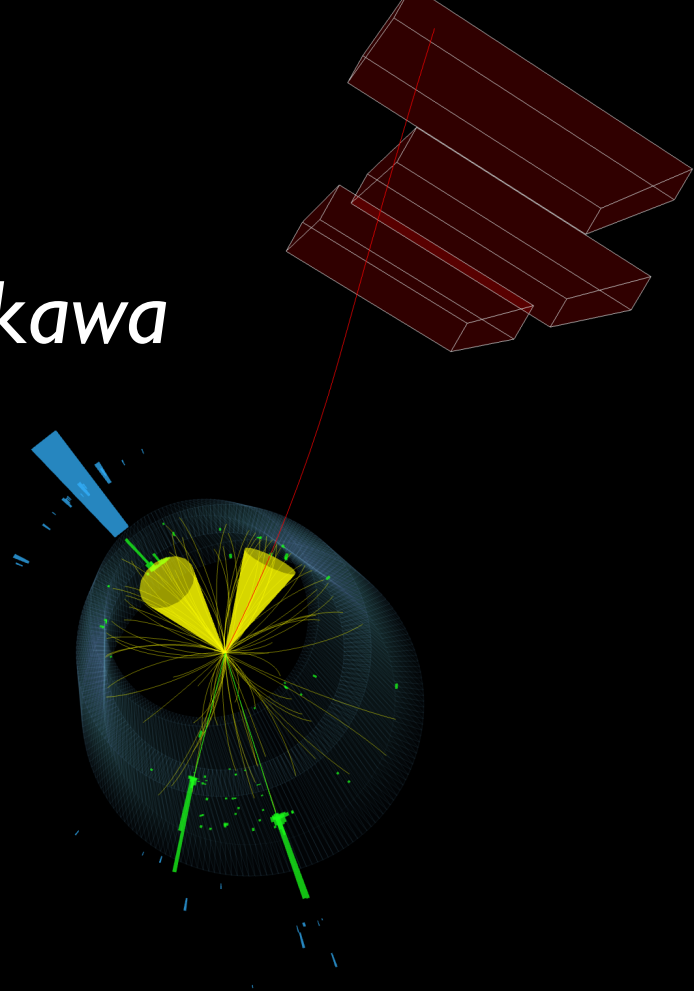




Constraining the Higgs-charm Yukawa coupling in the CMS experiment

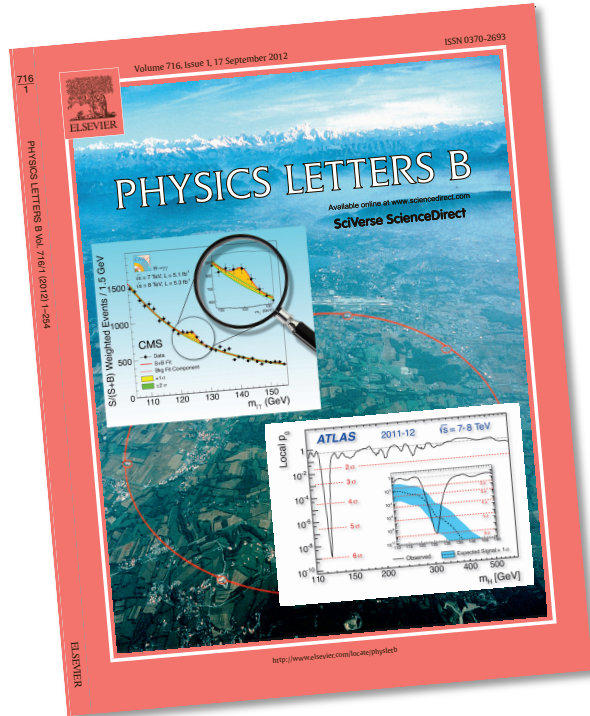
曲慧麟 (CERN)

中山-北大联合高能物理青年论坛
2022.05.18



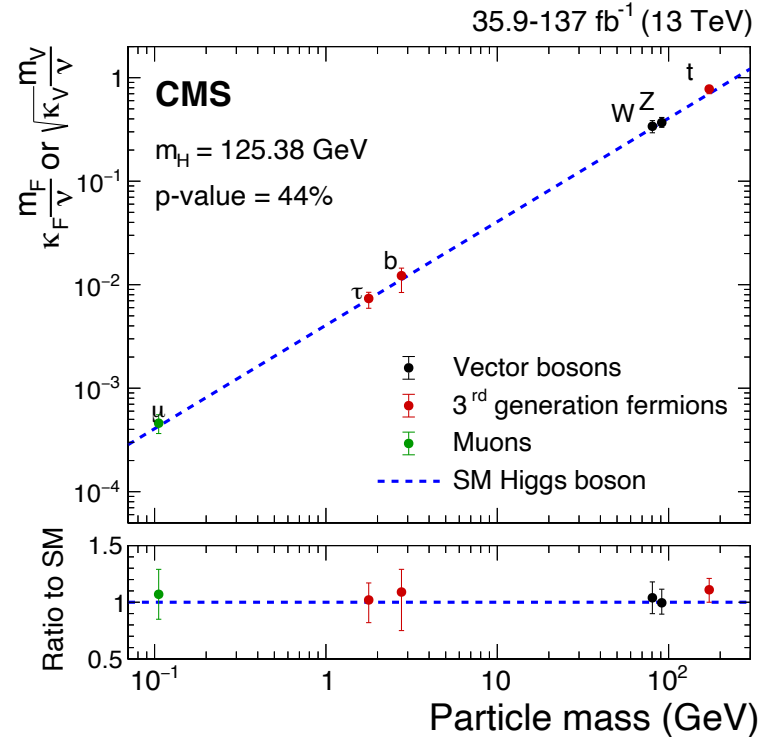
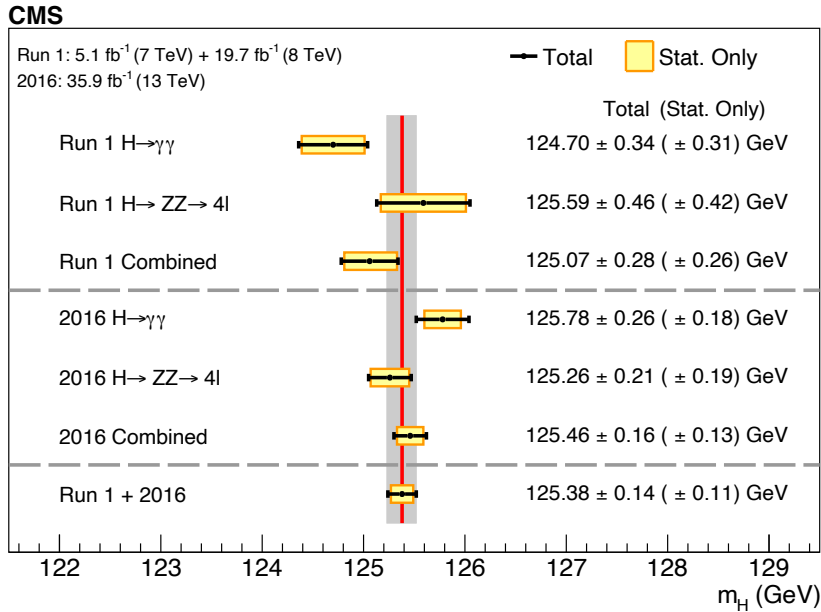
Introduction

- Discovery of the Higgs boson in 2012: A new chapter of particle physics



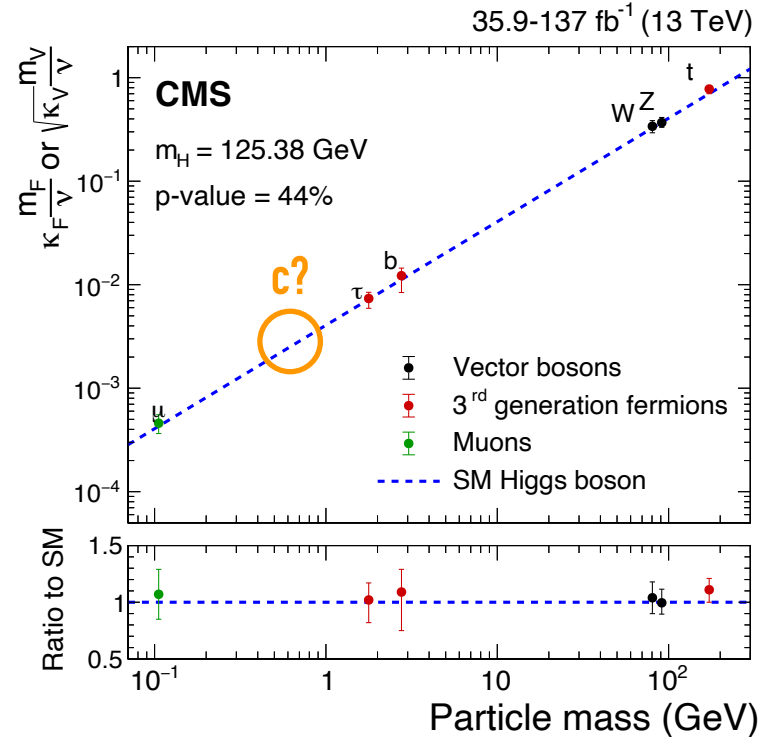
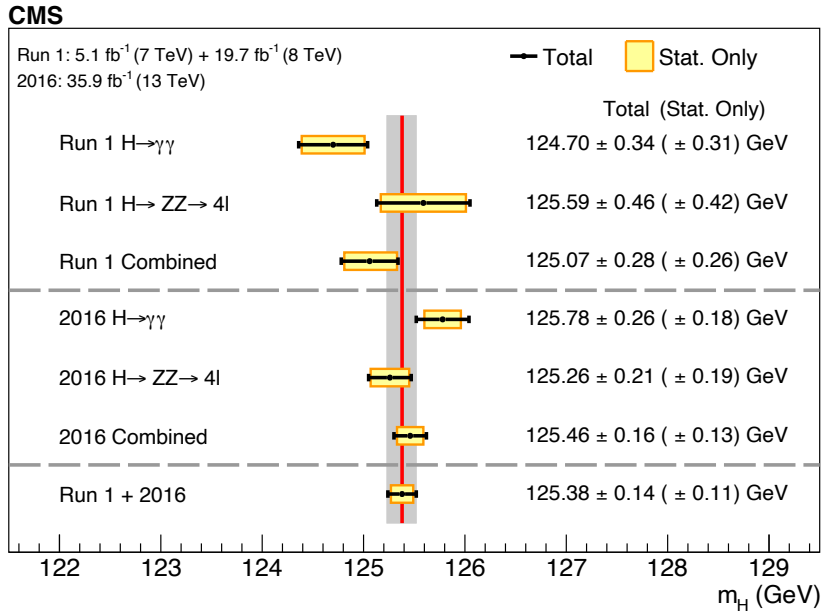
Understanding the Higgs boson

- Tremendous progress in our understanding of the Higgs boson in the past ten years



How charming is the Higgs boson?

- Tremendous progress in our understanding of the Higgs boson in the past ten years



Direct search for $H \rightarrow c\bar{c}$

❑ Search for $H \rightarrow c\bar{c}$ decays: directly sensitive to y_c , but very challenging

- small branching fraction ($\sim 3\%$) vs. large backgrounds (at a hadron collider)
- charm quark identification is the key

❑ Exploit associated VH production ($V = W, Z$)

- three channels: $Z \rightarrow \nu\nu$ (0L), $W \rightarrow \ell\nu$ (1L), $Z \rightarrow \ell\ell$ (2L) [$\ell = e, \mu$]

❑ Main backgrounds

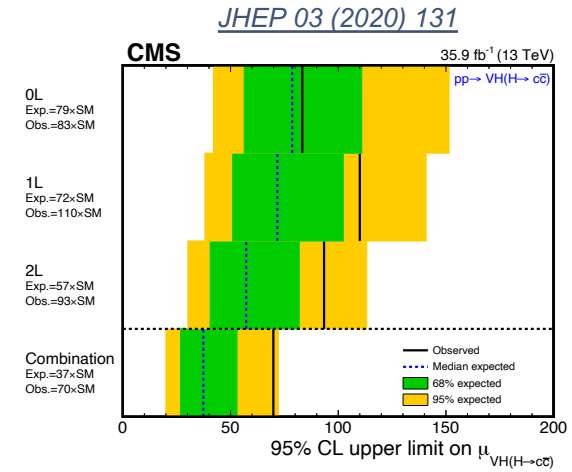
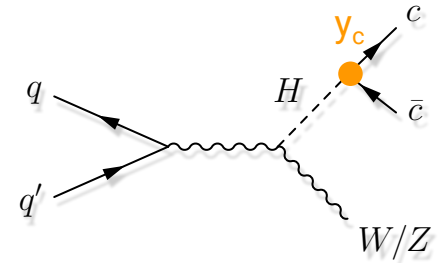
- V + jets, single and pair production of top quarks, dibosons
- $VH(H \rightarrow b\bar{b})$: small but largely irreducible

❑ Baseline event selections

- (high- p_T) vector boson recoiling against a Higgs boson candidate
- veto events with high jet multiplicity to suppress $t\bar{t}$ contribution (0L & 1L)

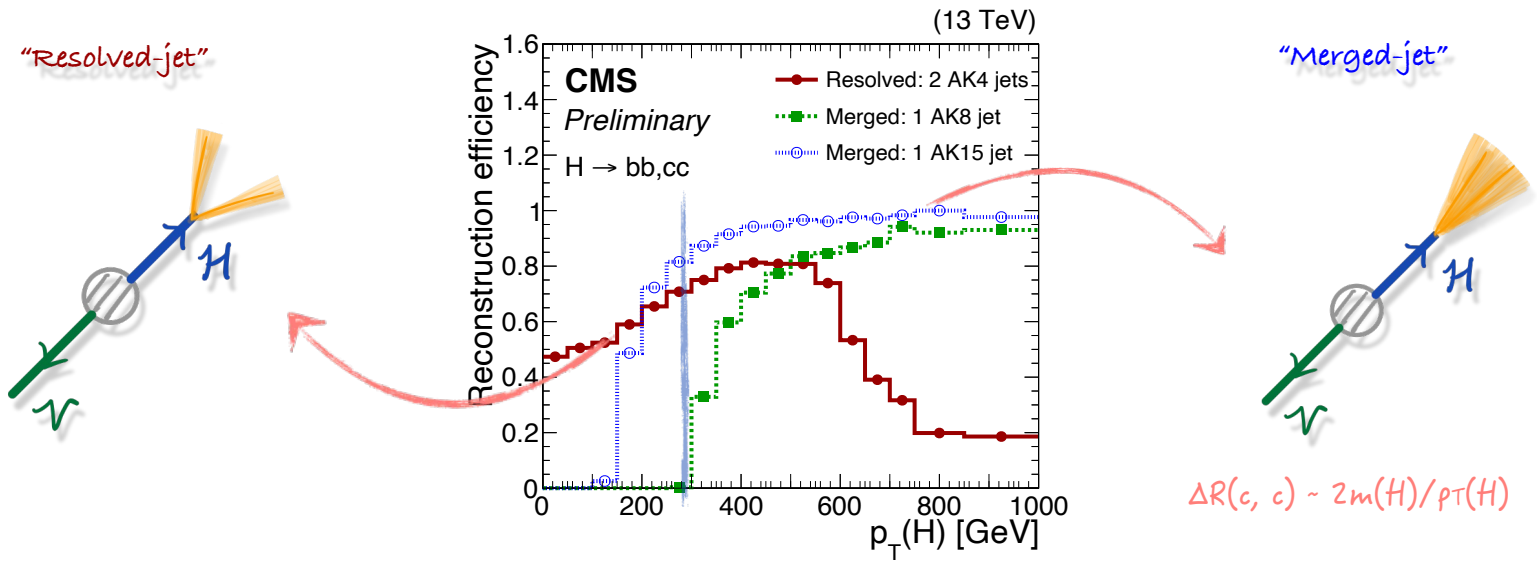
❑ Previous result (36 fb^{-1}): [JHEP 03 \(2020\) 131](#)

❑ Today: result with the full Run 2 data set (138 fb^{-1}) [arXiv:2205.05550](#) (submitted to PRL)



Analysis overview

- Two complementary approaches for Higgs boson candidate reconstruction



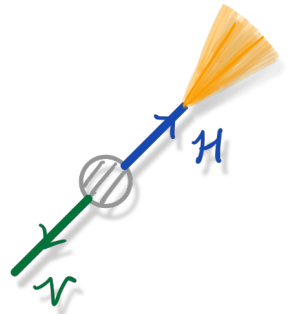
- Resolved-jet topology**

- reconstructs $H \rightarrow cc$ decay with two small-R jets ($R=0.4$, "AK4")
- probes the bulk (>95%) of the signal phase space

- Merged-jet topology**

- reconstructs $H \rightarrow cc$ decay with one large-R jets ($R=1.5$, "AK15")
- small signal acceptance (<5%) but higher purity
- better exploits the correlation between the two charm quarks

Merged-jet topology



H → cc identification

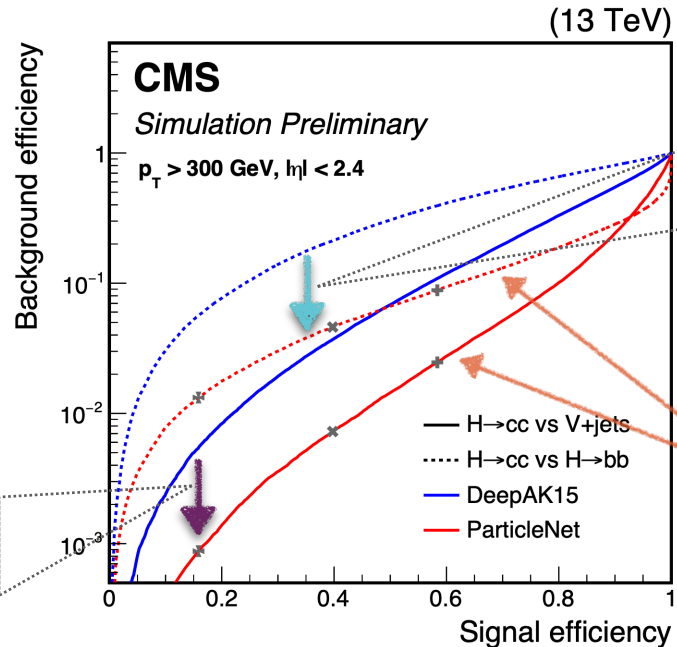
- ❑ Merged-jet topology: Higgs boson candidate reconstructed via a single large-R jet ($p_T > 300$ GeV)
- ❑ A major improvement: **ParticleNet** tagger used to identify H → cc decay

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

[DeepAK8 \(DeepAK15\)](#) [INST 15 (2020) P06005]

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

~5x better
V+jet rejection



~5x better
H → 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**

>2x improvement in the final sensitivity

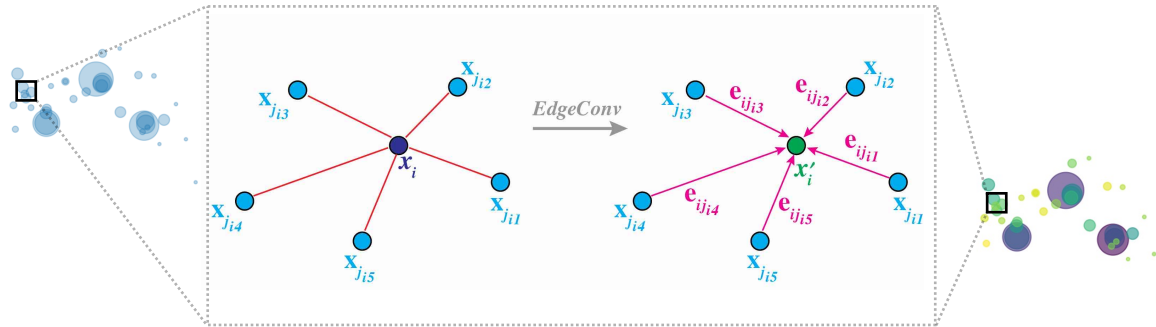
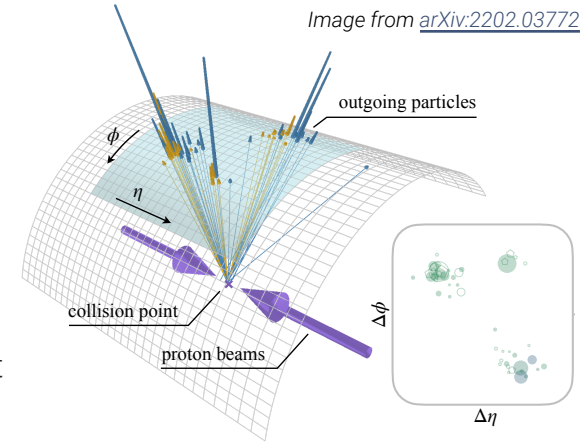
ParticleNet architecture

□ New jet representation: “particle cloud”

- treating a jet as an unordered set of particles, distributed in the $\eta - \phi$ space

□ ParticleNet [H. Qu and L. Gouskos, [Phys.Rev.D 101 \(2020\) 5, 056019](#)]

- graph neural network architecture adapted from DGCNN [[arXiv:1801.07829](#)]
- permutation-invariant architecture leads to significant performance improvement

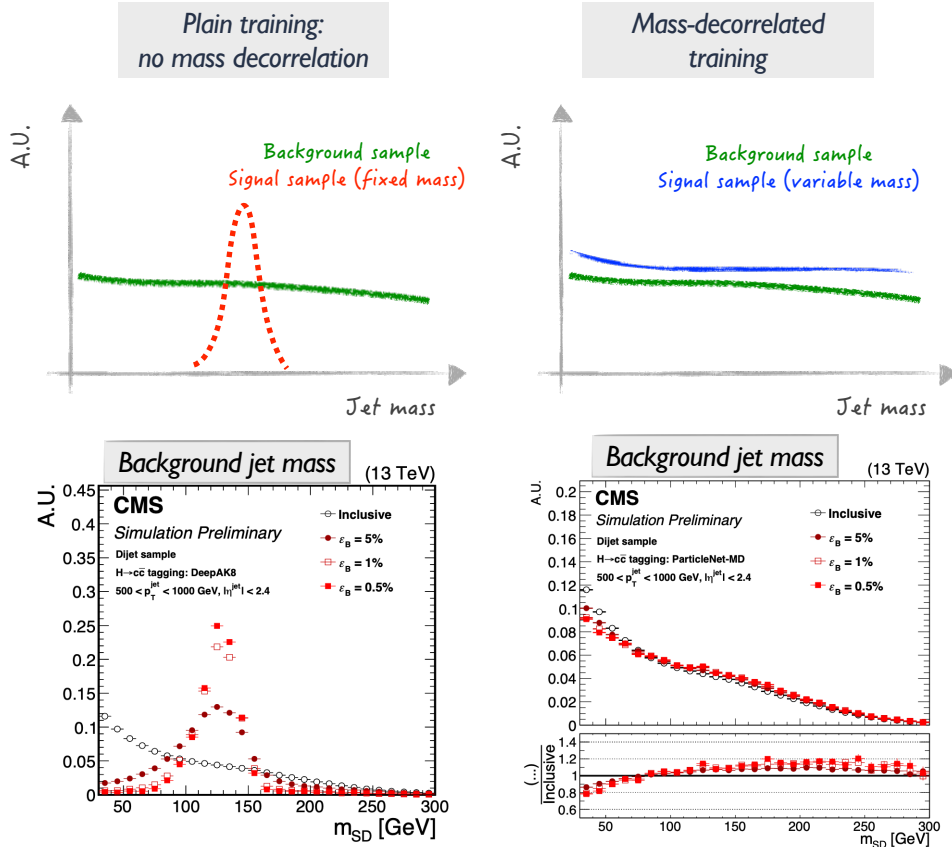


Performance on top quark tagging benchmark
[[SciPost Phys. 7, 014 \(2019\)](#)]

	$1/\epsilon_b$ at $\epsilon_s = 30\%$
ResNeXt-50	1147 ± 58
P-CNN	759 ± 24
PFN	888 ± 17
ParticleNet-Lite	1262 ± 49
ParticleNet	1615 ± 93

Mass decorrelation

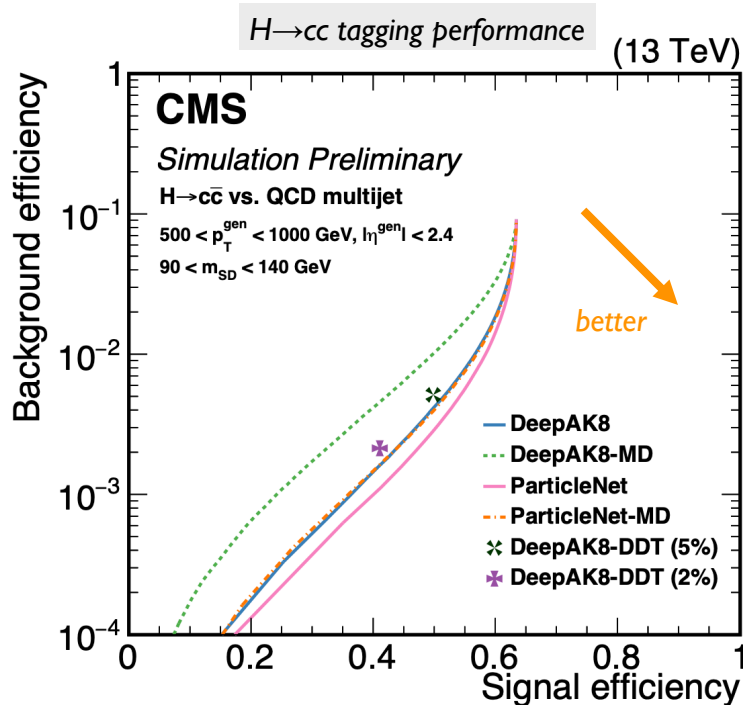
CMS-DP-2020-002



- ❑ “Mass sculpting”: background jet mass shape becomes similar to signal after tagger selection
- ❑ New approach to prevent mass sculpting
 - using a special signal sample for training
 - hadronic decays of a spin-0 particle X
 - $X \rightarrow b\bar{b}$, $X \rightarrow c\bar{c}$, $X \rightarrow q\bar{q}$
 - not a fixed mass, but a **flat mass spectrum**
 - $m(X) \in [15, 250]$ GeV
 - allows to easily reweight both signal and background to a \sim flat 2D distribution in (p_T, mass) for the training
- ❑ Signal and background have the same (\sim flat) mass spectrum, thus no sculpting will develop in the training

Mass decorrelation (II)

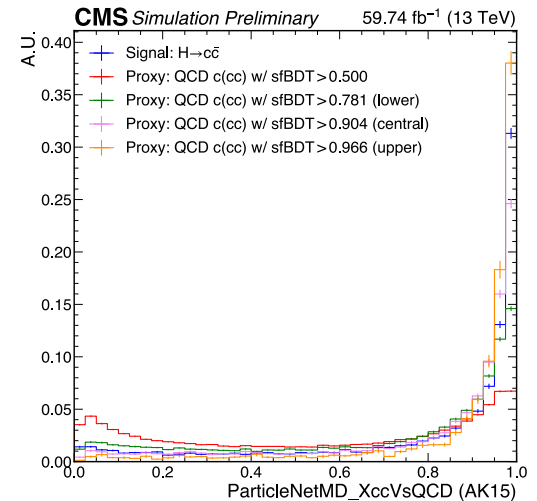
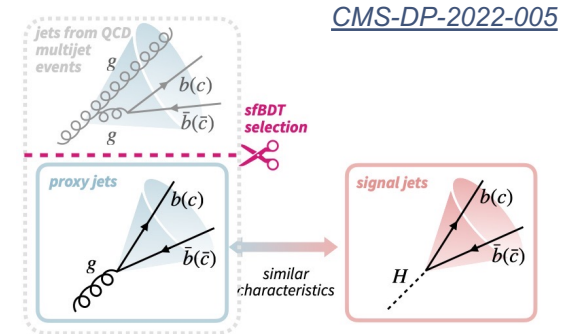
CMS-DP-2020-002



- ❑ “Mass sculpting”: background jet mass shape becomes similar to signal after tagger selection
- ❑ New approach to prevent mass sculpting
 - using a special signal sample for training
 - hadronic decays of a spin-0 particle *X*
 - $X \rightarrow b\bar{b}$, $X \rightarrow c\bar{c}$, $X \rightarrow q\bar{q}$
 - not a fixed mass, but a **flat mass spectrum**
 - $m(X) \in [15, 250]$ GeV
 - allows to easily reweight both signal and background to a \sim flat 2D distribution in (p_T , mass) for the training
- ❑ Performance loss due to mass decorrelation greatly reduced compared to the previous approach (DeepAK8-MD, based on “adversarial training”)

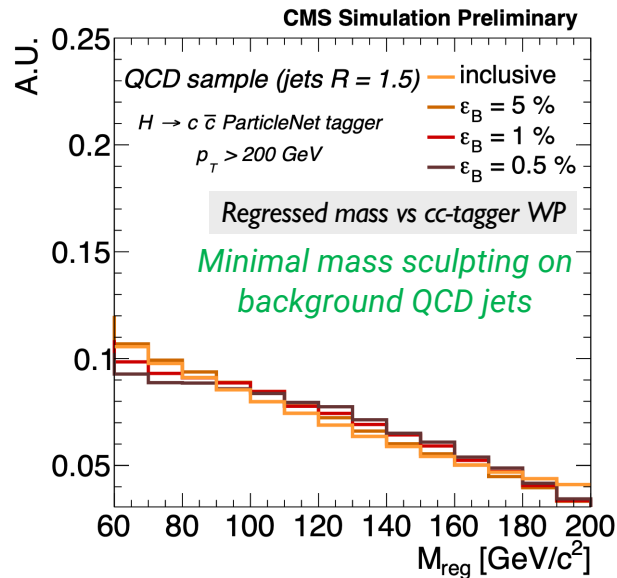
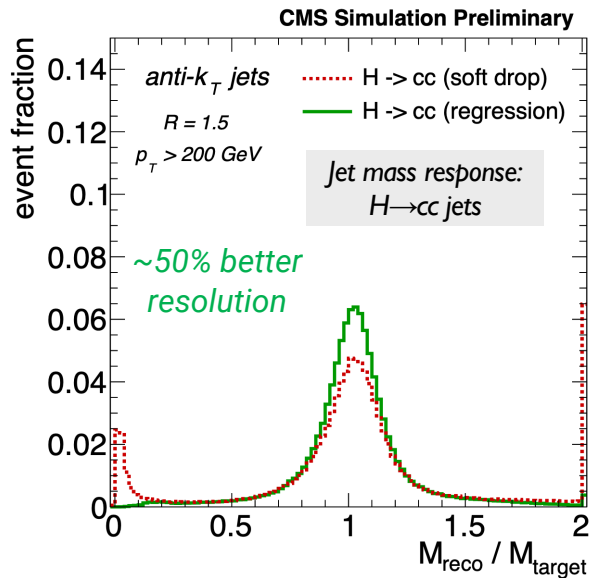
Calibration of the cc-tagger

- ❑ Need to measure ParticleNet cc-tagging efficiency in data
 - no pure sample of $H \rightarrow cc$ jets (or even $Z \rightarrow cc$) in data
 - using $g \rightarrow cc$ in QCD multi-jet events as a proxy
- ❑ Difficulty: select a phase-space in $g \rightarrow cc$ that resembles $H \rightarrow cc$
 - solution: a **dedicated BDT** developed to distinguish **hard 2-prong splittings** (i.e., high quark contribution to the jet momentum) from **soft cc radiations** (i.e., high gluon contribution to the jet momentum)
 - also allows to adjust the similarity between proxy and signal jets
 - by varying the sfBDT cut – treated as a systematic uncertainty
- ❑ Perform a fit to the secondary vertex mass shapes in the “passing” and “failing” regions simultaneously to extract the scale factors
 - three templates: cc (+ single c), bb (+ single b), light flavor jets
- ❑ Derived cc-tagging scale factors typically 0.9–1.3
 - corresponding uncertainties are 20–30%



Large-R jet mass regression

- ❑ Jet mass: one of the most powerful observable to distinguish signal and backgrounds [CMS DP-2021/017](#)
- ❑ New ParticleNet-based regression algorithm to improve the large-R jet mass reconstruction
 - training setup similar to the ParticleNet tagger; the regression target:
 - signal ($X \rightarrow bb/cc/qq$): generated particle mass of X [flat spectrum in 15 – 250 GeV]
 - background (QCD) jets: soft drop mass of the particle-level jet



**20 – 25% improvement
in the final sensitivity**

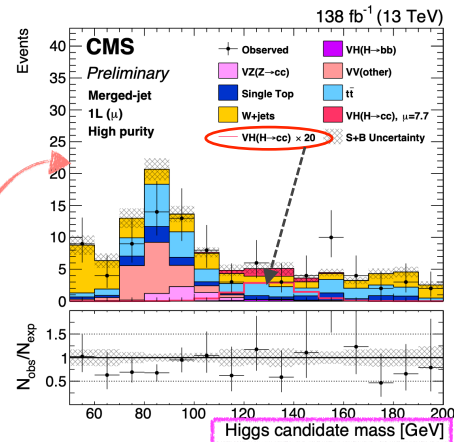
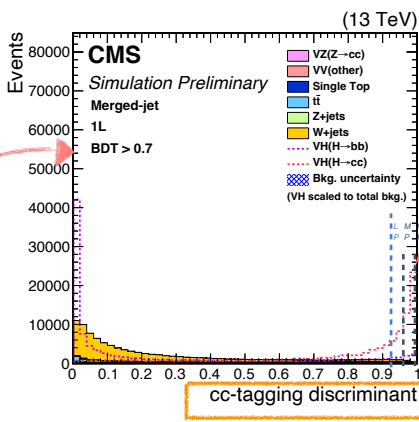
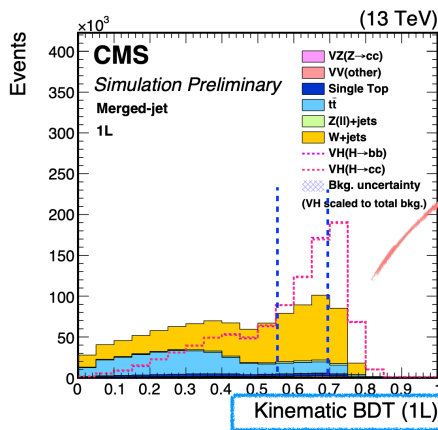
Analysis strategy

Factorized approach for analysis design

- event-level **kinematic BDT** developed in each channel to better suppress main backgrounds (V+jets, tt)
 - using only *event kinematics*, no intrinsic properties (e.g., mass/flavor) of the large-R jet
- ParticleNet cc-tagger** then used to define 3 cc-flavor enriched regions and reject light/bb-flavor jets
- finally: fit to the **ParticleNet-regressed large-R jet mass** shape for signal extraction

Kinematic BDT, ParticleNet cc-tagger and regressed jet mass largely independent of each other

- allowing for a simple and robust strategy for background estimation and signal extraction



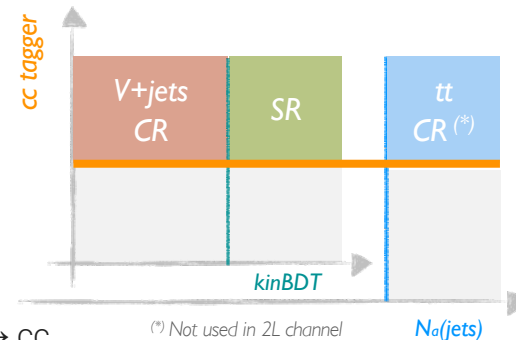
Background estimation

□ Normalizations of main backgrounds estimated via dedicated data control regions (CRs)

- V+jets CR: use the low kinematic BDT region
- tt CR (0L & 1L): invert the cut on the number of additional small-R jets (i.e., $N_{aj} \geq 2$)
- free-floating parameters scale the normalizations in CRs and signal regions (SRs) simultaneously

□ CRs designed to have similar jet flavor composition as the SR

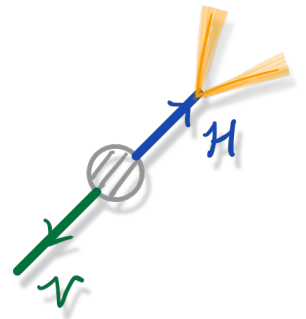
- flavor-independent kinematic BDT + same cc-tagging requirement in CRs as in SR
- allows to correct cc-tagging efficiency for backgrounds directly from data
- cc-tagging SFs only needed for the signal $VH(H \rightarrow cc)$ process (and $VZ(Z \rightarrow cc)$)
 - conservative uncertainty (2x/0.5x) for the misidentification of $H(Z) \rightarrow bb$ as $H(Z) \rightarrow cc$



□ Minor backgrounds (single top, dibosons, $VH(H \rightarrow bb)$) estimated from simulation

- dibosons: applying differential NNLO QCD + NLO EW corrections as a function of $p_T(V)$ [[JHEP 2002 \(2020\) 087](#)]

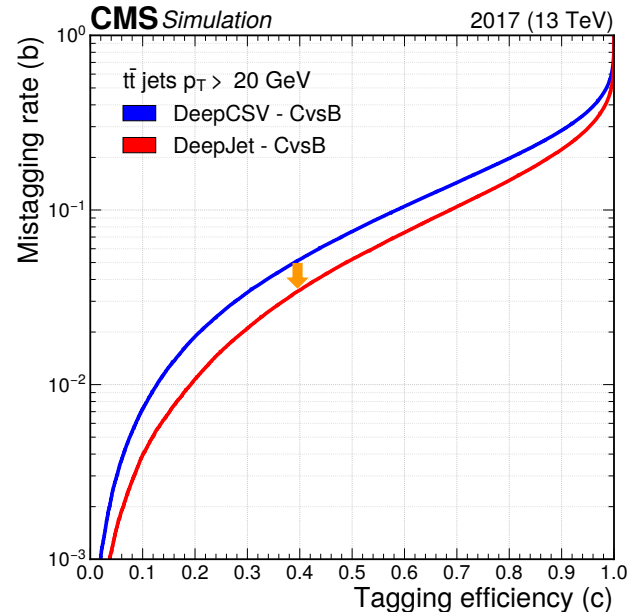
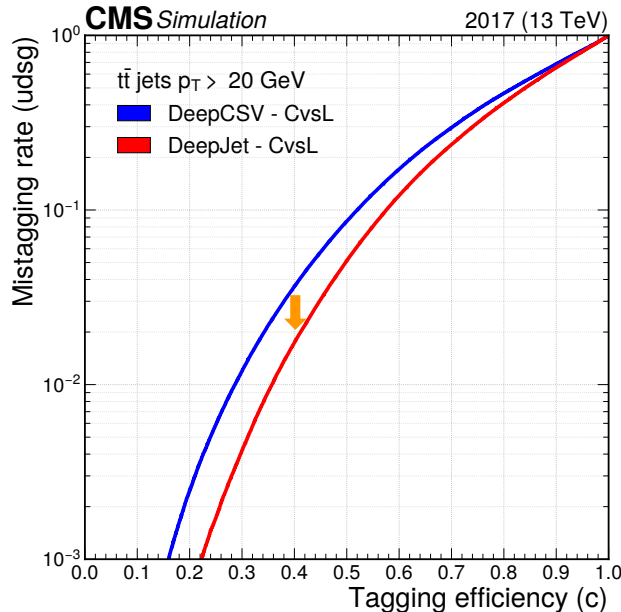
Resolved-jet topology



Charm quark identification

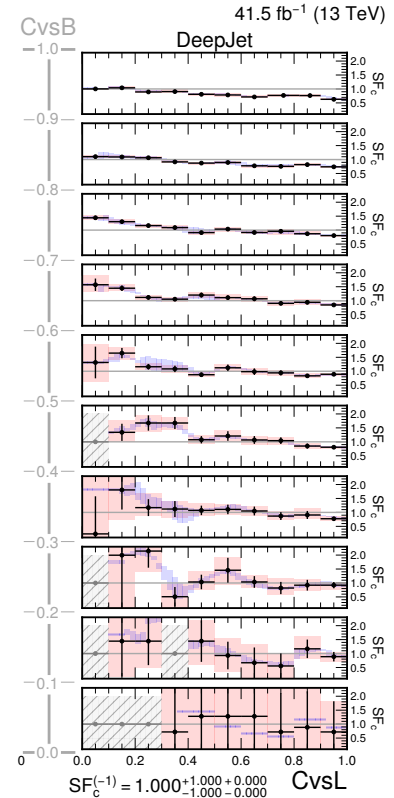
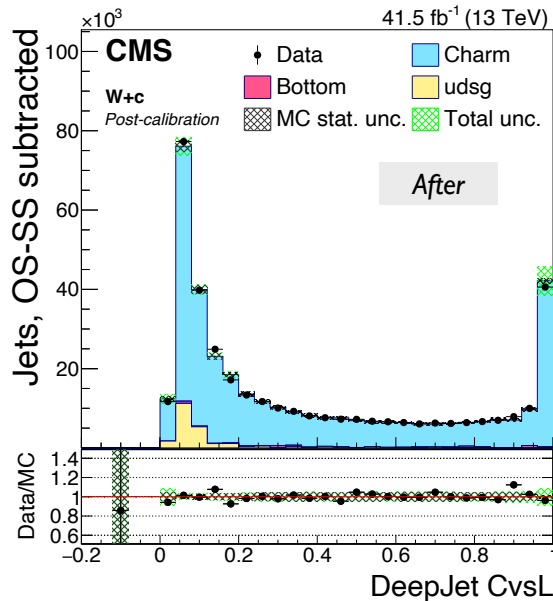
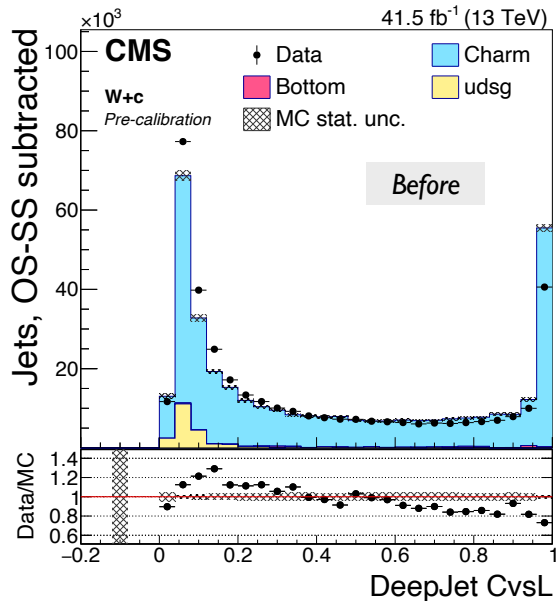
- Resolved-jet topology: Higgs boson candidate reconstructed with two small-R jets
- Charm quark jet identification: **DeepJet** algorithm
 - $\sim 2\times$ ($\sim 40\%$) improvement in light (b) jet rejection at 40% c jet efficiency compared to DeepCSV

JINST 17 (2022) P03014



Charm tagging calibration

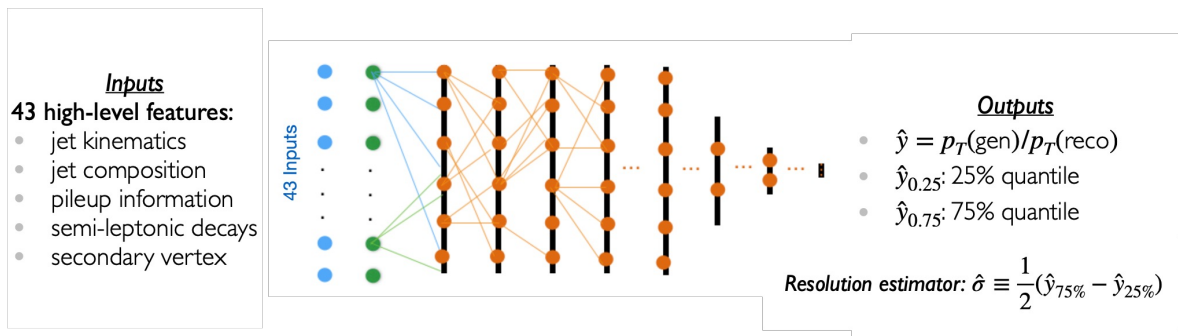
- Novel calibration method to correct the entire distributions of the c-tagging discriminants
 - per-jet SFs derived as a function of (CvsL, CvsB | truth flavor)
- SFs derived with an iterative approach using three samples
 - $Z(\ell\ell)$ + jets (light jet enriched); $W + c$ (c-jet enriched); $t\bar{t}$ (b-jet enriched)



Charm jet energy regression

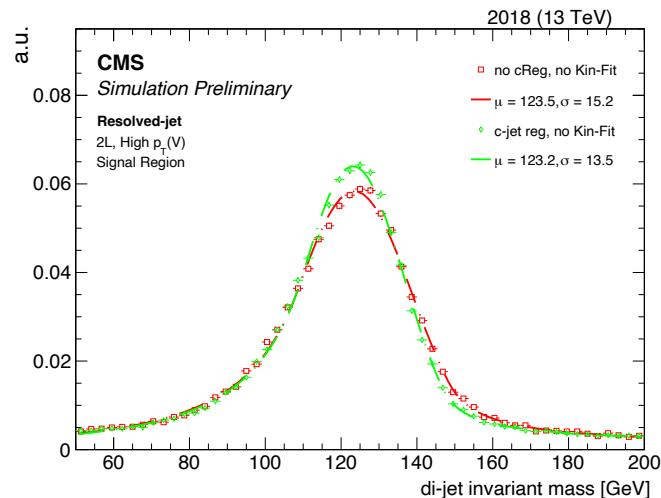
□ Dedicated jet energy regression algorithm developed to improve the c-jet energy scale and resolution

- based on the b jet energy regression [[Comput.Softw.Big Sci. 4 \(2020\) 10](#)] used in several CMS H→bb analyses
- re-trained for c jets instead of b jets
 - c jets collected from W→cx decay in tt MC events
- provides simultaneous estimation of the c jet energy and its resolution
 - both used as inputs to the signal extraction BDT



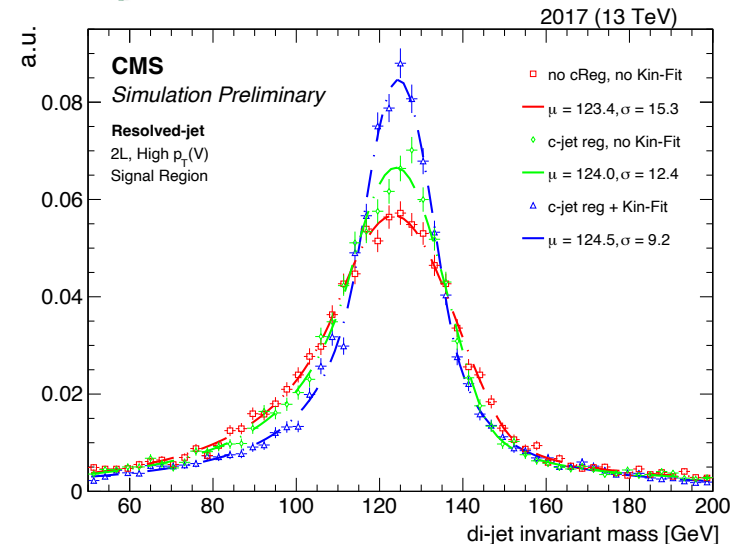
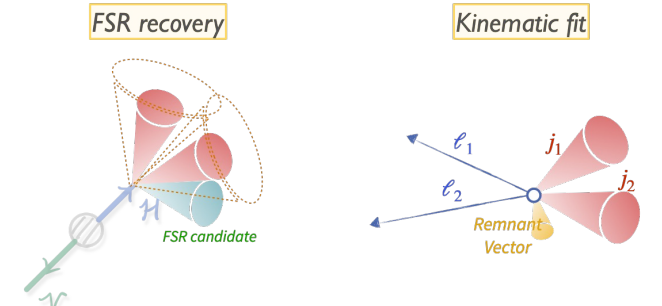
Joint loss function for correction (Huber) and resolution (quantiles):

$$\text{Loss} = \text{Huber}(y, F(x)) + \rho_{0.75}(y - F(x)) + \rho_{0.25}(y - F(x))$$



Higgs boson candidate reconstruction

- ❑ Higgs boson candidate reconstructed using the two small-R jets with highest CvsL scores
- ❑ To improve the Higgs candidate mass resolution:
 - recovery of final state radiations (FSR)
 - additional jets within $\Delta R < 0.8$ from either of the two selected jets are included in the calculation of the Higgs candidate's 4-momentum
 - improves the Higgs mass resolution by a few percent
 - new DNN-based c-jet energy regression
 - ~20% improvement in Higgs candidate mass resolution
 - improved kinematic fit in the 2L channel
 - better reconstruction of the Higgs candidate's 4-momentum using constraints from the $Z \rightarrow \ell\ell$ system
 - up to 30% improvement in Higgs candidate mass resolution



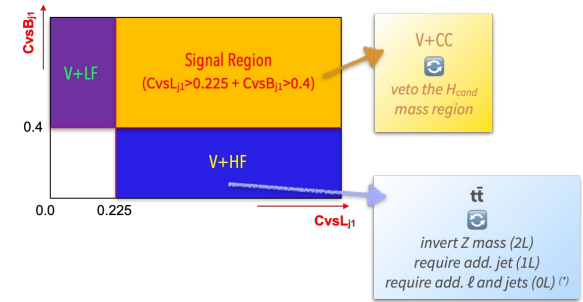
Signal extraction strategy

Event-level BDT trained in each channel to maximize the signal vs background separation

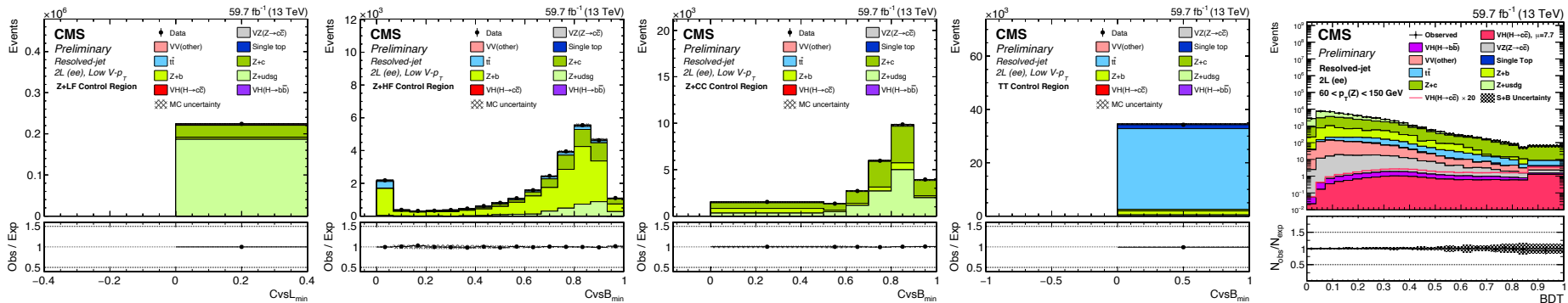
- inputs: event kinematics, Higgs candidate properties, c-tagging discriminants
 - + kinematic fit variables in 2L

Background estimation

- dedicated CRs to constrain the normalizations of main backgrounds (V + jets, tt)
 - V + jets split based on flavor: V + b, V + c, V + udsg



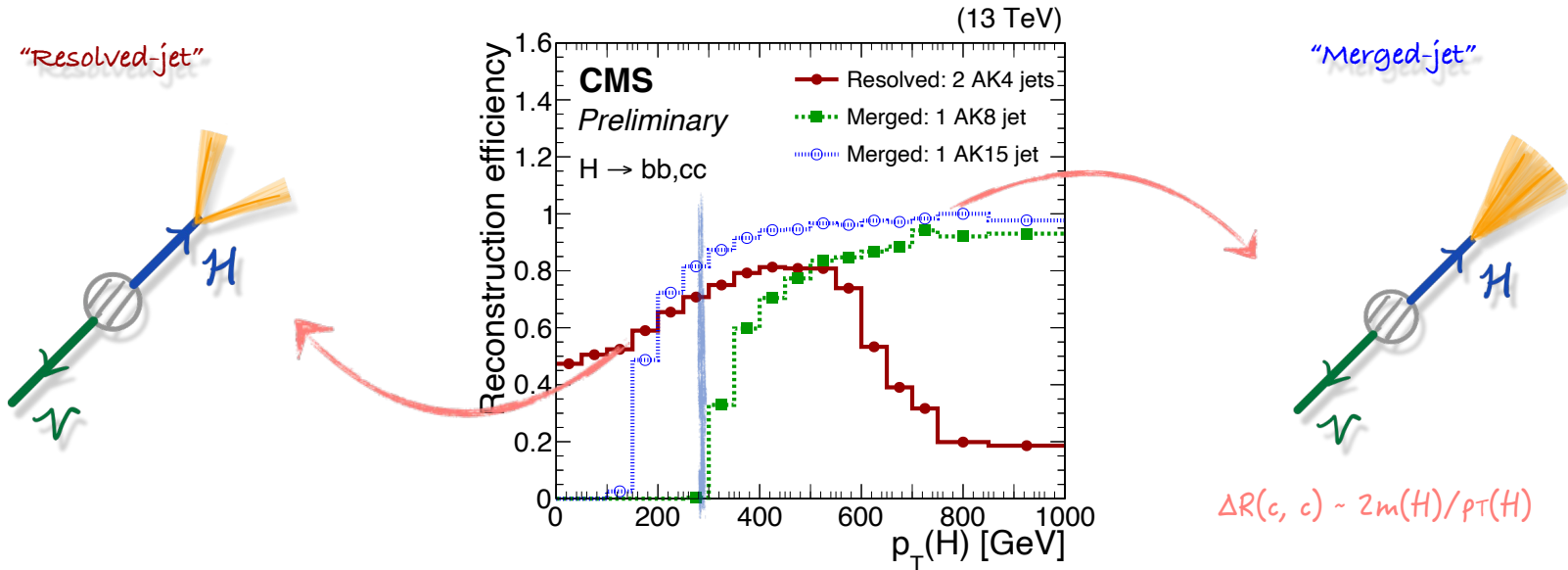
Simultaneous fit of SRs (BDT shapes) and CRs (c-tagging discriminants) for signal extraction



Results

Combination of the two topologies

- The two topologies are made orthogonal via the presence of large- R jet with $p_T > 300$ GeV
 - p_T threshold chosen to maximize expected sensitivity



Uncertainties

❑ Systematic uncertainties correlated between topologies, except:

- background normalizations for V+jets and tt
- charm quark identification efficiencies

❑ Main uncertainties

- limited statistics of the data set
- size of simulated samples (especially NLO V+jets)
- charm quark identification efficiencies

Relative contributions to the total uncertainty on μ

Uncertainty source	$\Delta\mu / (\Delta\mu)_{\text{tot}}$
Statistical	85%
Background normalizations	37%
Experimental	48%
Sizes of the simulated samples	37%
Charm identification efficiencies	23%
Jet energy scale and resolution	15%
Simulation modeling	11%
Luminosity	6%
Lepton identification efficiencies	4%
Theory	22%
Backgrounds	17%
Signal	15%

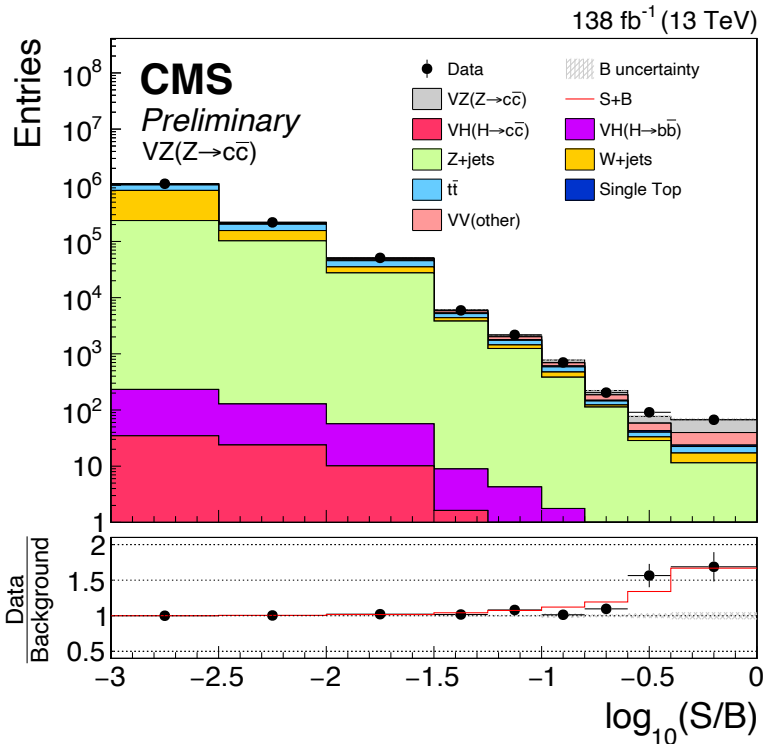
VZ($Z \rightarrow cc$) results

- The full analysis procedure is validated by measuring the VZ($Z \rightarrow cc$) process
 - resolved-jet topology:
 - BDT re-trained using VZ($Z \rightarrow cc$) as signal
 - fit to the BDT shapes to extract the signal
 - merged-jet topology:
 - no change to the analysis procedure
 - fit to the large-R jet mass shapes to extract the signal

VZ(Z → cc) results

□ The full analysis procedure is validated by measuring the VZ(Z → cc) process

[arXiv:2205.05550](https://arxiv.org/abs/2205.05550)



Observed significance for VZ(Z → cc): **5.7σ**

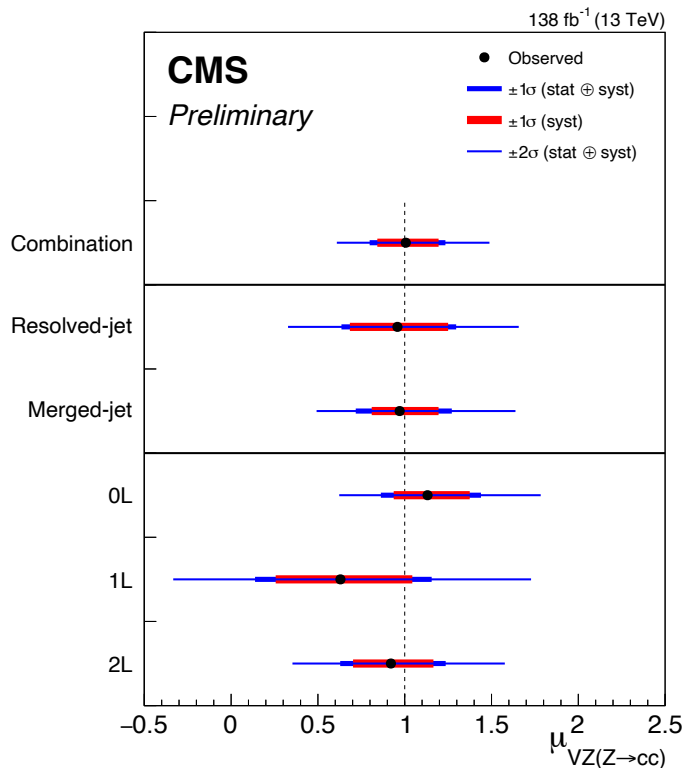
- expected significance: 5.9σ

First observation of Z → cc at a hadron collider!

VZ(Z → cc) results

- The full analysis procedure is validated by measuring the VZ(Z → cc) process

[arXiv:2205.05550](https://arxiv.org/abs/2205.05550)



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

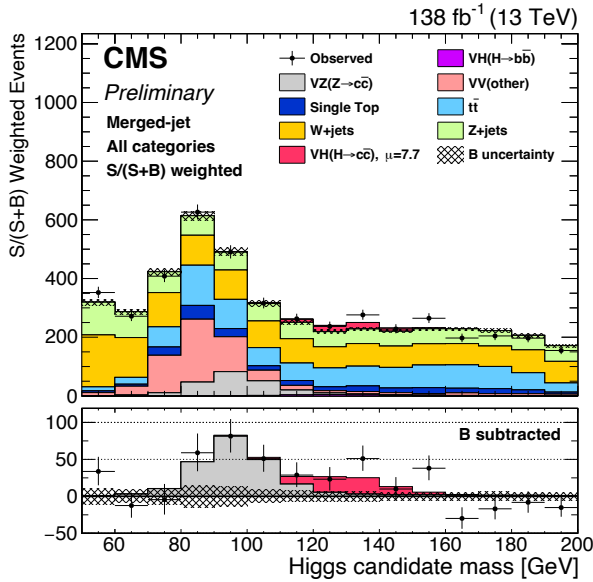
- very good agreement with SM expectation
- consistent results between topologies/channels

VH(H → cc) results

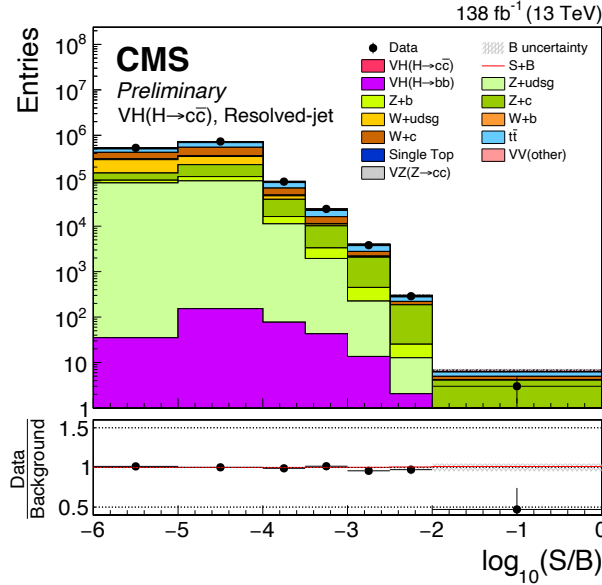
Post-fit distributions in the two topologies and the combination

[arXiv:2205.05550](https://arxiv.org/abs/2205.05550)

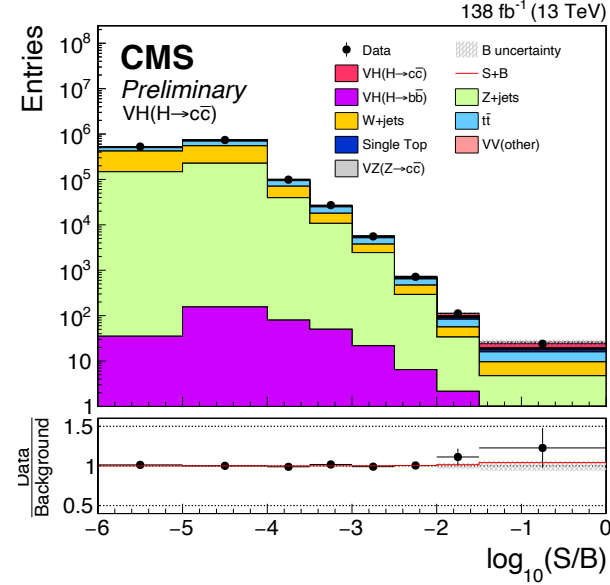
Merged-jet topology



Resolved-jet topology



Combination



VH(H → cc) results

Upper limits on the VH(H → cc) signal strength at 95% CL:

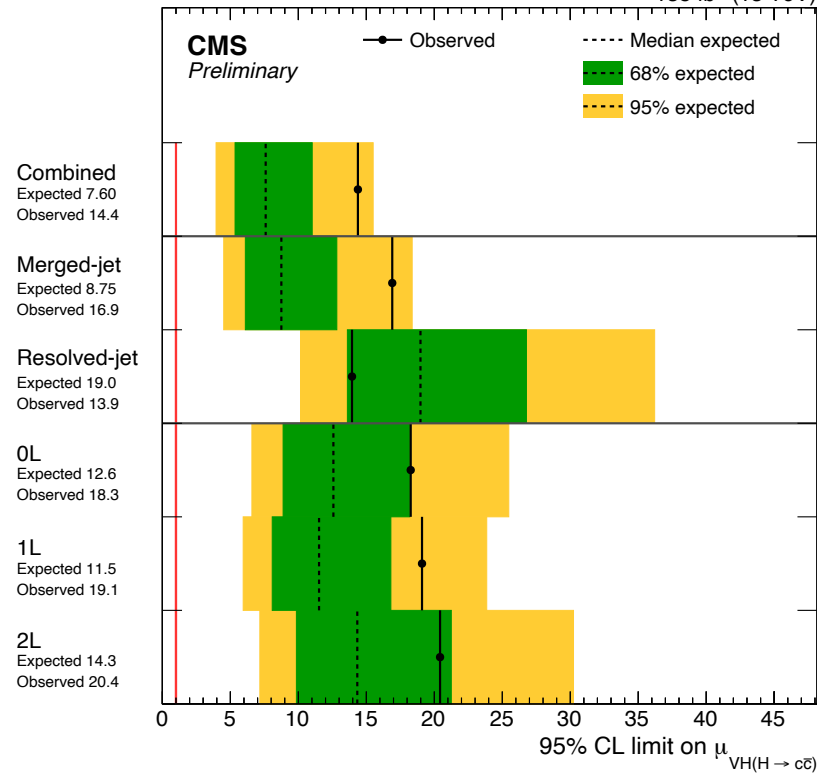
- $\mu_{VH(H \rightarrow cc)} < 14$ (7.6) observed (expected)
- substantially stronger than ATLAS full Run 2 result
 - $\mu_{VH(H \rightarrow cc)} < 26$ (31) obs. (exp.) [arXiv:2201.11428]

Best fit signal strength

- $\mu_{VH(H \rightarrow cc)} = 7.7^{+3.8}_{-3.5}$
- consistent with the SM prediction within 2 σ

arXiv:2205.05550

138 fb⁻¹ (13 TeV)



VH(H → cc) results

□ Results used to set a constraint on the charm quark Yukawa coupling modifier $\kappa_c = y_c / y_c^{\text{SM}}$

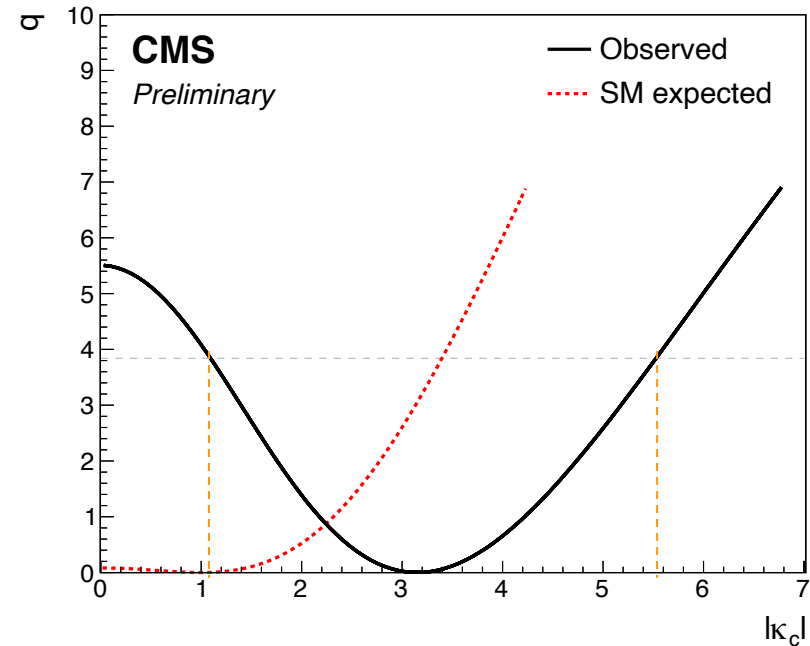
- for simplicity, only considering effects on $B(H \rightarrow cc)$ and fixing all other couplings at SM values:

$$\mu_{\text{VH}(H \rightarrow c\bar{c})} = \frac{\kappa_c^2}{1 + \mathcal{B}_{\text{SM}}(H \rightarrow c\bar{c}) \times (\kappa_c^2 - 1)}$$

□ The 95% CL interval on κ_c :

- observed: $1.1 < |\kappa_c| < 5.5$
- expected: $|\kappa_c| < 3.4$
- **most stringent constraint on κ_c to date!**
 - comparable to the previous projection for HL-LHC w/ 3000 fb^{-1} : $|\kappa_c| < 3.0$ [ATL-PHYS-PUB-2021-039]

[arXiv:2205.05550](https://arxiv.org/abs/2205.05550)



Prospects: HL-LHC

Projection at HL-LHC: Setup

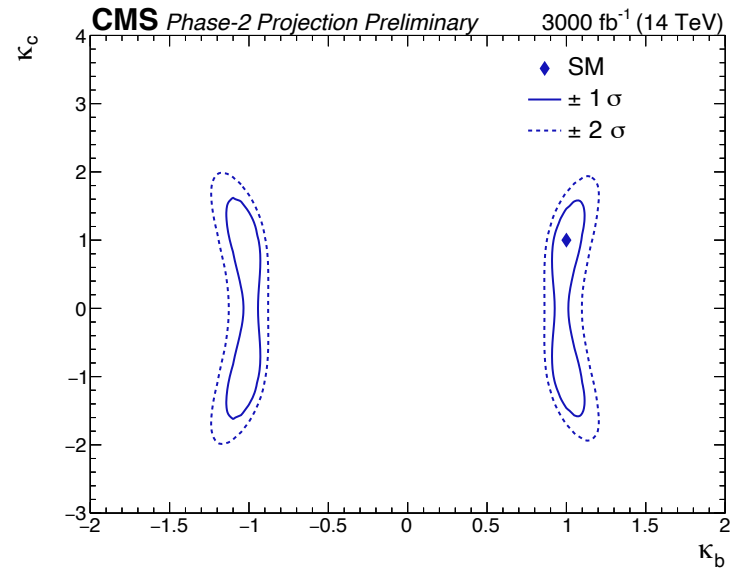
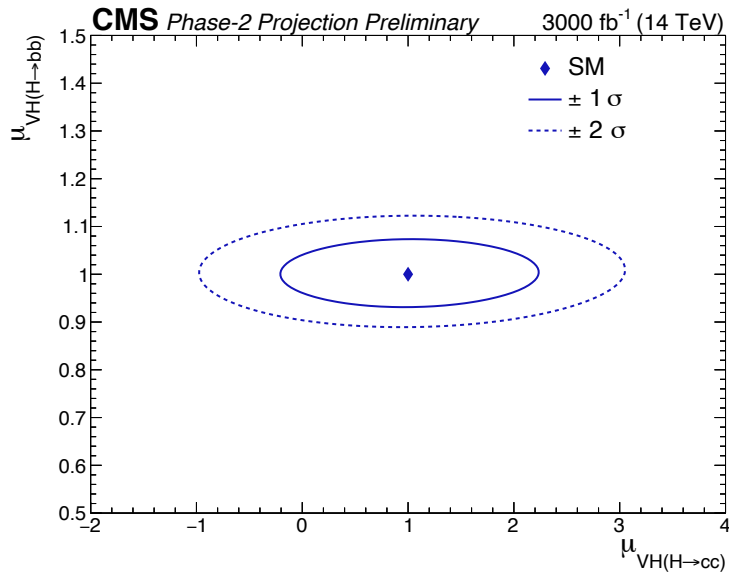
- ❑ Extrapolation of the merged-jet analysis to HL-LHC with 3000 fb^{-1} data
- ❑ Modifications to the Run 2 analysis to allow for a simultaneous constraint on $H \rightarrow bb$ and $H \rightarrow cc$
 - addition of 3 categories enriched in $H \rightarrow bb$ decays, selected with the ParticleNet bb-tagging discriminant
 - very small (1-2%) overlap of bb and cc categories – events assigned to a unique category
 - large-R jet p_T threshold lowered from 300 GeV to 200 GeV – increasing signal acceptance
- ❑ Systematic uncertainties adjusted according to the Yellow Report [[CERN-2019-007](#)]
 - theoretical uncertainties: reduced by half
 - most experimental uncertainties: scaled down with $\sqrt{\mathcal{L}}$
 - bb and cc tagging efficiencies: constrained by $VZ(Z \rightarrow bb)$ and $VZ(Z \rightarrow cc)$ events to $\sim 3\%$ and $\sim 5\%$
 - misidentification of $H \rightarrow bb$ as $H \rightarrow cc$: a prominent uncertainty on $H \rightarrow cc$ measurement at HL-LHC
 - assumed to be reduced from $\sim 100\%$ (Run 2) to 20% in the projection

Projection at HL-LHC

□ Simultaneous extraction of the $H \rightarrow bb$ and $H \rightarrow cc$ signal strengths

CMS-PAS-HIG-21-008

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



Expected sensitivity approaches the SM value for the Higgs-charm coupling.

Prospects: ML for jet tagging

Beyond ParticleNet

Particle Transformer (ParT): a new Transformer-based architecture for jet tagging

- Transformer + *interaction*

Interaction features

$$\Delta = \sqrt{(y_a - y_b)^2 + (\phi_a - \phi_b)^2},$$

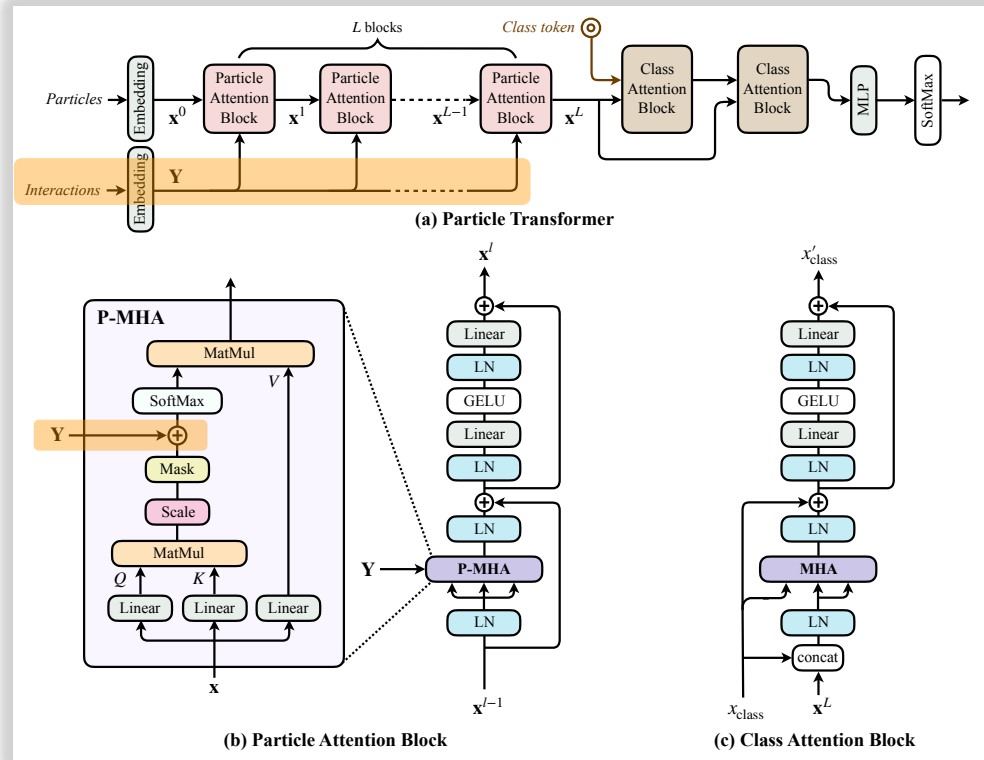
$$k_T = \min(p_{T,a}, p_{T,b})\Delta,$$

$$z = \min(p_{T,a}, p_{T,b}) / (p_{T,a} + p_{T,b}),$$

$$m^2 = (E_a + E_b)^2 - \|\mathbf{p}_a + \mathbf{p}_b\|^2,$$

Motivated by LundNet
[F. Dreyer and H. Qu,
[JHEP 03 \(2021\) 052](#)]

H. Qu, C. Li, S. Qian,
[arXiv:2202.03772](#)
[GitHub](#)

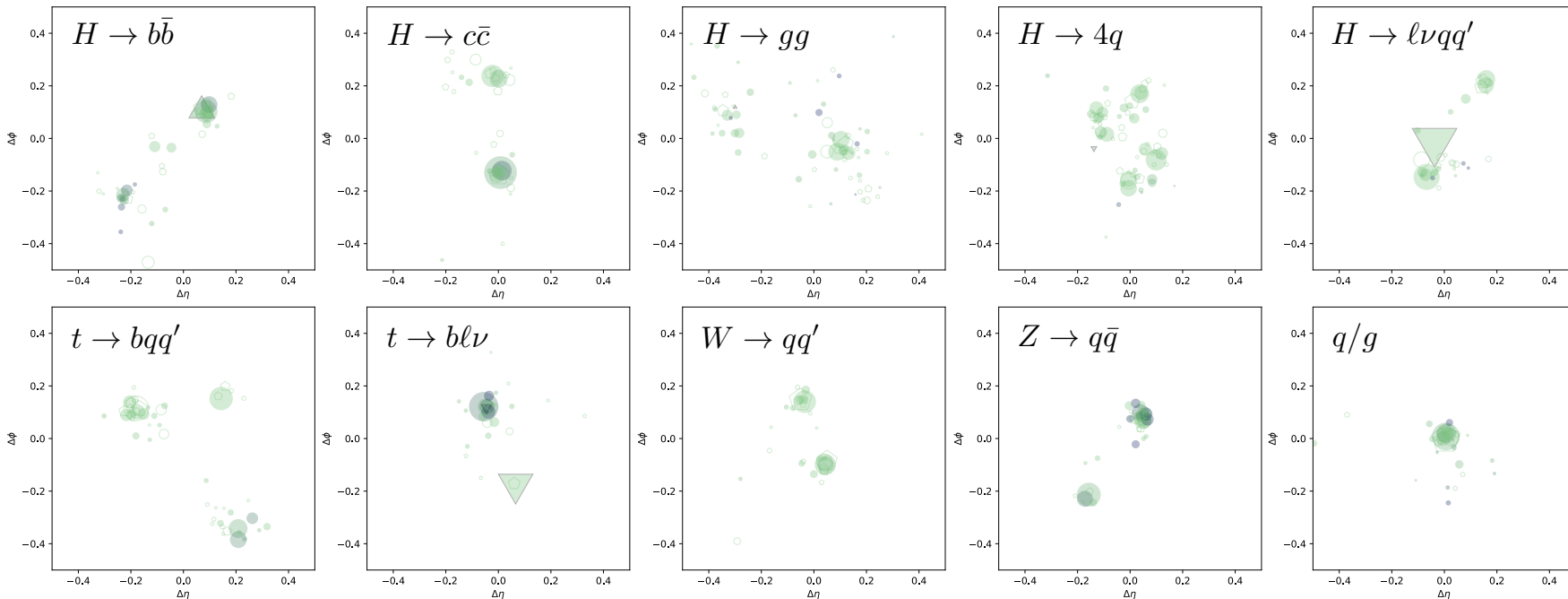


Beyond ParticleNet: JetClass dataset

□ A new large-scale public jet dataset: 100M jets in 10 classes

■ MadGraph + Pythia (Herwig) + Delphes

H. Qu, C. Li, S. Qian,
[arXiv:2202.03772](https://arxiv.org/abs/2202.03772)
[GitHub](#)



Beyond ParticleNet: Particle Transformer

Constraining Higgs-charm coupling in CMS – May 18, 2022 - Huijin Qu (CERN)

Performance comparison on the JetClass dataset

$$\text{Rej}_{50\%} = 1/\epsilon_{\text{bkg}} @ \epsilon_{\text{sig}} = 50\%$$

H. Qu, C. Li, S. Qian,
arXiv:2202.03772, [GitHub](#)

	All classes		$H \rightarrow b\bar{b}$	$H \rightarrow c\bar{c}$	$H \rightarrow gg$	$H \rightarrow 4q$	$H \rightarrow \ell\nu qq'$	$t \rightarrow bqq'$	$t \rightarrow bl\nu$	$W \rightarrow qq'$	$Z \rightarrow q\bar{q}$
	Accuracy	AUC	$\text{Rej}_{50\%}$	$\text{Rej}_{50\%}$	$\text{Rej}_{50\%}$	$\text{Rej}_{50\%}$	$\text{Rej}_{99\%}$	$\text{Rej}_{50\%}$	$\text{Rej}_{99.5\%}$	$\text{Rej}_{50\%}$	$\text{Rej}_{50\%}$
PFN	0.772	0.9714	2924	841	75	198	265	797	721	189	159
P-CNN	0.809	0.9789	4890	1276	88	474	947	2907	2304	241	204
ParticleNet	0.844	0.9849	7634	2475	104	954	3339	10526	11173	347	283
ParT	0.861	0.9877	10638	4149	123	1864	5479	32787	15873	543	402
ParT (plain)	0.849	0.9859	9569	2911	112	1185	3868	17699	12987	384	311

Fine-tuning result on Top-tagging Benchmark (~2M jets) [SciPost Phys. 7 (2019) 014]

	Accuracy	AUC	$\text{Rej}_{50\%}$	$\text{Rej}_{30\%}$
P-CNN	0.930	0.9803	201 ± 4	759 ± 24
PFN	—	0.9819	247 ± 3	888 ± 17
ParticleNet	0.940	0.9858	397 ± 7	1615 ± 93
JEDI-net (w/ $\sum O$)	0.930	0.9807	—	774.6
PCT	0.940	0.9855	392 ± 7	1533 ± 101
LGN	0.929	0.964	—	435 ± 95
rPCN	—	0.9845	364 ± 9	1642 ± 93
ParT	0.940	0.9858	413 ± 16	1602 ± 81
ParT-ft.	0.944	0.9877	691 ± 15	2766 ± 130

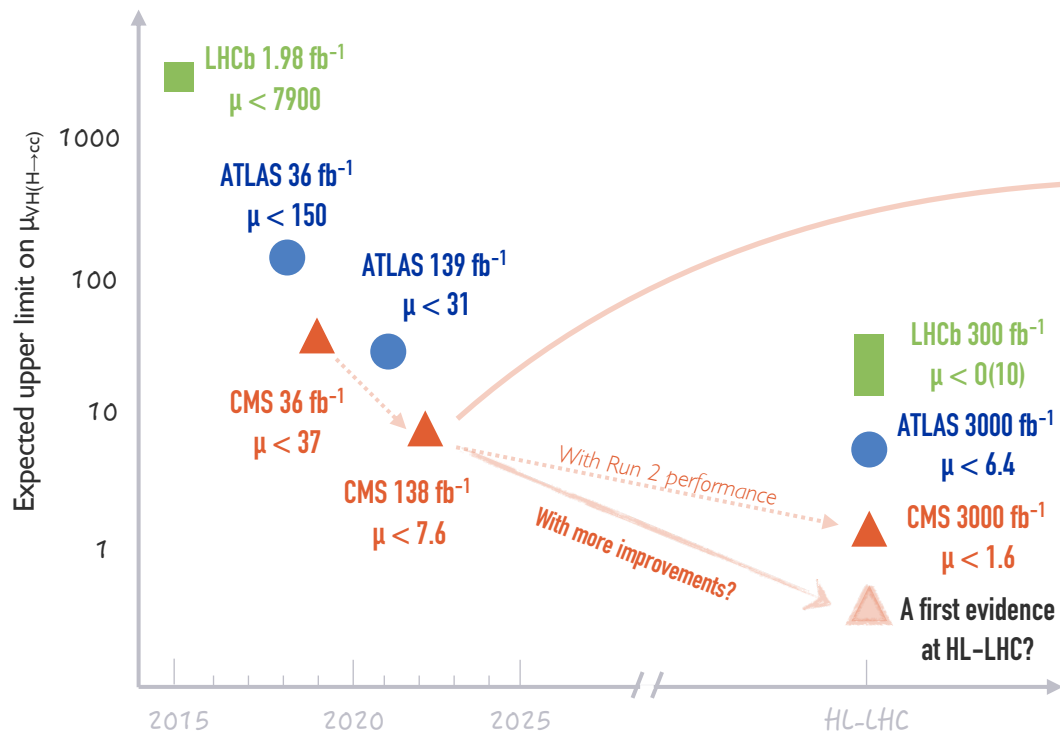
Model complexity

	Accuracy	# params	FLOPs
PFN	0.772	86.1 k	4.62 M
P-CNN	0.809	354 k	15.5 M
ParticleNet	0.844	370 k	540 M
ParT	0.861	2.14 M	340 M
ParT (plain)	0.849	2.13 M	260 M

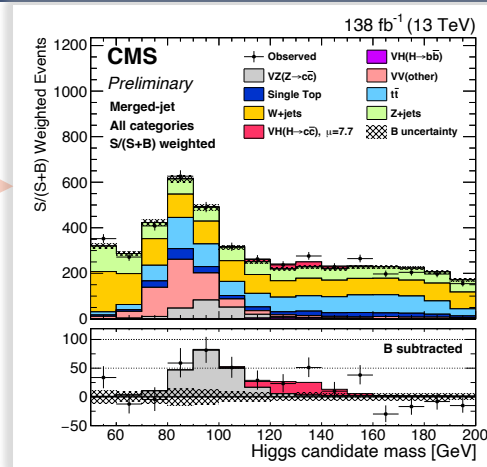
Significant performance improvement.
Similar computational cost.

Summary & Outlook

A charming journey



From $\mathcal{O}(1000)$ to $\mathcal{O}(100)$ to $\mathcal{O}(10)$ in ~ 5 years.
A combined effort and creativity from instrumentation,
physics objects and analysis techniques!



First observation of $Z \rightarrow cc$ at a hadron collider!
Opening a new era for future explorations.

- More channels: $t\bar{t}(cc)$, VBF $H(cc)$, indirect constraints, etc.
- Improvements in advanced analysis techniques (e.g., Deep Learning) and instrumentation (e.g., tracker)
- Reduction of systematic uncertainties: c -tagging, event modeling, theoretical uncertainties, ...

A charming journey ahead!

Backups

H \rightarrow cc searches at the LHC

□ ATLAS:

- [[Phys. Rev. Lett. 120 \(2018\) 211802](#)] (36 fb⁻¹)
- [[arXiv:2201.11428](#)] (139 fb⁻¹)
- [[ATL-PHYS-PUB-2021-039](#)] (HL-LHC projection, 3000 fb⁻¹)

□ CMS:

- [[JHEP 03 \(2020\) 131](#)] (36 fb⁻¹)
- [[CMS-PAS-HIG-21-008](#)] (138 fb⁻¹; HL-LHC projection, 3000 fb⁻¹)

□ LHCb:

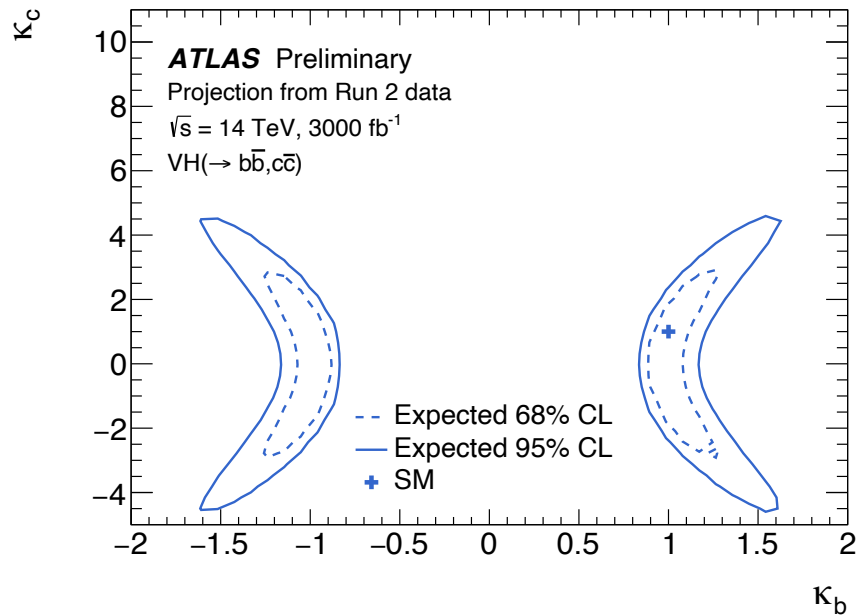
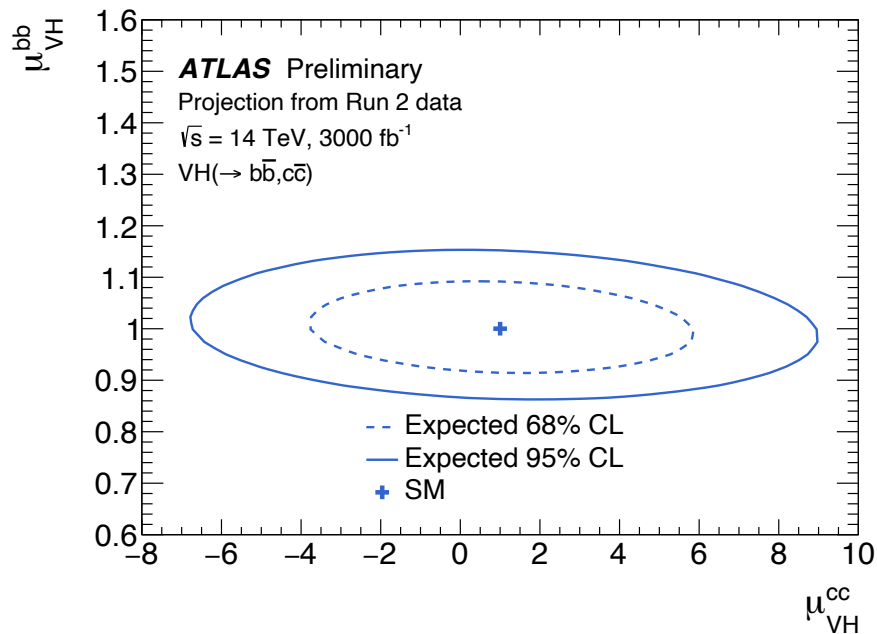
- [[LHCb-CONF-2016-006](#)] (1.98 fb⁻¹)
- [[LHCb-PUB-2018-009](#)] (HL-LHC projection, 300 fb⁻¹)

Higgs-charm coupling at HL-LHC

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

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

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



Baseline event selections

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

□ Breakdown of the uncertainties in each topology

Merged-jet topology

Table 3: The relative contributions to the total uncertainty on $\mu_{\text{VH}(H \rightarrow c\bar{c})}$ in the merged-jet analysis, with a best fit value $\mu_{\text{VH}(H \rightarrow c\bar{c})} = 8.7^{+4.6}_{-4.0}$.

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

Table 4: The relative contributions to the total uncertainty on $\mu_{\text{VH}(H \rightarrow c\bar{c})}$ in the resolved-jet analysis, with a best fit value $\mu_{\text{VH}(H \rightarrow c\bar{c})} = -9.5 \pm 9.6$.

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%

BDT input variables: Merged-jet topology

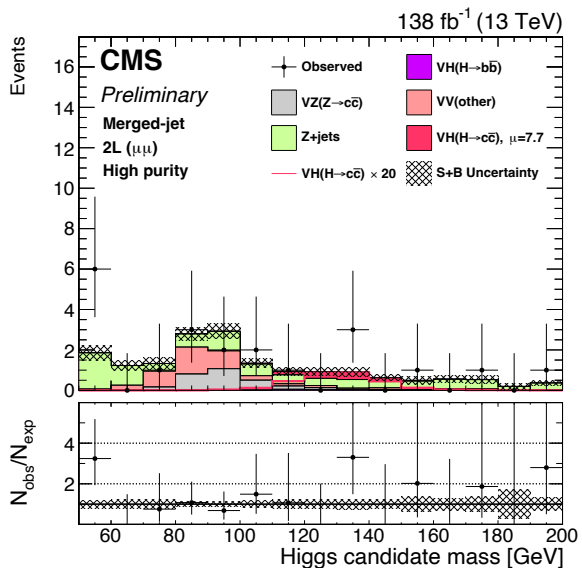
Variable	Description	0-lepton	1-lepton	2-lepton
$p_T(V)$	vector boson transverse momentum	✓	✓	✓
$\Delta R(\ell, \ell)$	angular separation between the two leptons	—	—	✓
$p_T(H_{\text{cand}})$	H_{cand} transverse momentum	✓	✓	✓
$ \eta(H_{\text{cand}}) $	absolute value of the H_{cand} pseudorapidity	✓	—	—
$\Delta\phi(V, H_{\text{cand}})$	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 $\vec{p}_T + \vec{p}_T^{\text{miss}}$	—	✓	—
$N_{\text{small-}R}^{\text{aj}}$	number of additional small- R jets	✓	✓	✓

BDT input variables: Resolved-jet topology

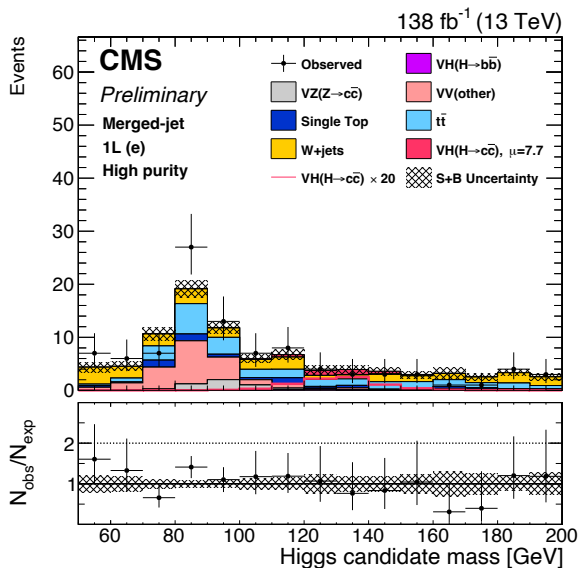
Variable	Description	0-lepton	1-lepton	2-lepton
$m(H_{\text{cand}})$	H_{cand} mass	✓	✓	✓
$p_T(H_{\text{cand}})$	H_{cand} transverse momentum	—	✓	✓
$p_T(V)$	vector boson transverse momentum	—	✓	✓
$m_T(V)$	vector boson transverse mass	—	✓	—
p_T^{miss}	missing transverse momentum	✓	✓	—
$p_T(V)/p_T(H_{\text{cand}})$	ratio between vector boson and H transverse momenta	✓	✓	✓
$CvsL_{\text{max}}$	$CvsL$ value of the leading $CvsL$ jet	✓	✓	✓
$CvsB_{\text{max}}$	$CvsB$ value of the leading $CvsL$ jet	✓	✓	✓
$CvsL_{\text{min}}$	$CvsL$ value of the subleading $CvsL$ jet	✓	✓	✓
$CvsB_{\text{min}}$	$CvsB$ value of the subleading $CvsL$ jet	✓	✓	✓
$p_{T\text{max}}$	p_T of the leading $CvsL$ jet	✓	✓	✓
$p_{T\text{min}}$	p_T of the subleading $CvsL$ jet	✓	✓	✓
$\Delta\phi(V, H_{\text{cand}})$	azimuthal angle between vector boson and H	✓	✓	✓
$\Delta R(j_1, j_2)$	ΔR between leading and subleading $CvsL$ jets	—	✓	✓
$\Delta\phi(j_1, j_2)$	azimuthal angle between leading and subleading $CvsL$ jets	✓	✓	—
$\Delta\eta(j_1, j_2)$	difference in pseudorapidity between leading and subleading $CvsL$ jets	✓	✓	—
$\Delta\phi(\ell_1, \ell_2)$	azimuthal angle between leading and subleading p_T leptons	—	—	✓
$\Delta\eta(\ell_1, \ell_2)$	difference in pseudorapidity between leading and subleading p_T leptons	—	—	✓
$\Delta\phi(\ell_1, j_1)$	azimuthal angle between leading p_T lepton and leading $CvsL$ jet	—	✓	—
$\Delta\phi(\ell_2, j_1)$	azimuthal angle between subleading p_T lepton and leading $CvsL$ jet	—	—	✓
$\Delta\phi(\ell_2, j_2)$	azimuthal angle between subleading p_T lepton and subleading $CvsL$ jet	—	—	✓
$\Delta\phi(\ell_1, p_T^{\text{miss}})$	azimuthal angle between leading p_T lepton and missing transverse momentum	—	✓	—
$\Delta\eta(\ell_1, b)$	difference in pseudorapidity between leading p_T lepton and b-tagged jet from top quark decay	—	✓	—
$\Delta\phi(\ell_1, b)$	azimuthal angle between leading p_T lepton and b-tagged jet from top quark decay	—	✓	—
$\Delta R(\ell_1, b)$	ΔR between leading p_T lepton and b-tagged jet from top quark decay	—	✓	—
$CvsL_b$	$CvsL$ value of the b-tagged jet from top quark decay	—	✓	—
$CvsB_b$	$CvsB$ value of the b-tagged jet from top quark decay	—	✓	—
$P(b+bb)_b$	$DeepJet$ $prob(b+bb)$ value of the b-tagged jet from top quark decay	—	✓	—
$m(\bar{t})$	Reconstructed top quark mass	—	✓	—
$N_{\text{small-R}}^{\text{add}}$	Number of small-R additional jets after the FSR subtraction	—	✓	—
$\sigma_{c,Reg}(j_1)$	leading p_T jet resolution from c-jet energy regression	✓	✓	✓
$\sigma_{c,Reg}(j_2)$	subleading p_T jet resolution from c-jet energy regression	✓	✓	✓
$\Delta\eta(V, H_{\text{cand}}) \parallel_{\text{kinfit}}$	difference in pseudorapidity between vector boson and H_{cand} , after kinematic-fit	—	—	✓
$\Delta\phi(V, H_{\text{cand}}) \parallel_{\text{kinfit}}$	azimuthal angle between vector boson and H_{cand} , after kinematic-fit	—	—	✓
$m(H_{\text{cand}}) \parallel_{\text{kinfit}}$	H_{cand} mass after kinematic-fit	—	—	✓
$p_T(H_{\text{cand}}) \parallel_{\text{kinfit}}$	H_{cand} transverse momentum after kinematic-fit	—	—	✓
$p_{T\text{max}} \parallel_{\text{kinfit}}$	p_T of the leading $CvsL$ jet after kinematic-fit	—	—	✓
$p_{T\text{min}} \parallel_{\text{kinfit}}$	p_T of the subleading $CvsL$ jet after kinematic-fit	—	—	✓
$p_T(V)/p_T(H_{\text{cand}}) \parallel_{\text{kinfit}}$	ratio between vector boson and H_{cand} transverse momenta after kinematic-fit	—	—	✓
$\sigma(H_{\text{cand}}) \parallel_{\text{kinfit}}$	H_{cand} invariant mass resolution from kinematic fit	—	—	✓

Merged-jet topology: signal regions

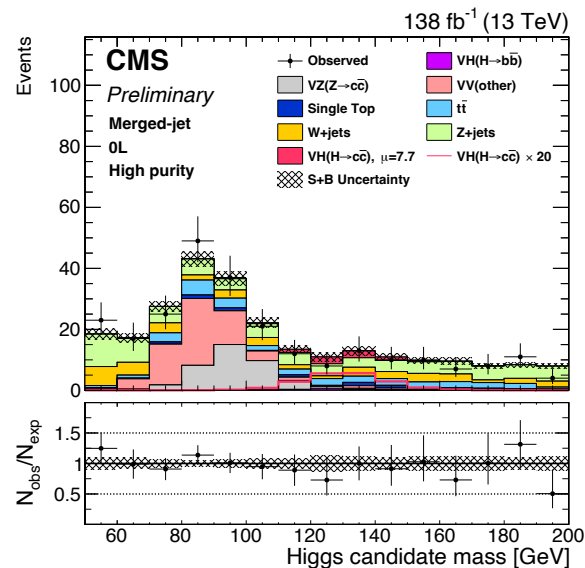
2L($\mu\mu$), high cc-purity



1L(e), high cc-purity

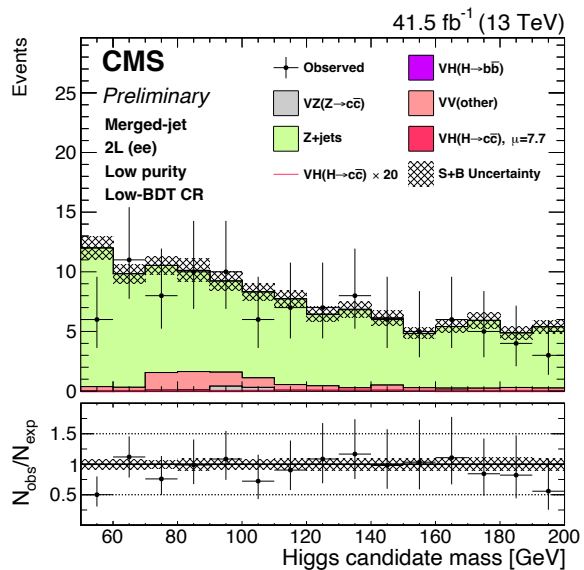


0L, high cc-purity

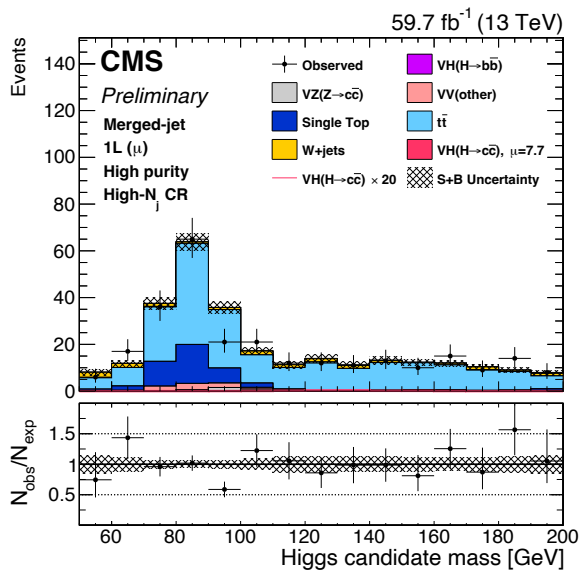


Merged-jet topology: control regions

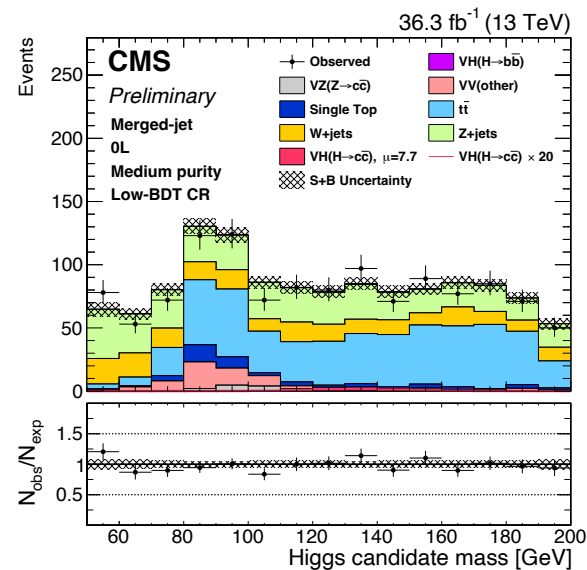
2L(ee), V+jets CR, low cc-purity



1L(μ), tt CR, high cc-purity

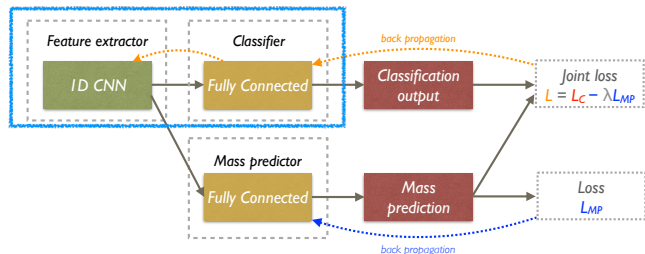


0L, V+jets CR, medium cc-purity

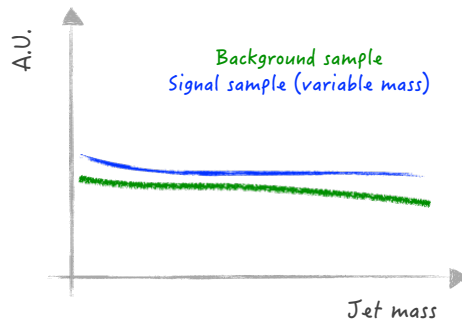


Comparison of mass decorrelation methods

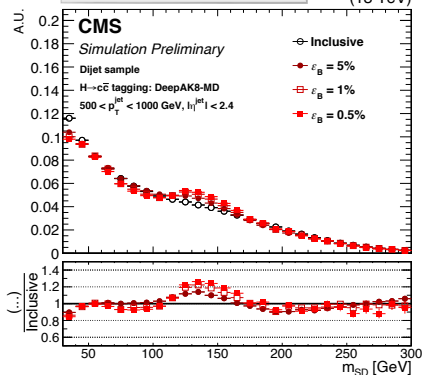
DeepAK8-MD



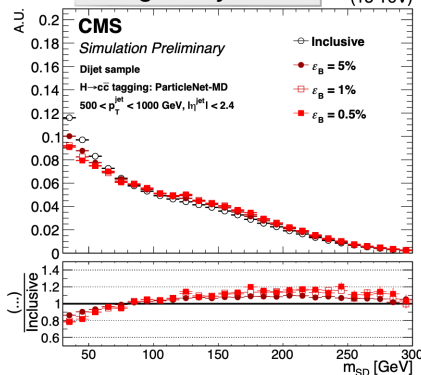
ParticleNet-MD



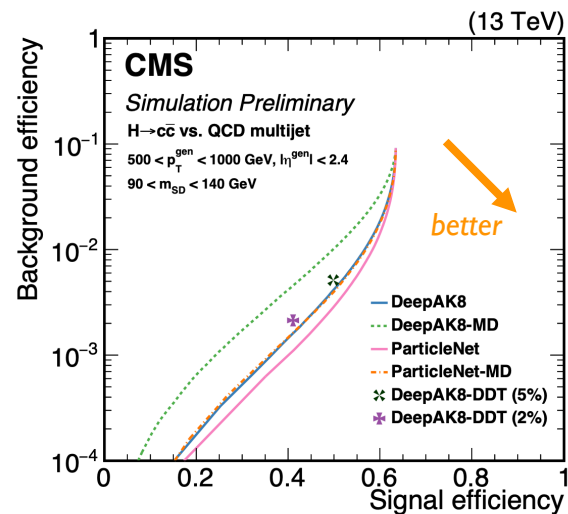
Background jet mass



Background jet mass



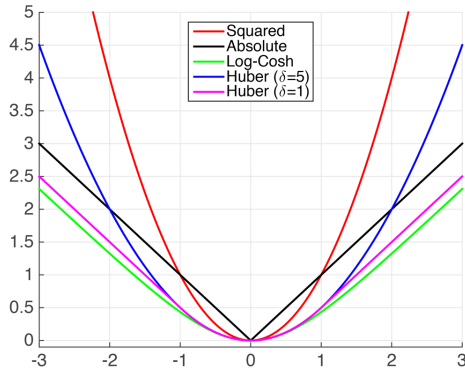
H -> cc tagging performance



Large-R jet mass regression

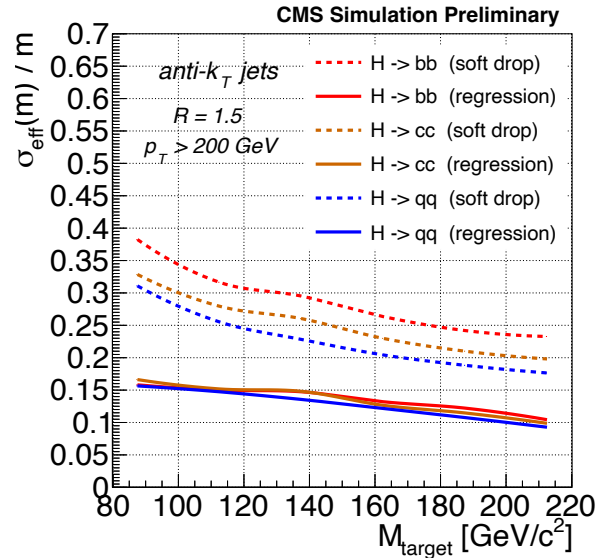
Loss function: LogCosh

$$L(y, y^p) = \sum_{i=1}^n \log(\cosh(y_i^p - y_i))$$

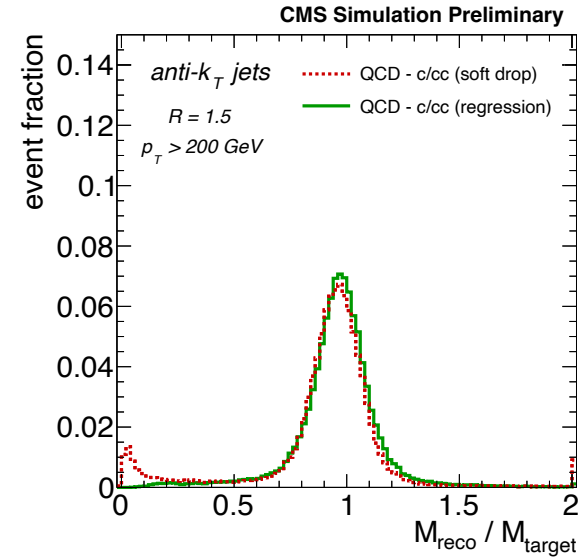


<https://www.cs.cornell.edu/courses/cs4780/2015fa/web/lecturenotes/lecturenote10.html>

Signal jet mass resolution

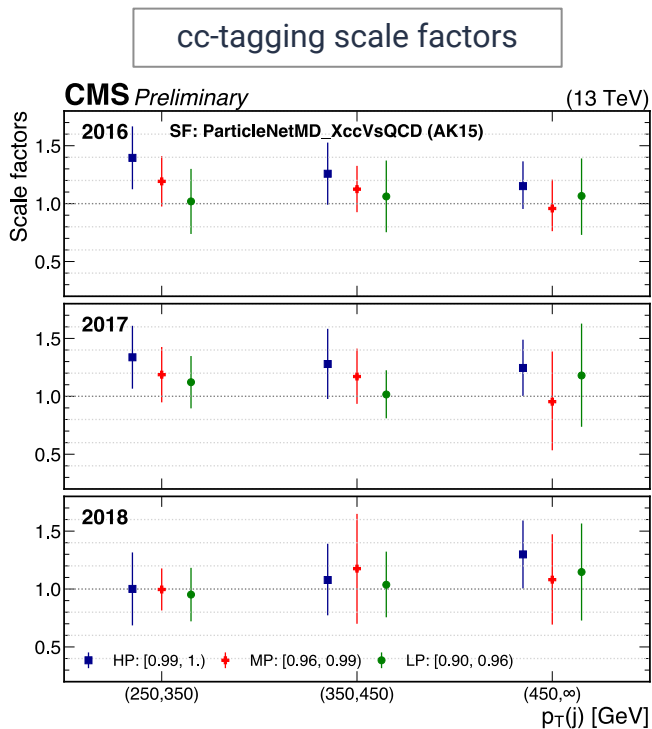
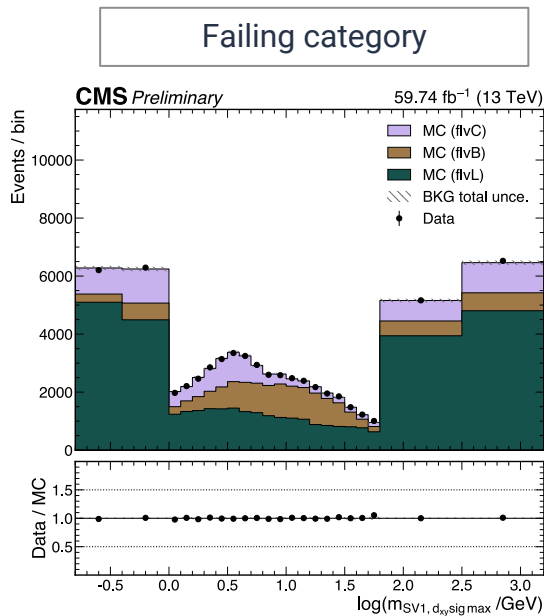
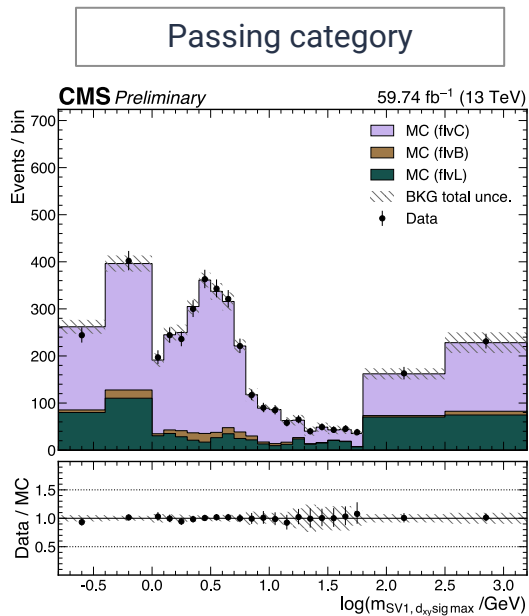


Background jet mass response



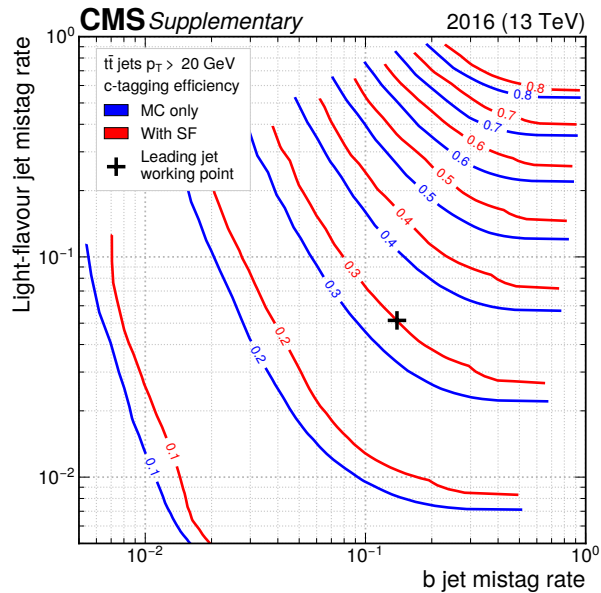
Calibration of the cc-tagger

CMS-DP-2022-005

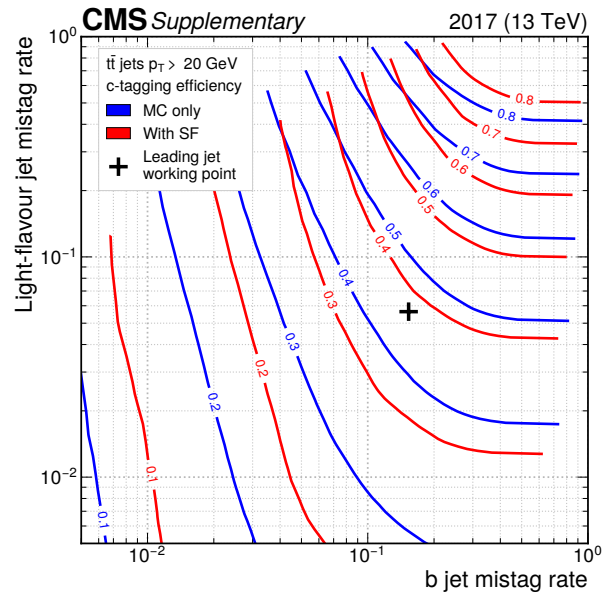


C-tagger ROC curves

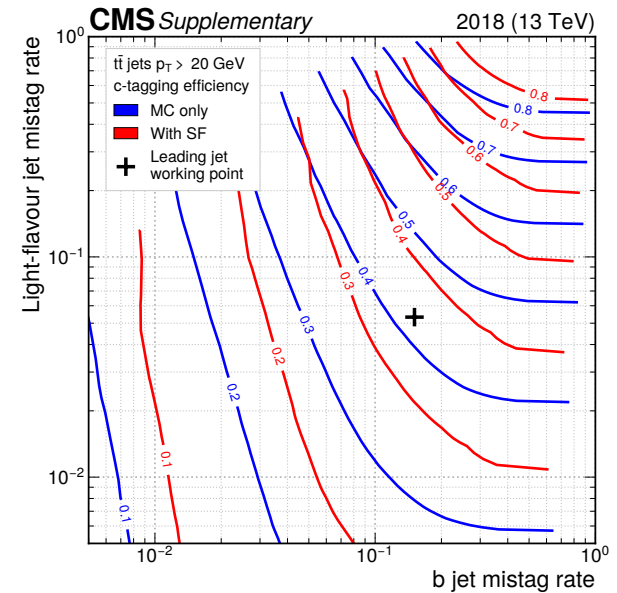
2016



2017



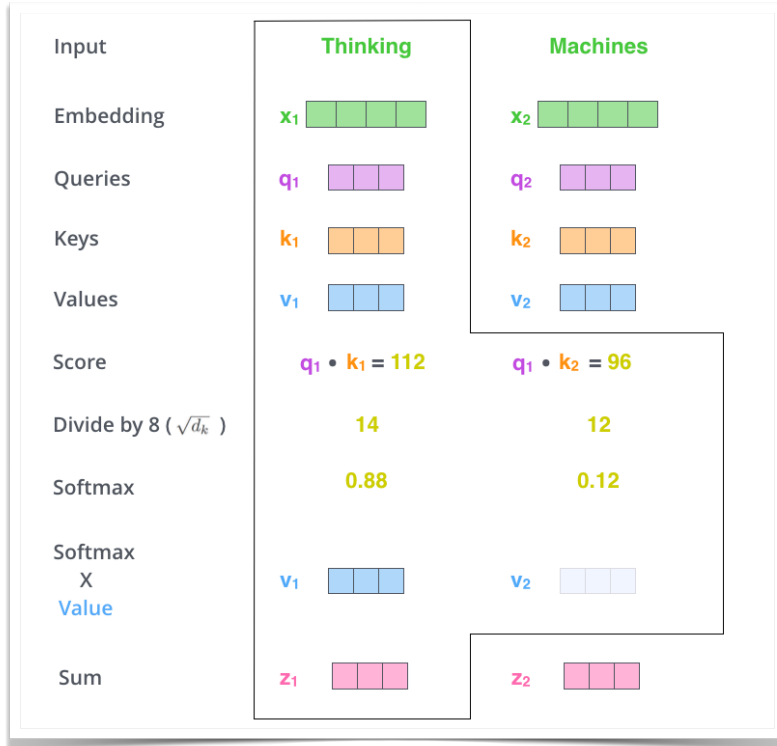
2018



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

Transformer 101

Vaswani, Shazeer, Parmar, Uszkoreit, Jones, Gomez, Kaiser, Polosukhin, arXiv:1706.03762



$$X \times W^Q = Q$$

$$X \times W^K = K$$

$$X \times W^V = V$$

$$\text{softmax} \left(\frac{Q \times K^T}{\sqrt{d_k}} \right) \times V = Z$$

<https://jalamar.github.io/illustrated-transformer/> ×