

# Some applications of machine learning and quantum computing in jet reconstruction

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# 1. A novel quantum realization of jet clustering in high-energy physics experiments

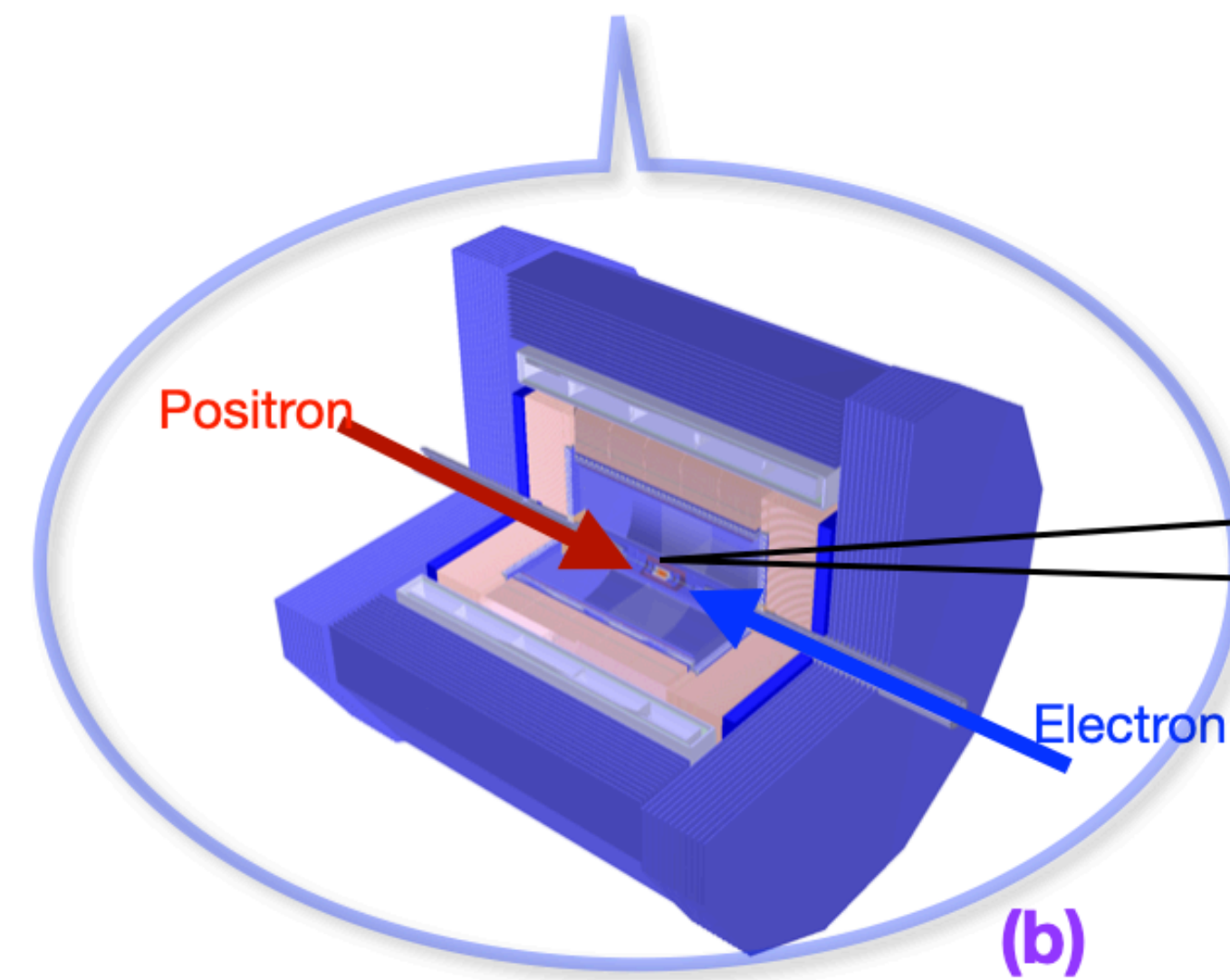
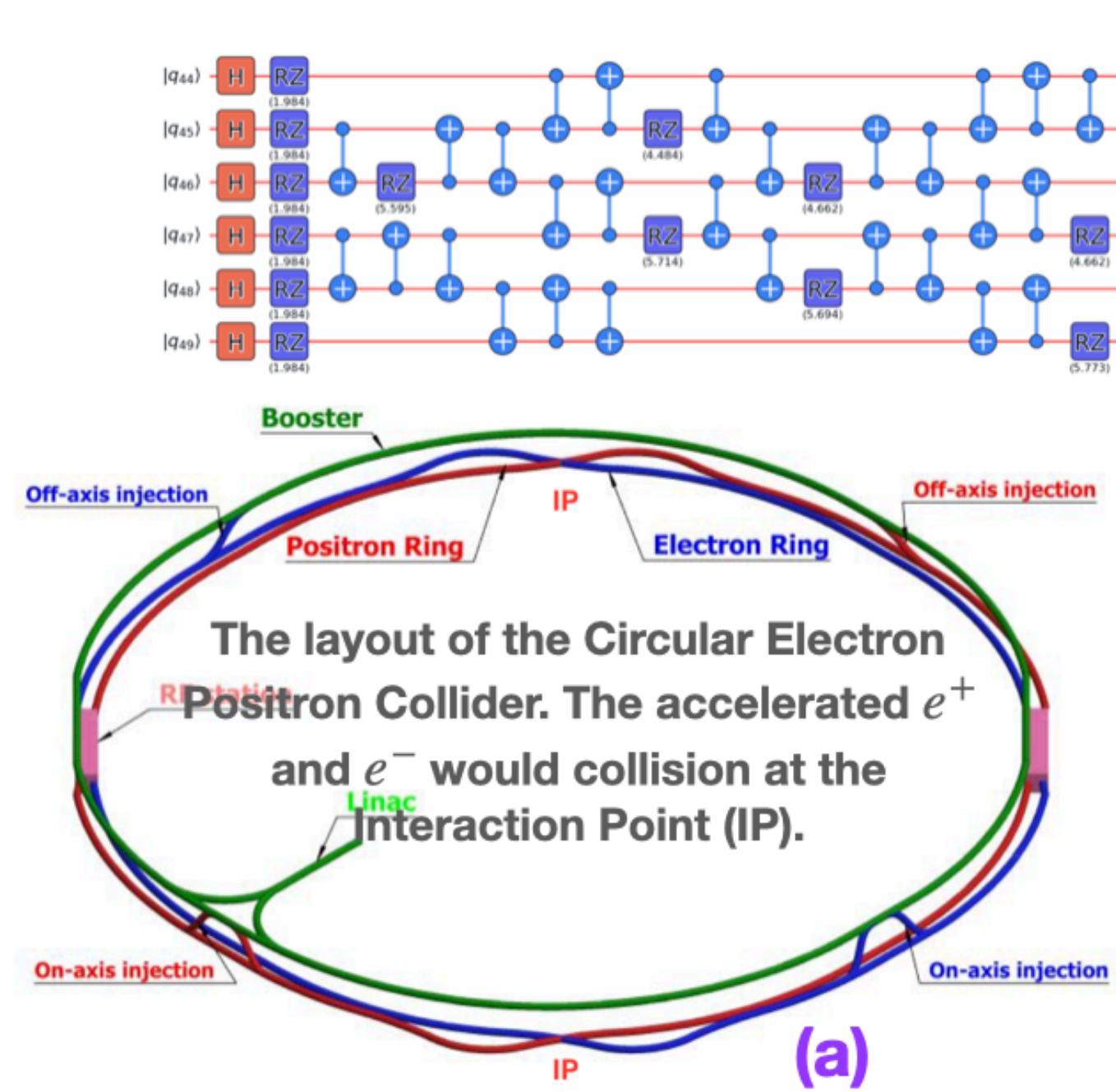
[Science Bulletin 70 \(2025\) 460](#)

In collaboration with Manqi Ruan (IHEP), Dong Liu (Tsinghua/BAQIS), and others

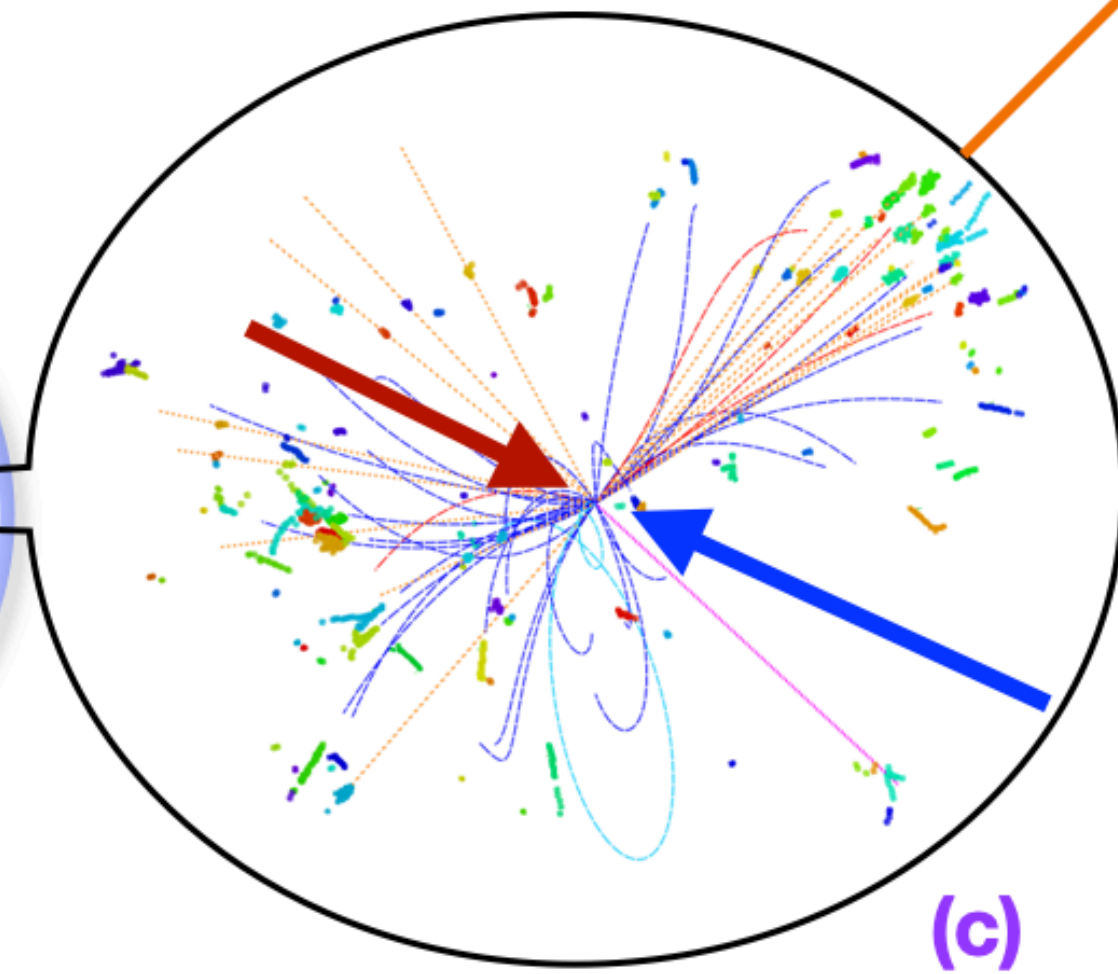
# A novel quantum realization of jet clustering

- Exploring the application of quantum technologies to fundamental sciences holds the key to fostering innovation for both sides
- Accurate jet clustering is crucial as it retains the information of the originating quark or gluon and forms the basis for many physics studies
- Quantum Approximate Optimization Algorithm (QAOA) is a hybrid quantum-classical algorithm for addressing classical combinatorial optimization problems with available quantum resources
- For the first time, by mapping collision events into graphs—with particles as nodes and their angular separations as edges—we realize jet clustering using QAOA

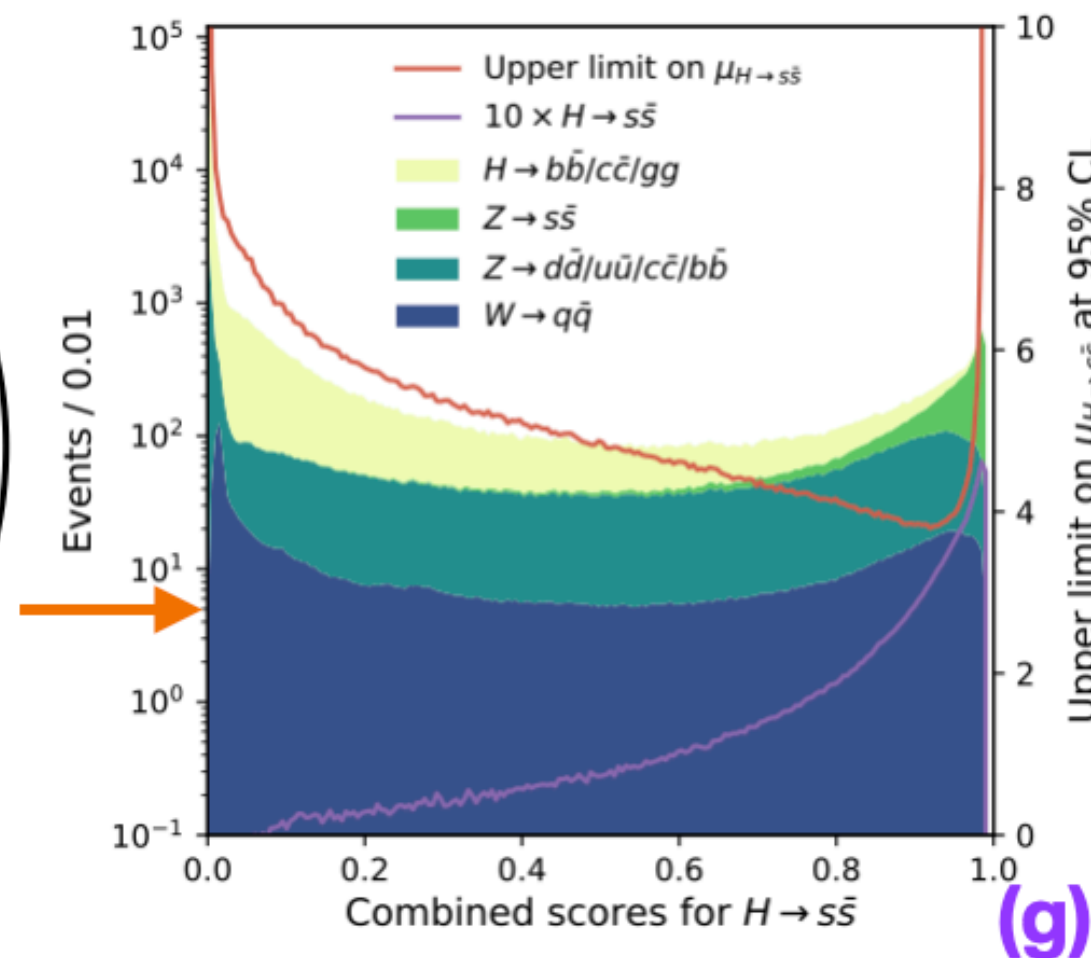




The collision of  $e^+$  and  $e^-$  can generate quarks, gluons, and leptons.

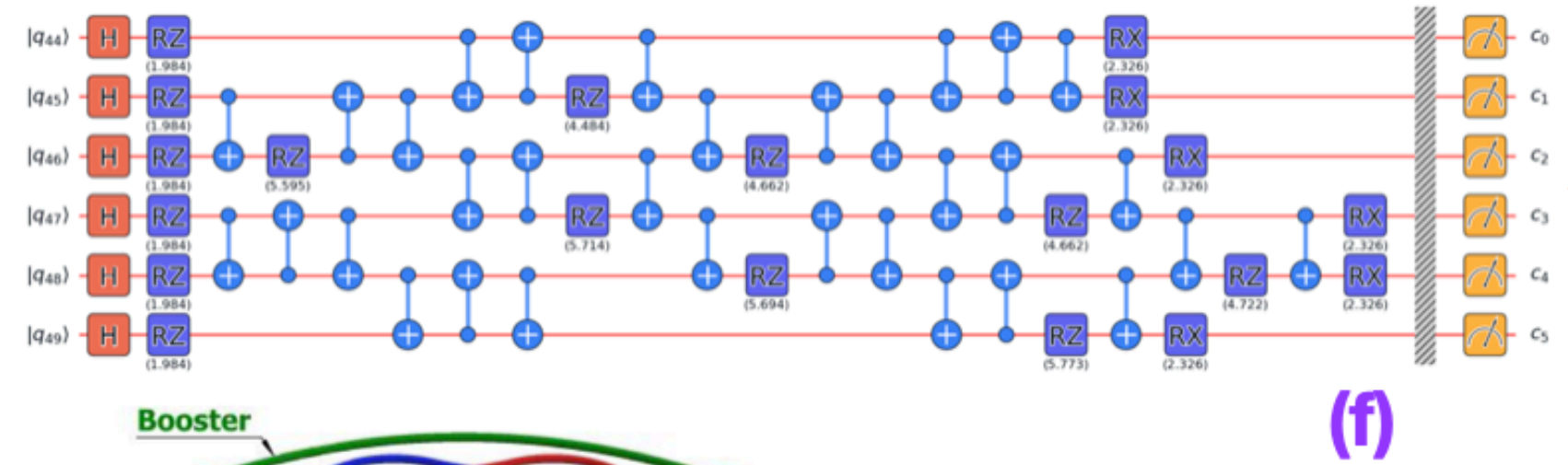
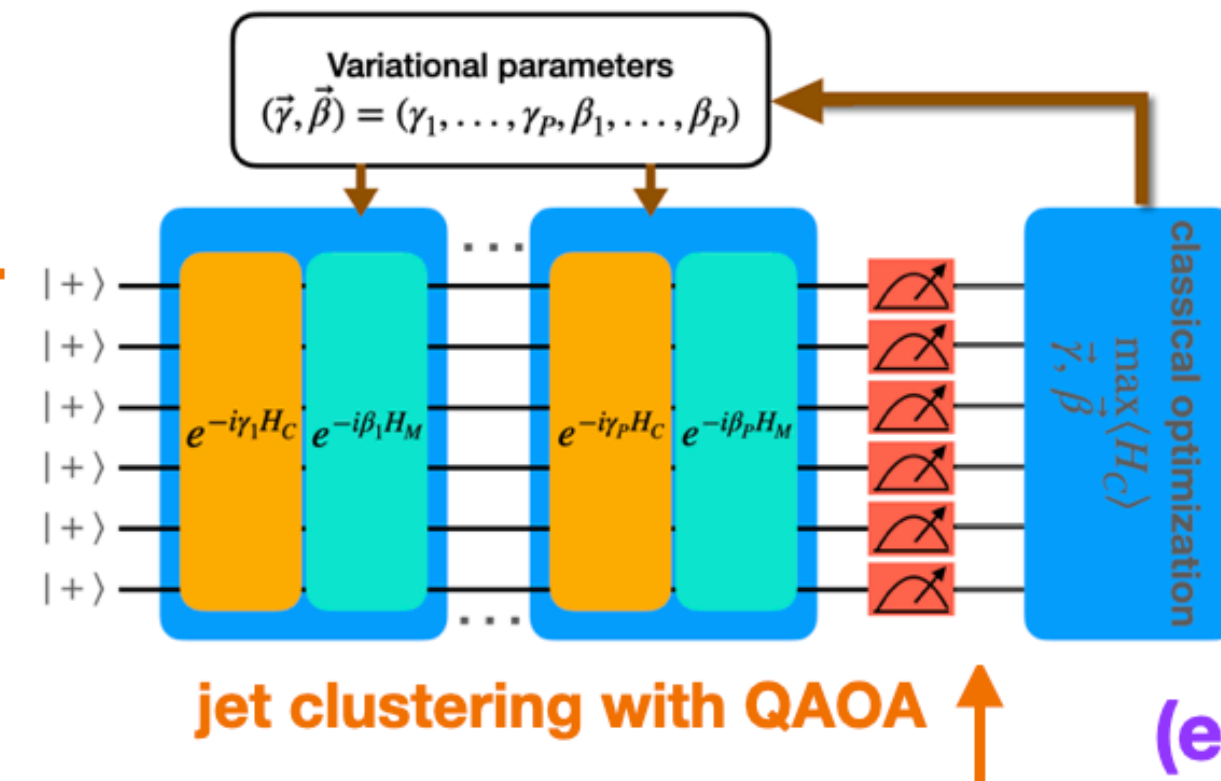
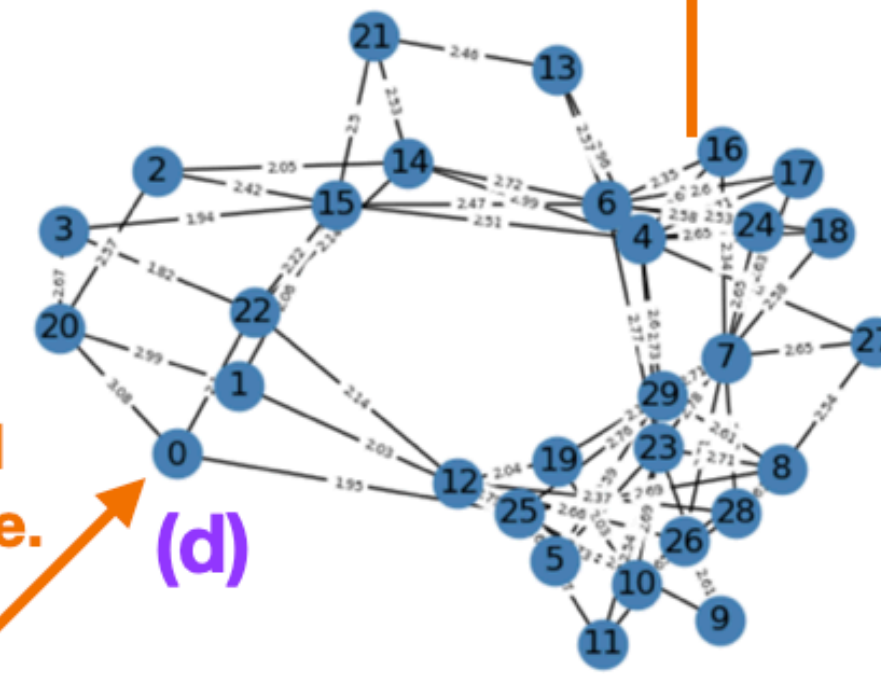


The quarks and gluons would immediately transform into collimated particle sprays known as jets.



With jet clustering and other techniques, the related physics analyses can be performed.

See the event as a graph, where particles as nodes and angle of two particles as edge.

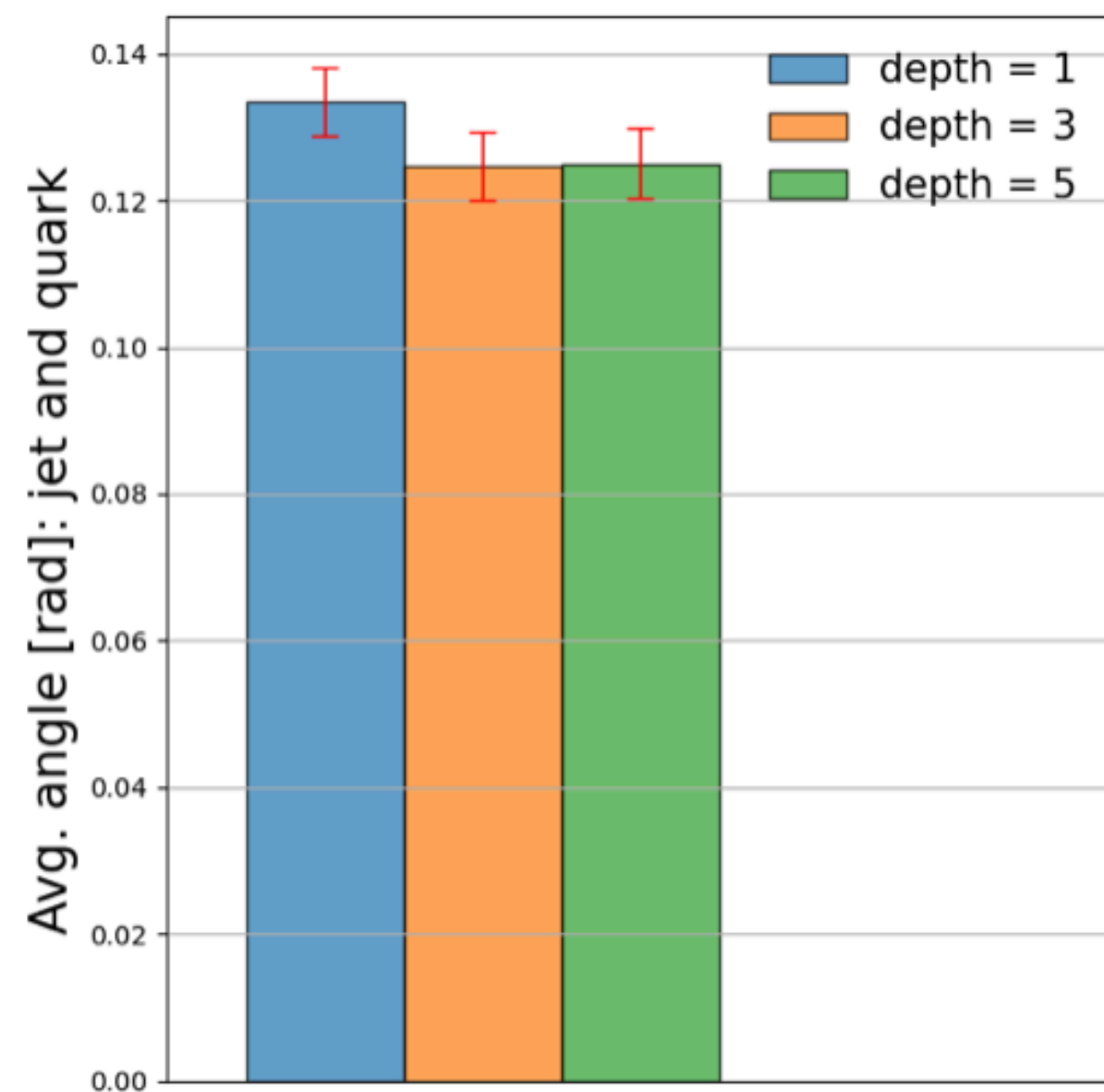
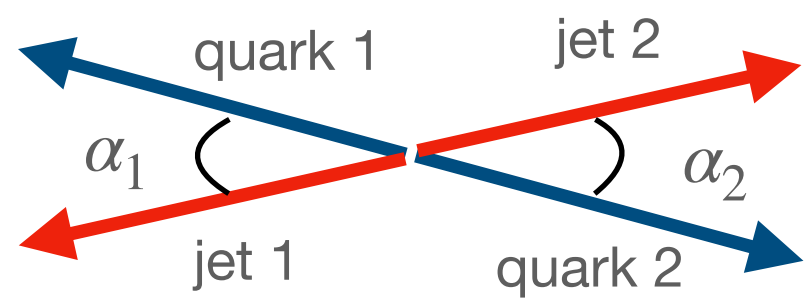


$$\hat{H}_C = \frac{1}{2} \sum_{(i,j) \in E} W_{ij} (I - \sigma_i^z \sigma_j^z)$$

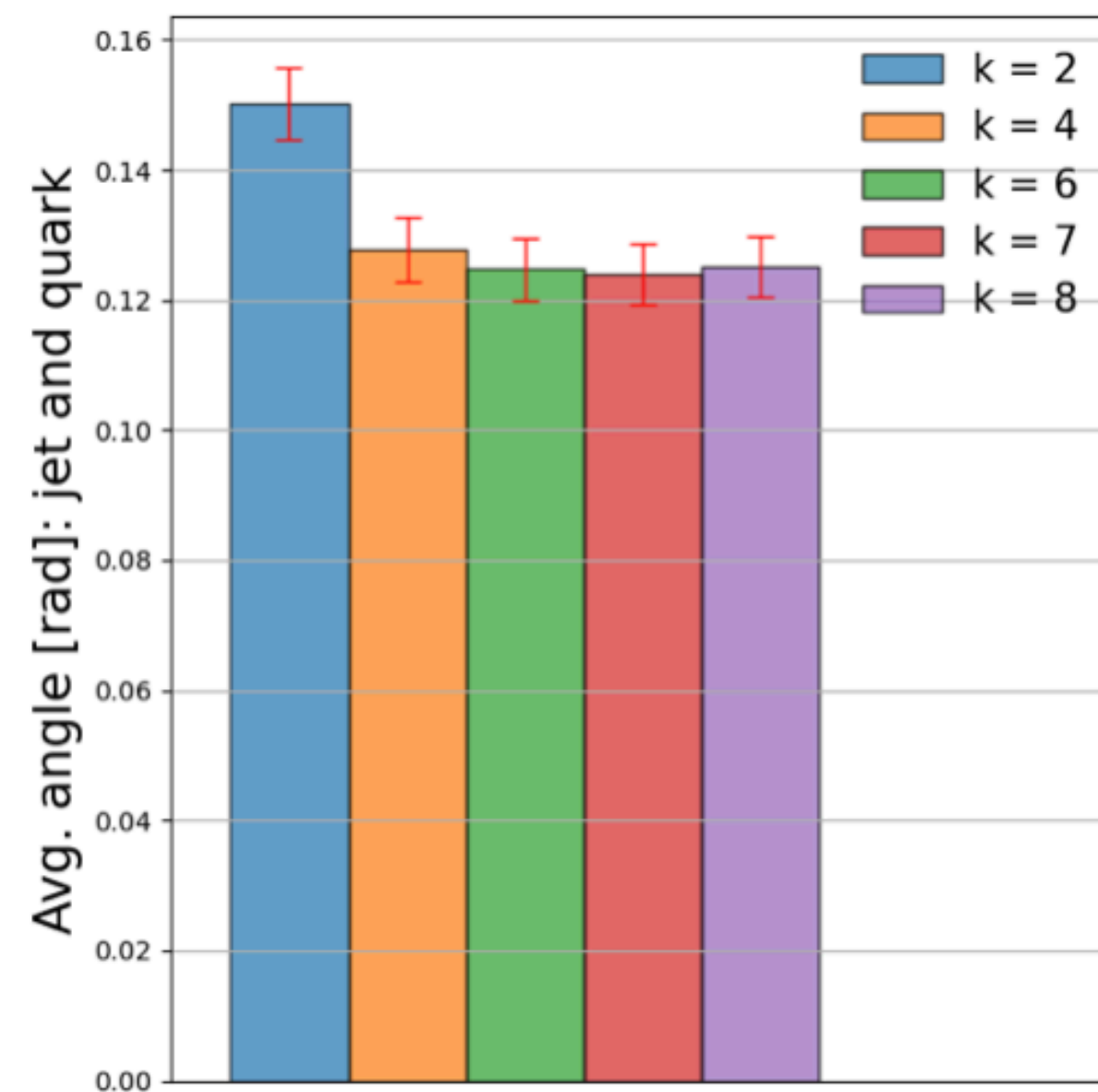
$$\hat{H}_M = \sum_{j=1}^n \sigma_j^x$$

$$C(\mathbf{x}) = \sum_{i,j=1}^{|V|} w_{ij} x_i (1 - x_j)$$

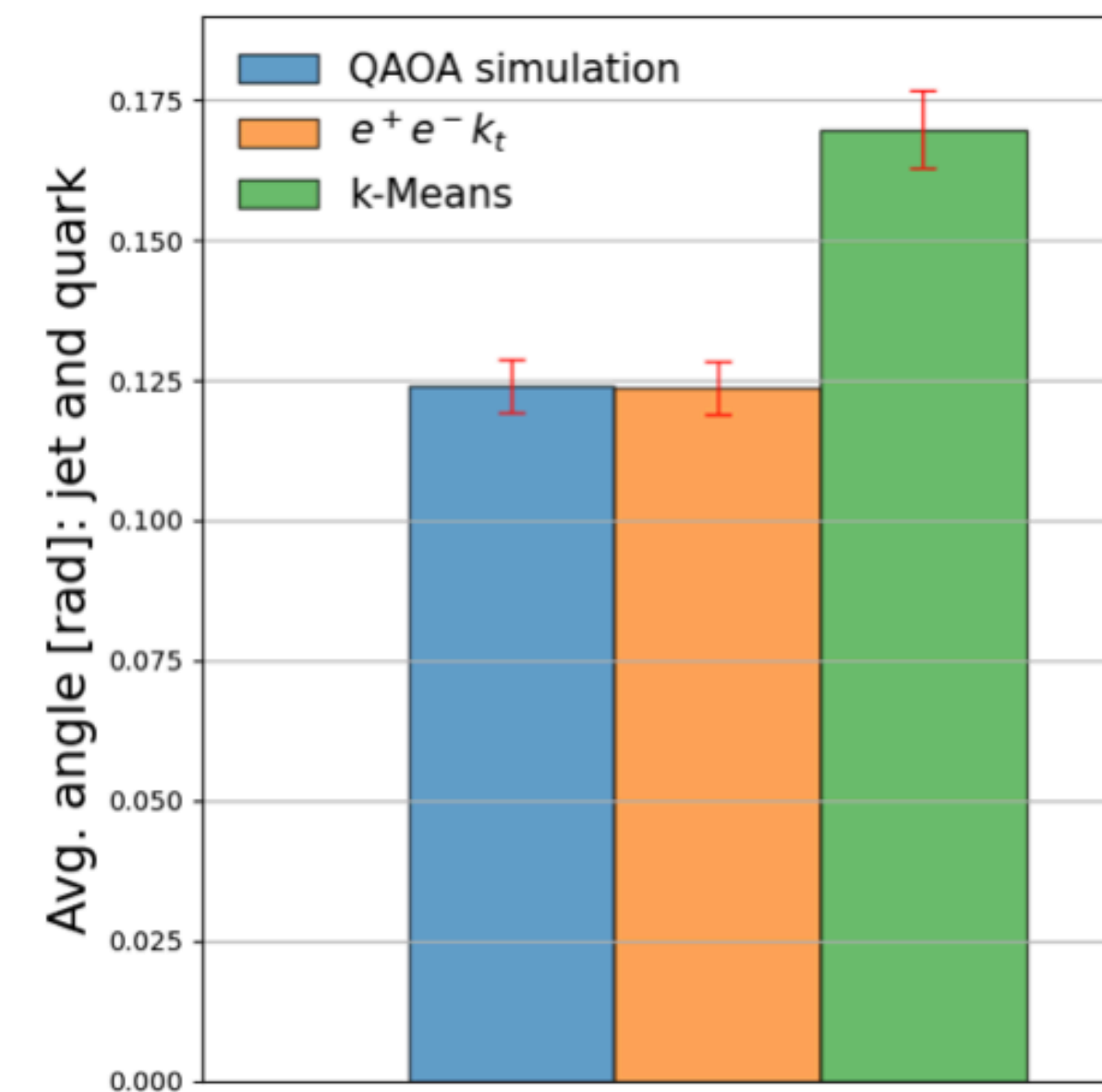




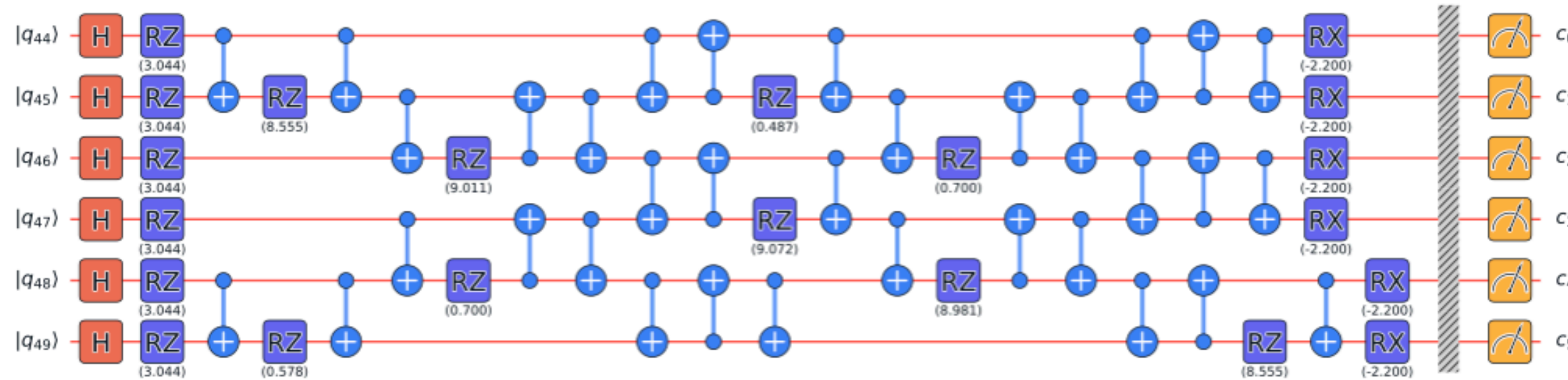
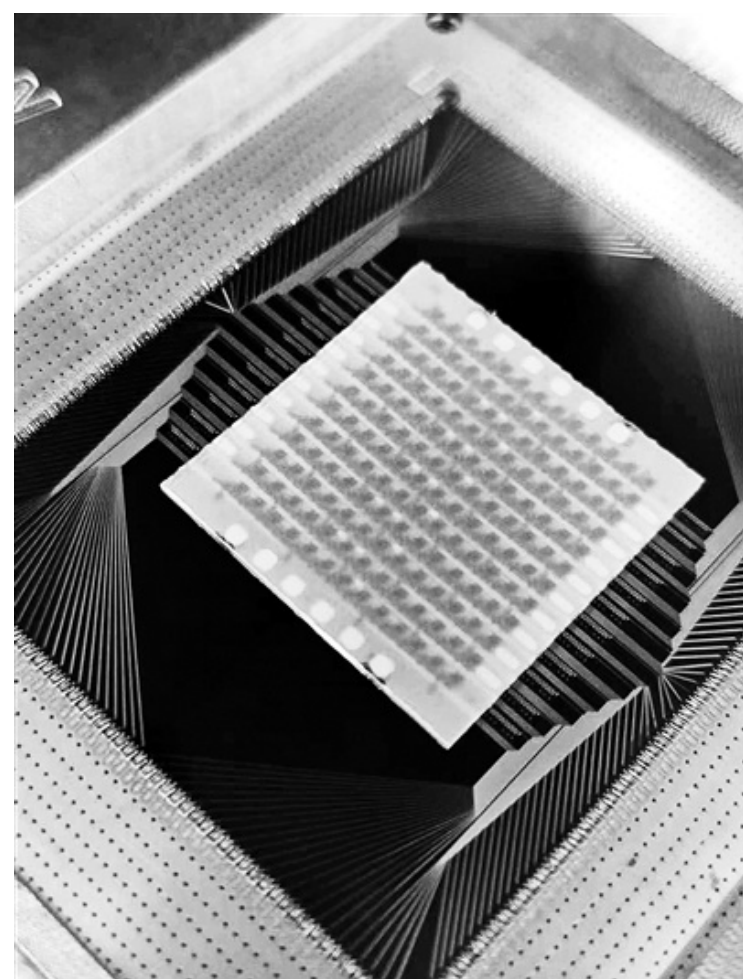
(a)



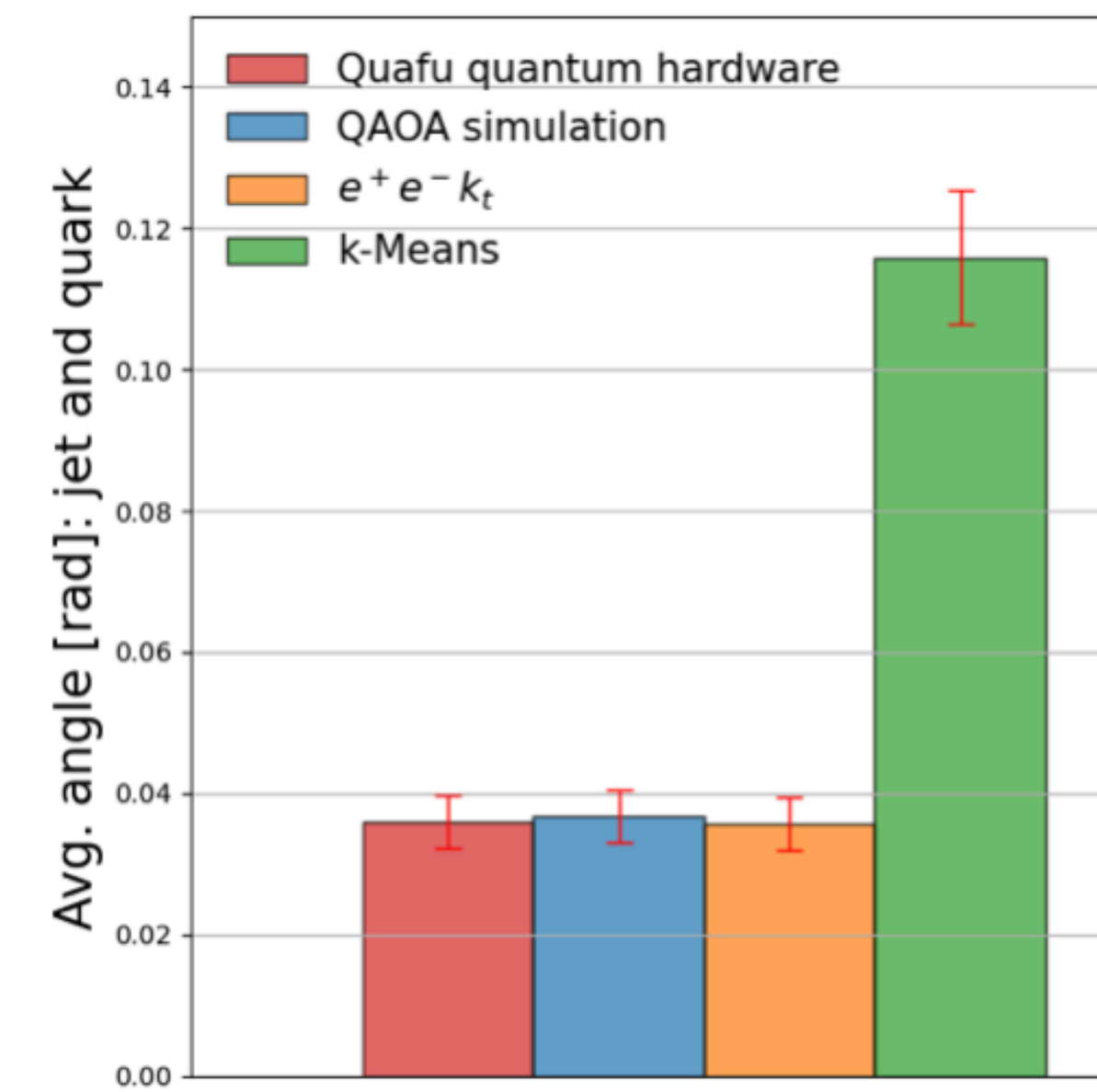
(b)



(c)



(d)



(e)

# A novel quantum realization of jet clustering

- Our results, derived from 30 qubits on quantum computer simulator and 6 qubits on quantum computer hardware, demonstrate that jet clustering performance with QAOA is comparable with classical algorithms for a small-sized problem
- This study highlights the feasibility of quantum computing to revolutionize jet clustering, bringing the practical application of quantum computing in high-energy physics experiments one step closer

## 2. Machine learning-based jet-origin identification and its application at an $e^-e^+$ Higgs factory

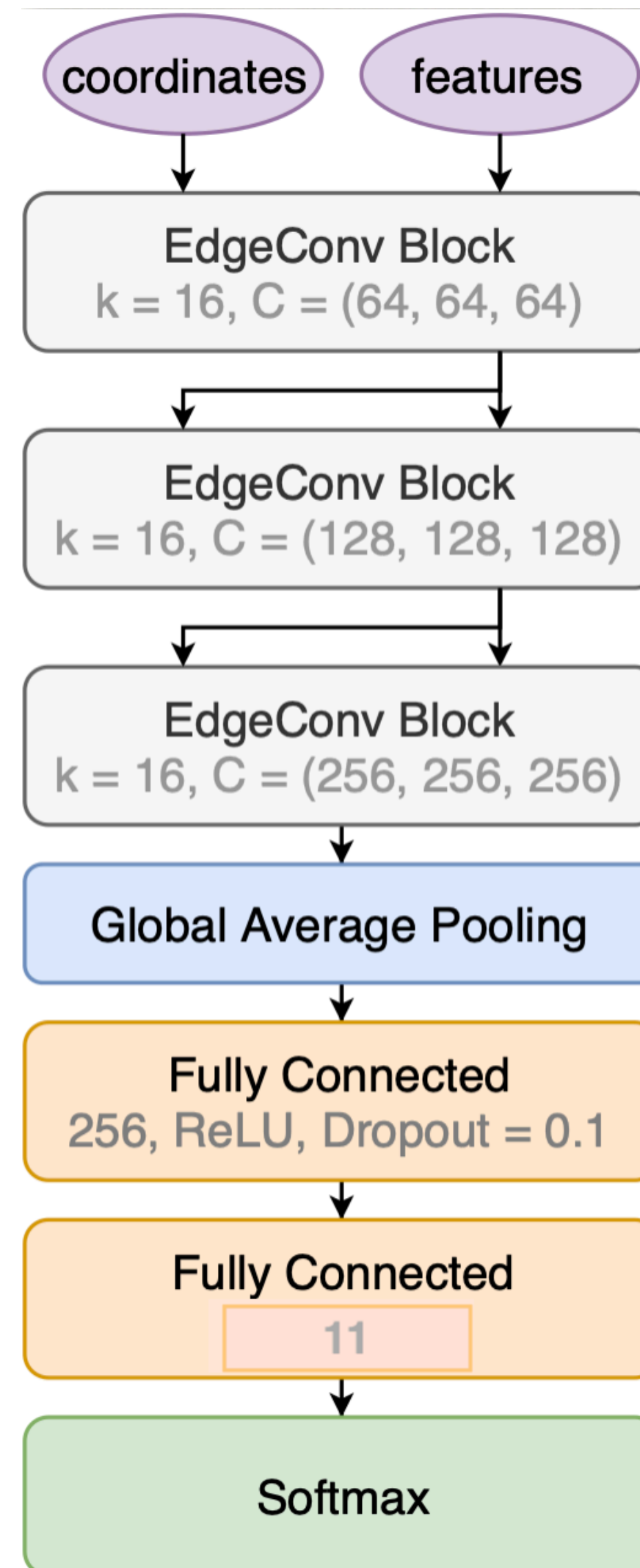
[Phys. Rev. Lett. 132 \(2024\) 221802](#)

In collaboration with Manqi Ruan (IHEP), Huilin Qu (CERN), and others



# Machine learning-based jet-origin identification

- We propose and realize the concept of jet-origin identification using state-of-the-art machine learning algorithms, such as ParticleNet

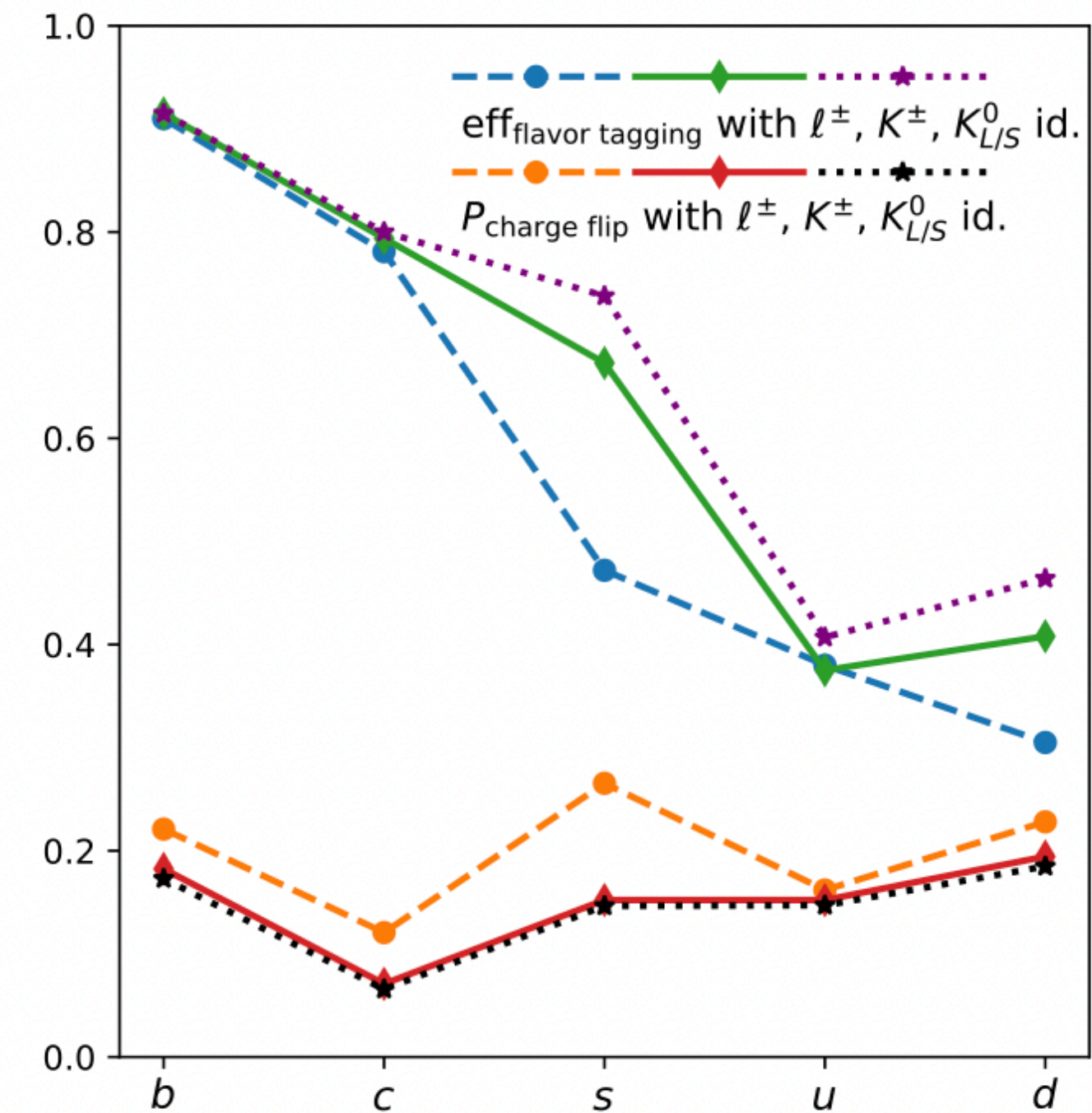
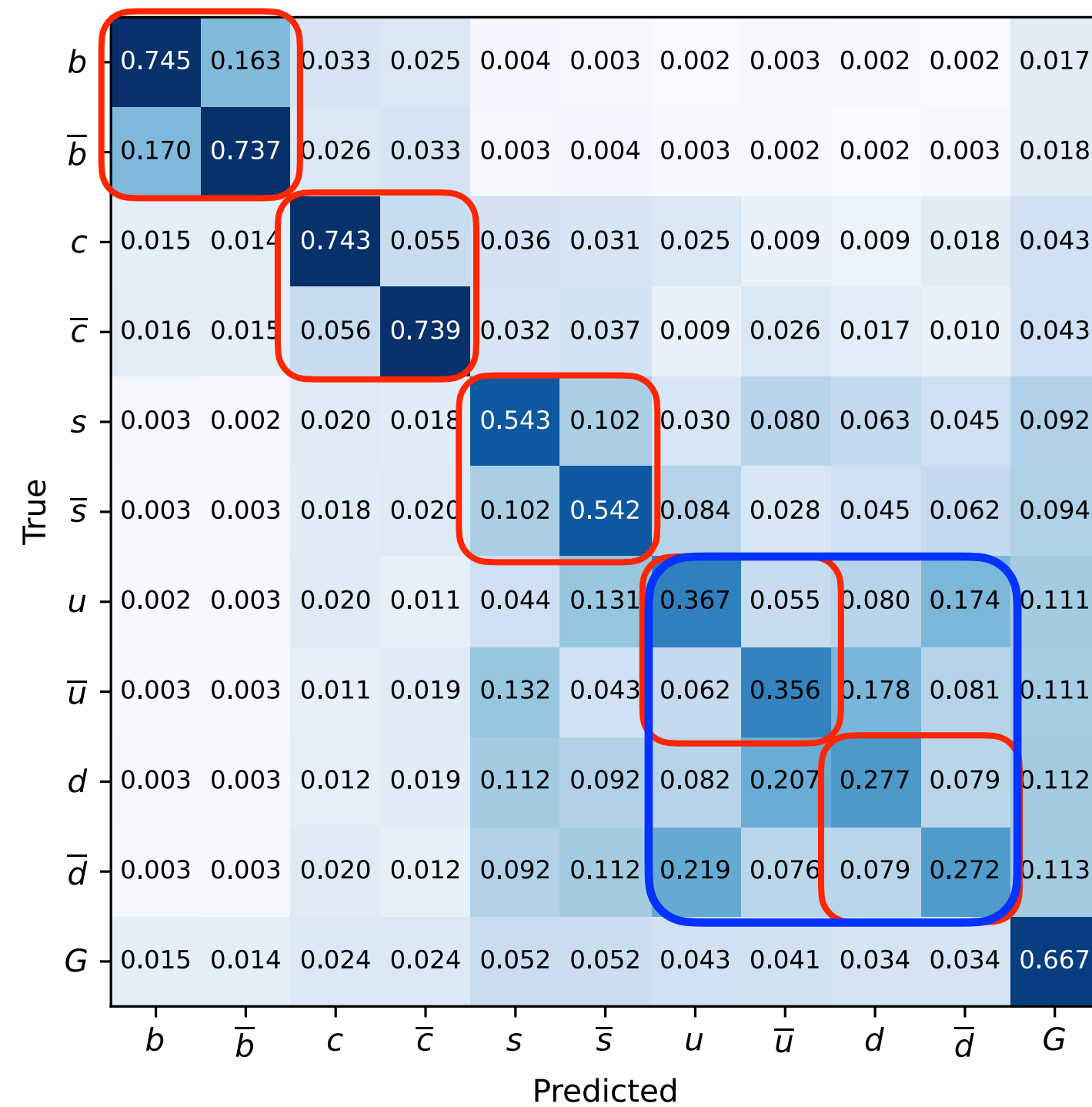


**Table 3** The input variables used in ParticleNet for jet flavor tagging at the CEPC

Variable	Definition
$\Delta \eta$	Difference in pseudorapidity between the particle and the jet axis
$\Delta \phi$	Difference in azimuthal angle between the particle and the jet axis
$\log P_t$	Logarithm of the particle's $P_t$
$\log E$	Logarithm of the particle's energy
$\log \frac{P_t}{P_t(jet)}$	Logarithm of the particle's $P_t$ relative to the jet $P_t$
$\log \frac{E}{E(jet)}$	Logarithm of the particle's energy relative to the jet energy
$\Delta R$	Angular separation between the particle and the jet axis
$d_0$	Transverse impact parameter of the track
$d_{0err}$	Uncertainty associated with the measurement of the $d_0$
$z_0$	Longitudinal impact parameter of the track
$z_{0err}$	Uncertainty associated with the measurement of the $z_0$
Charge	Electric charge of the particle
isElectron	Whether the particle is an electron
isMuon	Whether the particle is a muon
isChargedKaon	Whether the particle is a charged Kaon
isChargedPion	Whether the particle is a charged Pion
isProton	Whether the particle is a proton
isNeutralHadron	Whether the particle is a neutral hadron
isPhoton	Whether the particle is a photon



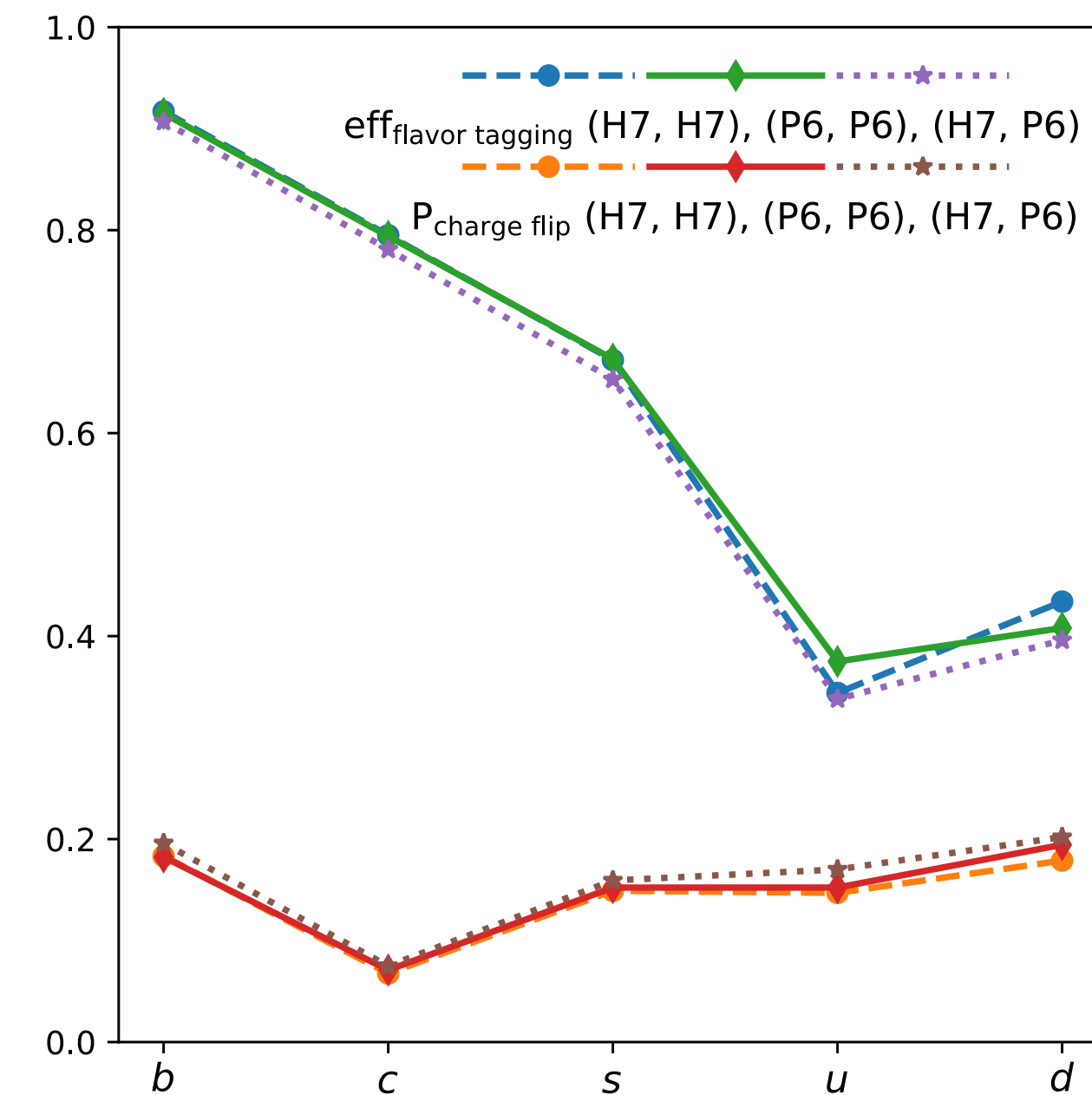
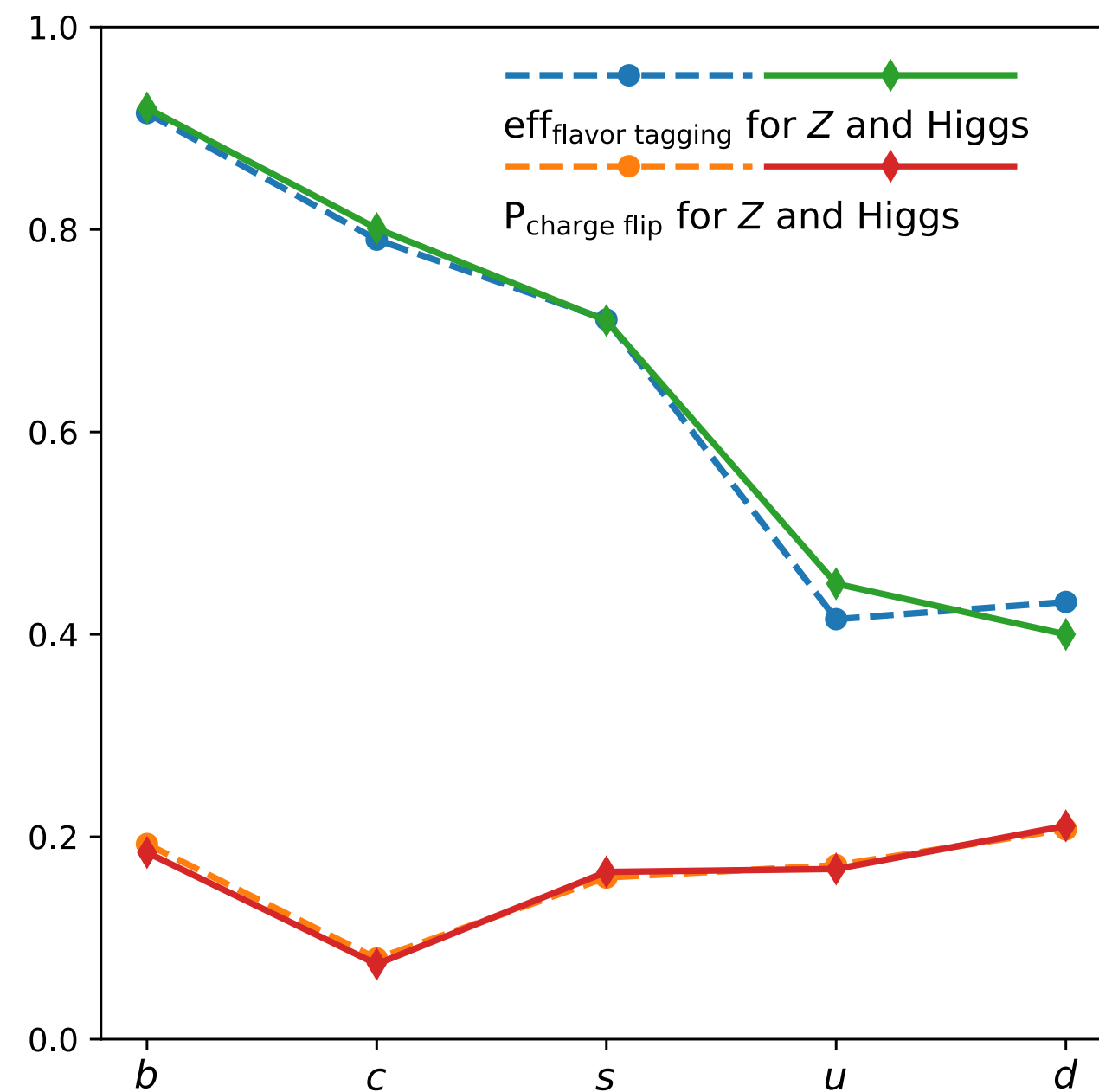
# Machine learning-based jet-origin identification



- Our jet-origin identification categorizes jets into five quark species (b, c, s, u, d), five antiquarks, and the gluon
- It reaches jet flavor tagging efficiencies ranging from 67% to 92% for b, c, and s quarks and jet charge flip rates of 7%–24% for all quark species

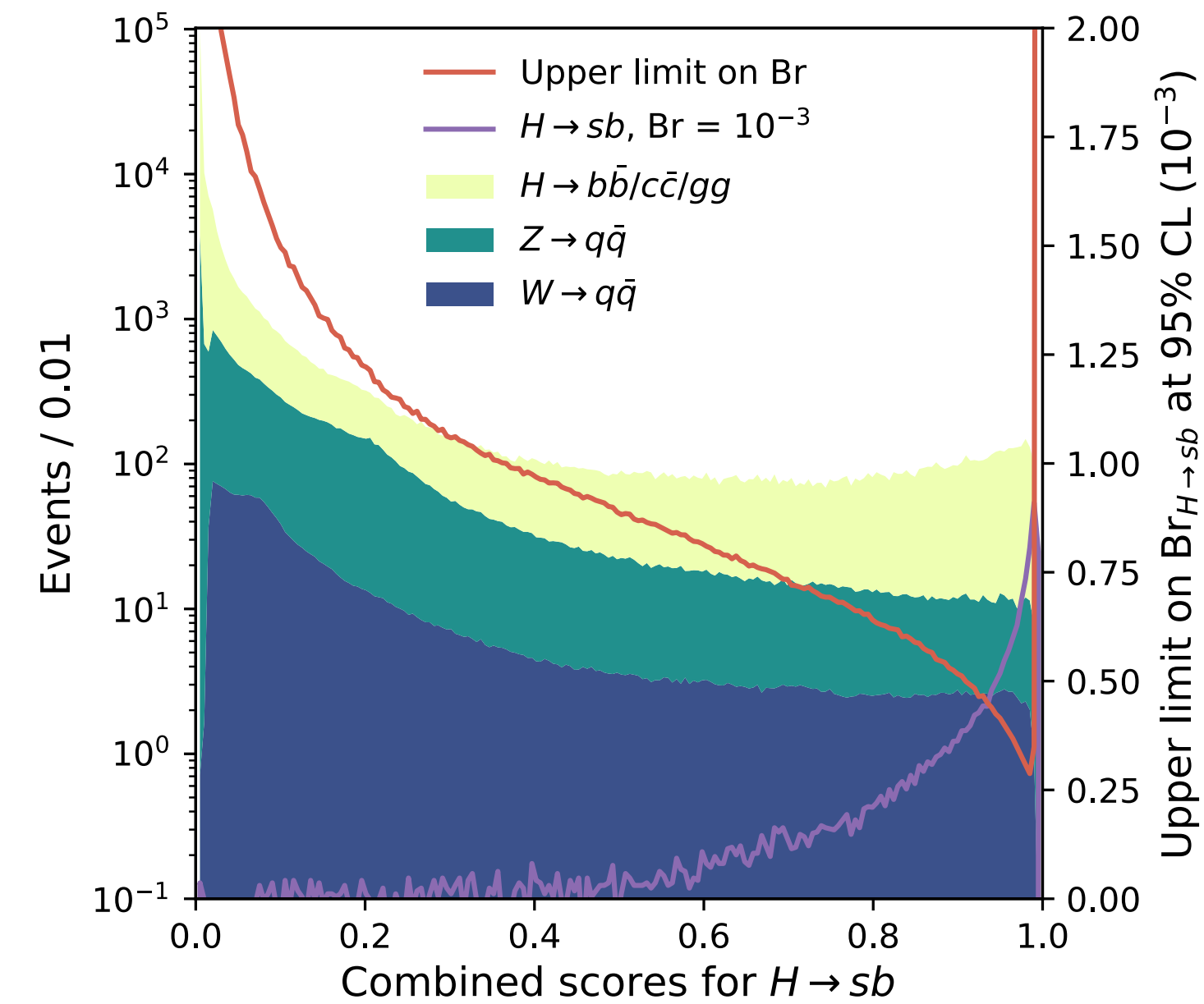
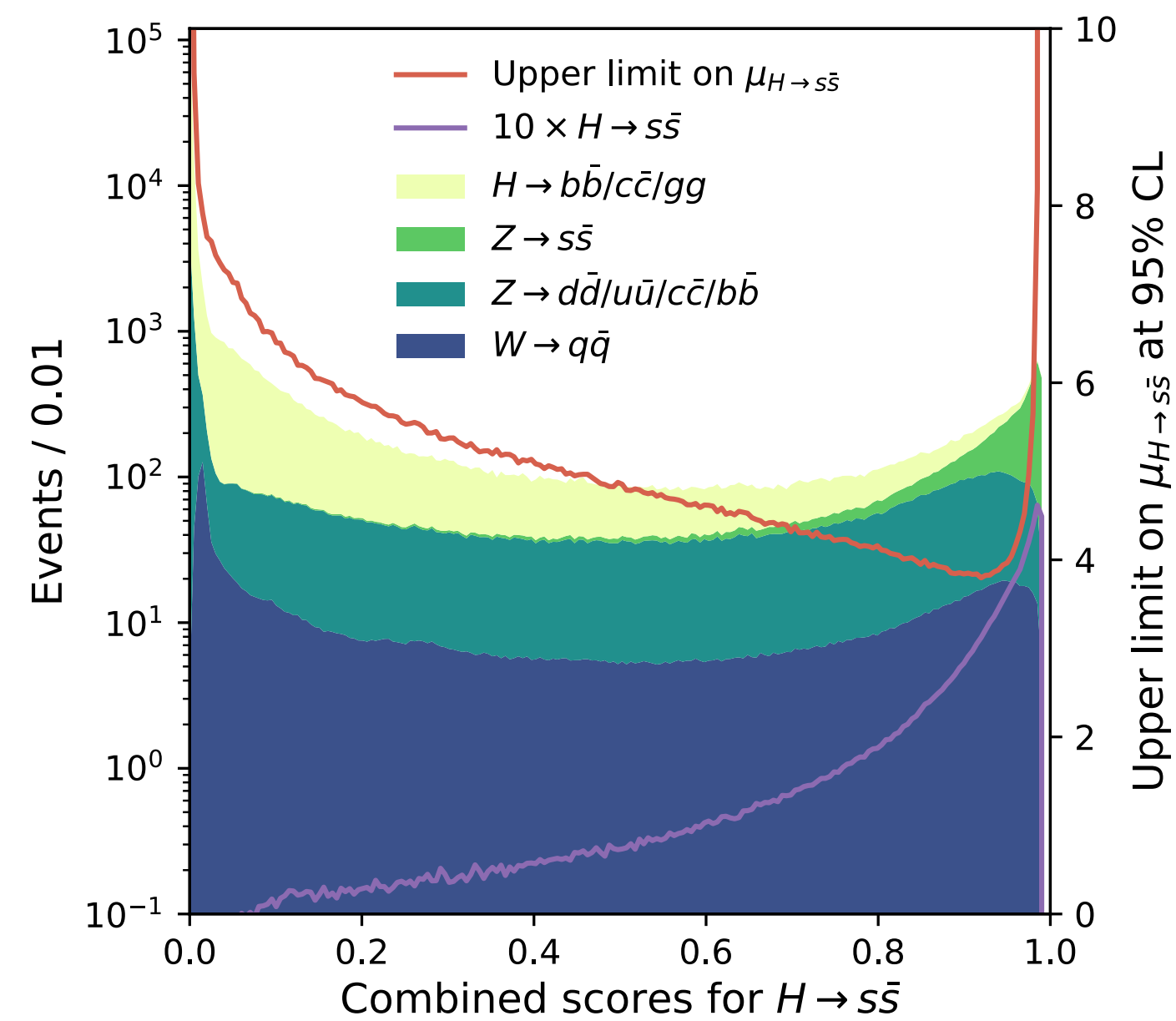
# Machine learning-based jet-origin identification

- We perform a series of companion studies to understand the systematic uncertainties for jet-origin identification
- We found the performance are consistent between different physics processes (left) and comparable between different hadronization models (right)



# JOI's application at an electron-positron Higgs factory

- We apply the jet-origin identification to Higgs rare and exotic decay measurements at CEPC



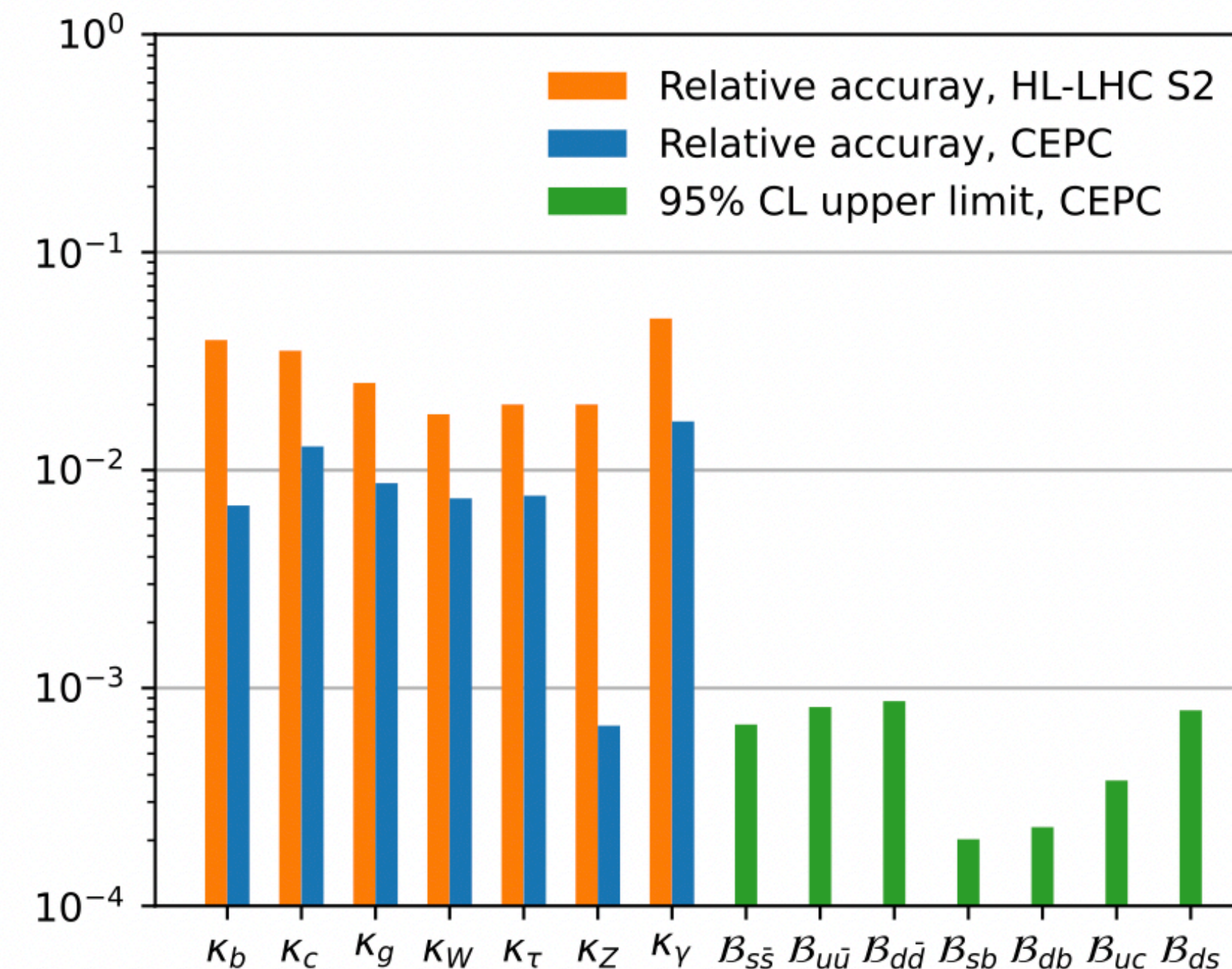


# JOI's application at an electron-positron Higgs factory

- The upper limits at 95% confidence level on the branching ratios of  $H \rightarrow ss, uu, dd$  and  $H \rightarrow sb, db, uc, ds$  can be determined to  $2 \times 10^{-4}$  to  $1 \times 10^{-3}$ , which are greatly improved upon previous studies

TABLE I. Summary of background yields from  $H \rightarrow b\bar{b}/c\bar{c}/gg, Z$ , and  $W$  prior to the flavor-based event selection, along with the expected upper limits on Higgs decay branching ratios at 95% CL under the background-only hypothesis.

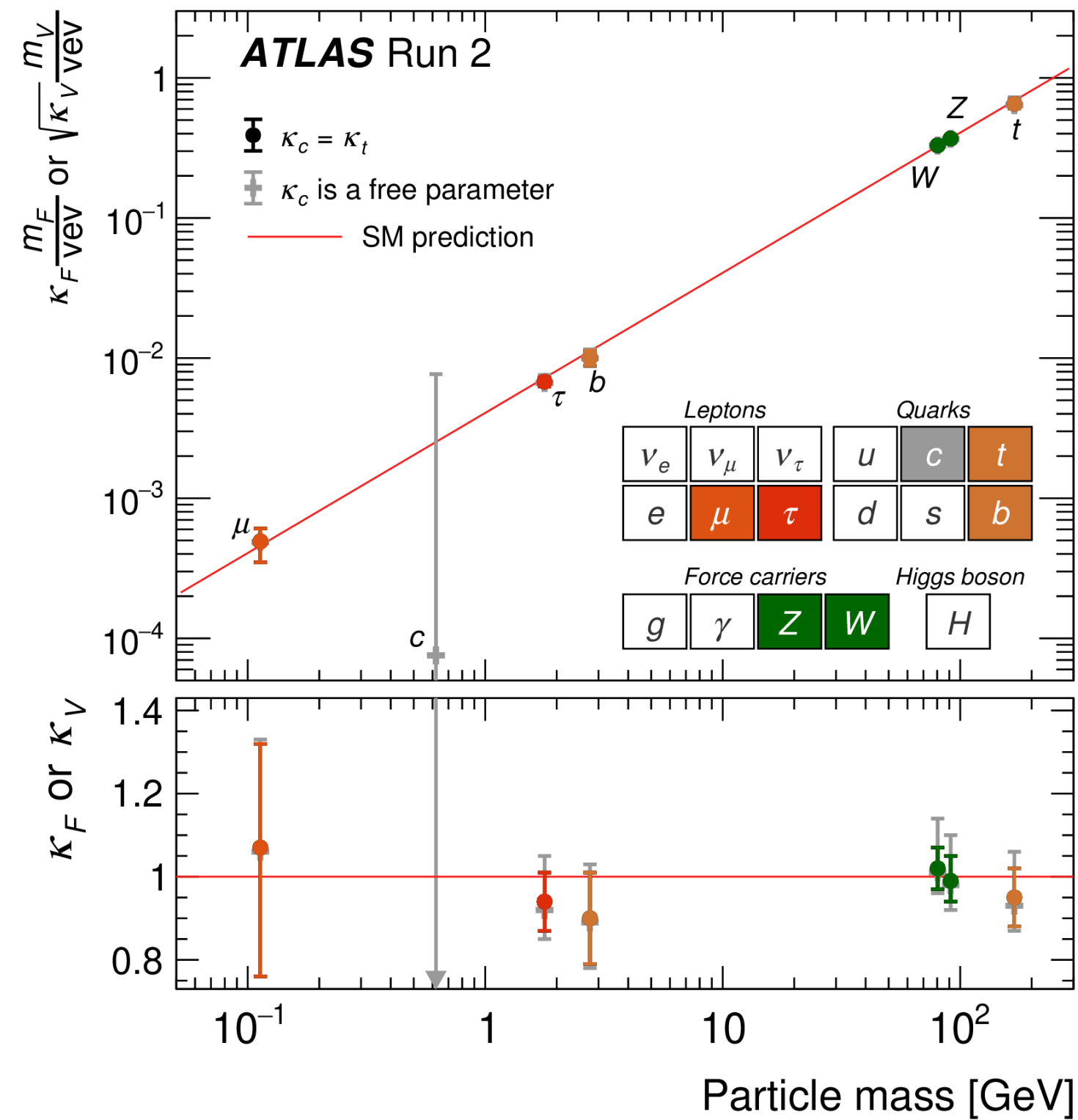
	Bkg ( $10^3$ )			Upper limits on Br ( $10^{-3}$ )						
	$H$	$Z$	$W$	$s\bar{s}$	$u\bar{u}$	$d\bar{d}$	$sb$	$db$	$uc$	$ds$
$\nu\bar{\nu}H$	151	20	2.1	0.81	0.95	0.99	0.26	0.27	0.46	0.93
$\mu^+\mu^-H$	50	25	0	2.6	3.0	3.2	0.5	0.6	1.0	3.0
$e^+e^-H$	26	16	0	4.1	4.6	4.8	0.7	0.9	1.6	4.3
Comb.	...	...	...	0.75	0.91	0.95	0.22	0.23	0.39	0.86



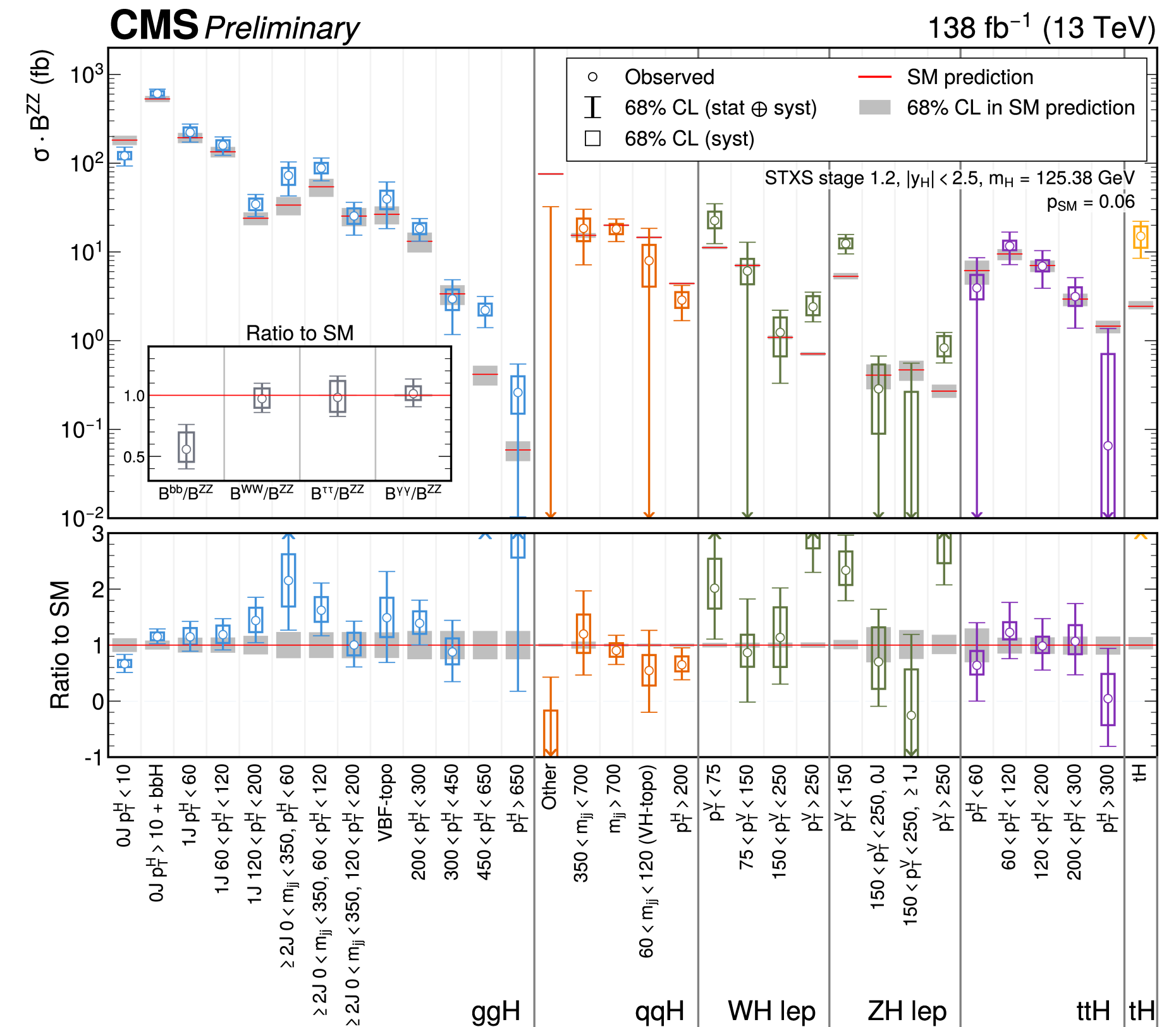
# **3. Some Higgs and new physics results that benefit from better jet reconstruction**

In this part, I will only discuss LHC results that I contributed to

# Higgs couplings with other fundamental particles



[Nature 607 \(2022\) 52-59](#)

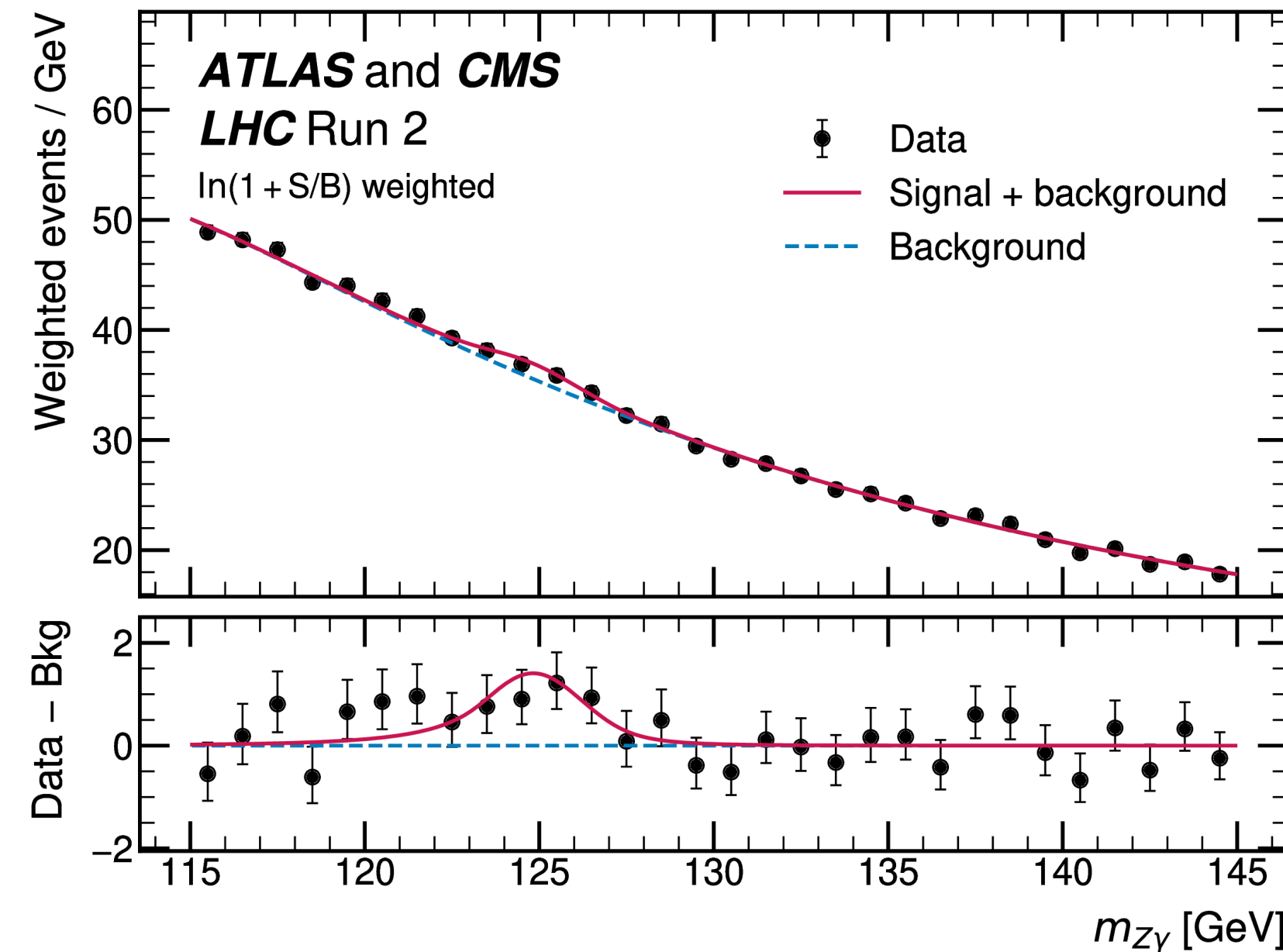
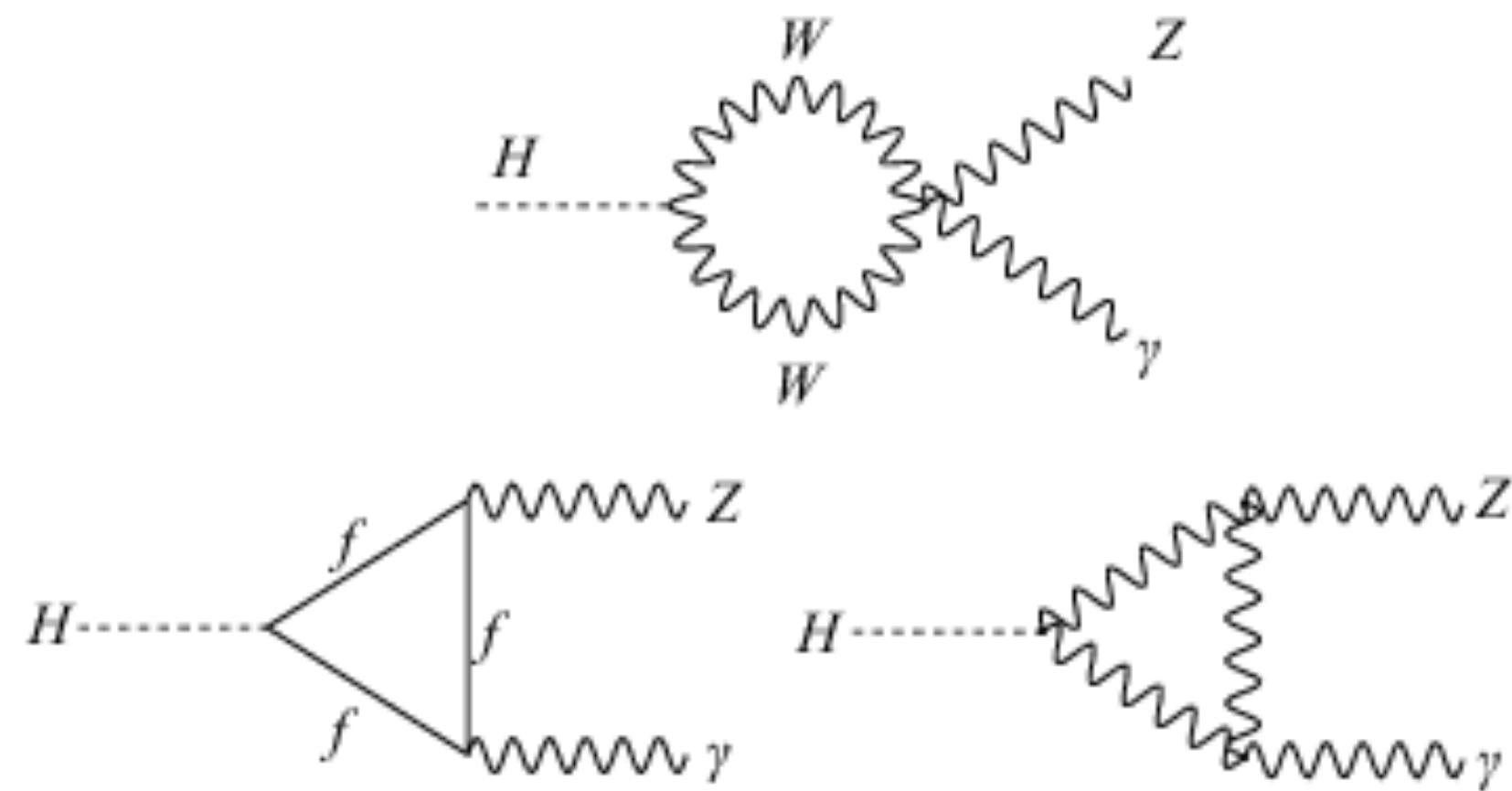


[Moriond 2025](#)

- “Kappa” framework: assign **coupling modifier** to each **interaction vertex** (e.g.  $\kappa_t$ , ...)
- “STXS” framework: split **production cross sections** into various **phase space regions**
- LHC Run 2: good agreement with the SM



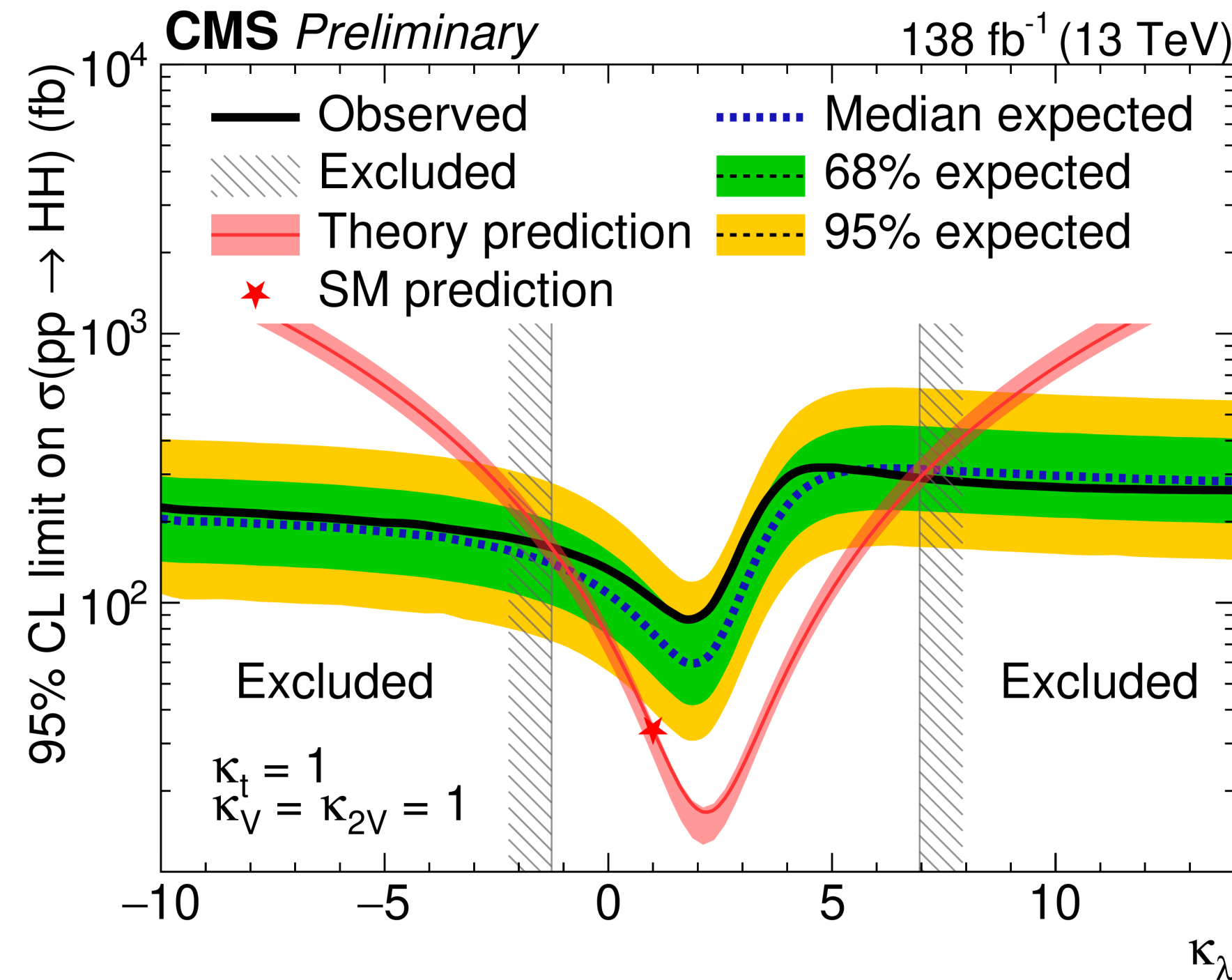
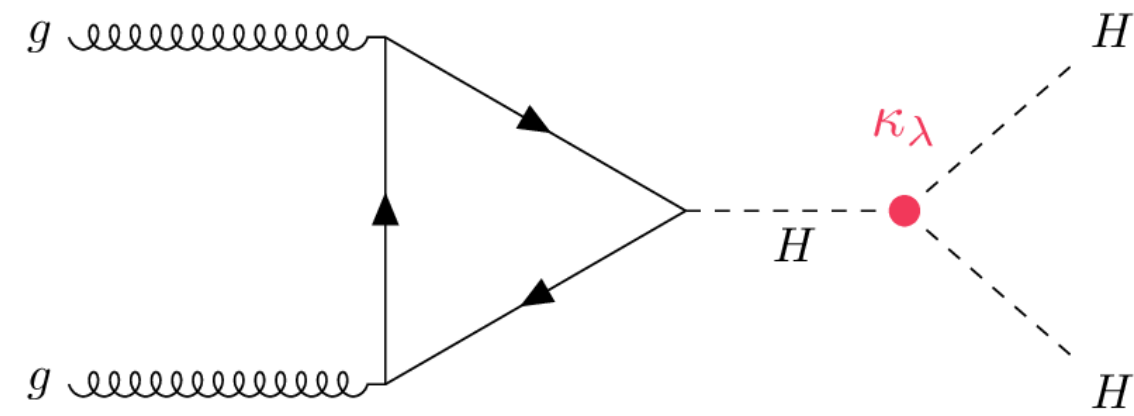
# H→Zγ decay



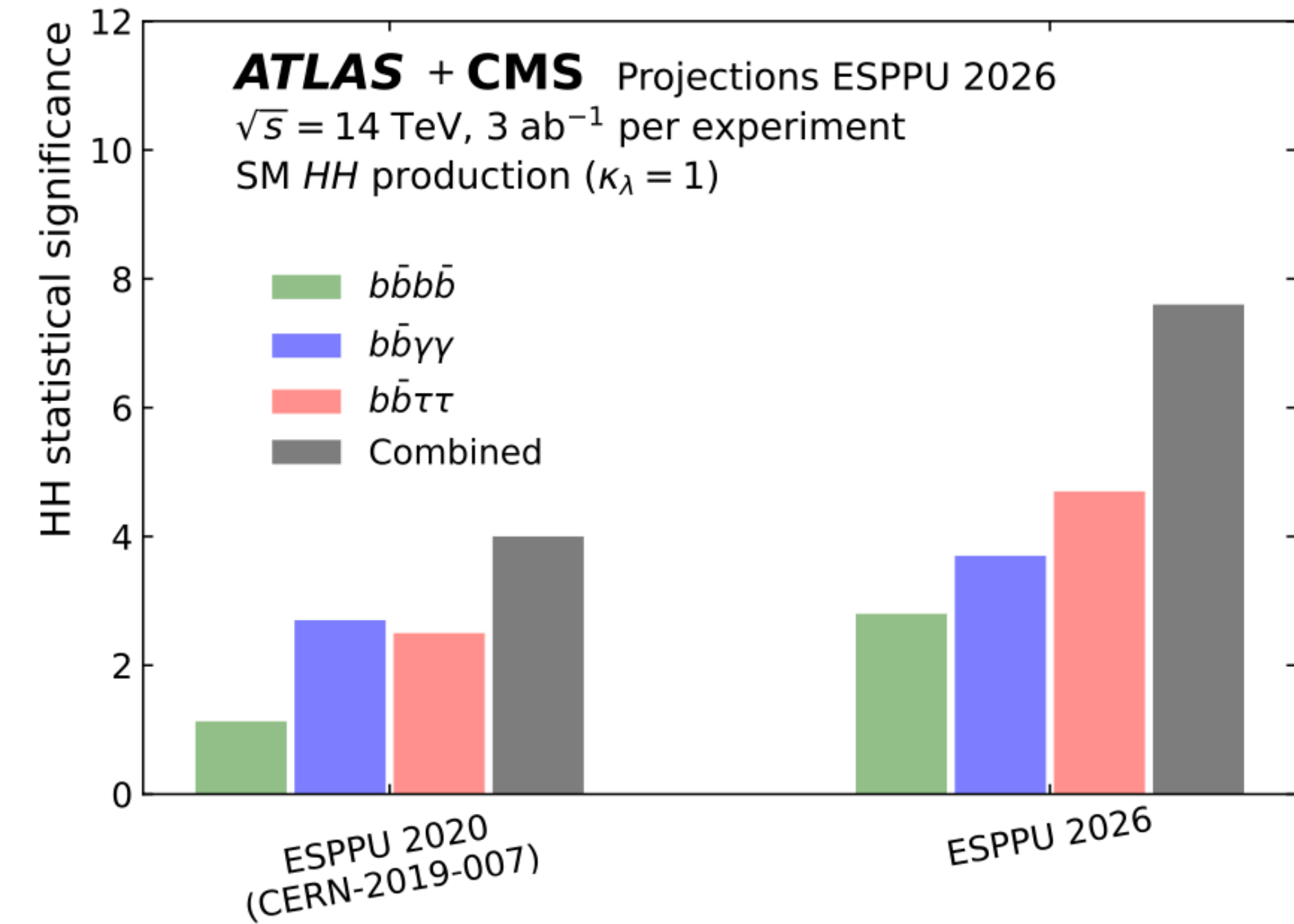
[Phys. Rev. Lett. 132 \(2024\) 021803](#),  
Editor's Suggestion,  
Featured in Physics,  
Collection of the Year

- The observed H→Zγ significance in ATLAS+CMS combined result is **3.4σ** (expected 1.6σ)
- **First evidence** of the H→Zγ decay
- VBF channel (with 2 forward quark jets) is crucial

# Higgs self-couplings



[CERN Seminar 2024](#)

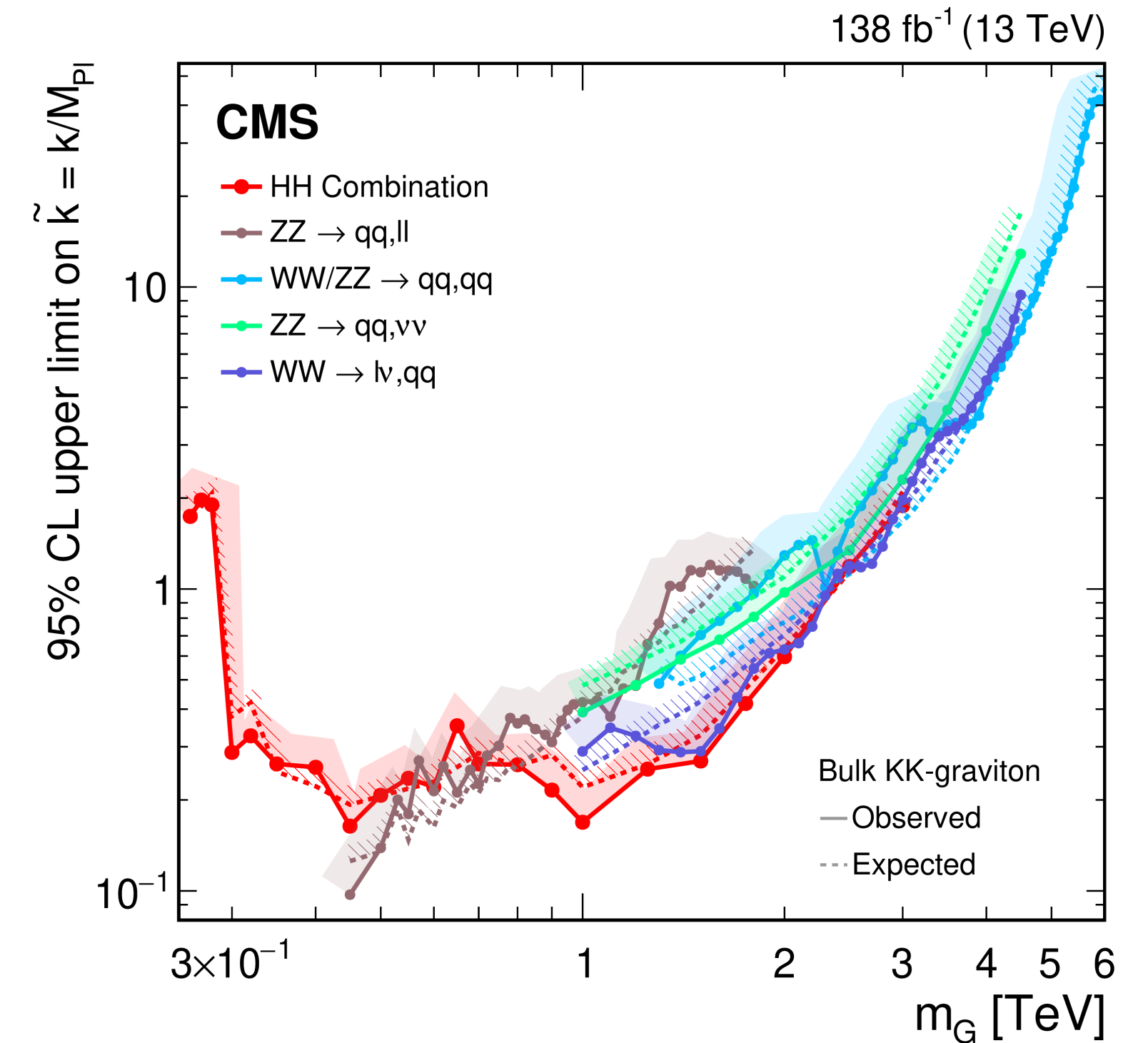
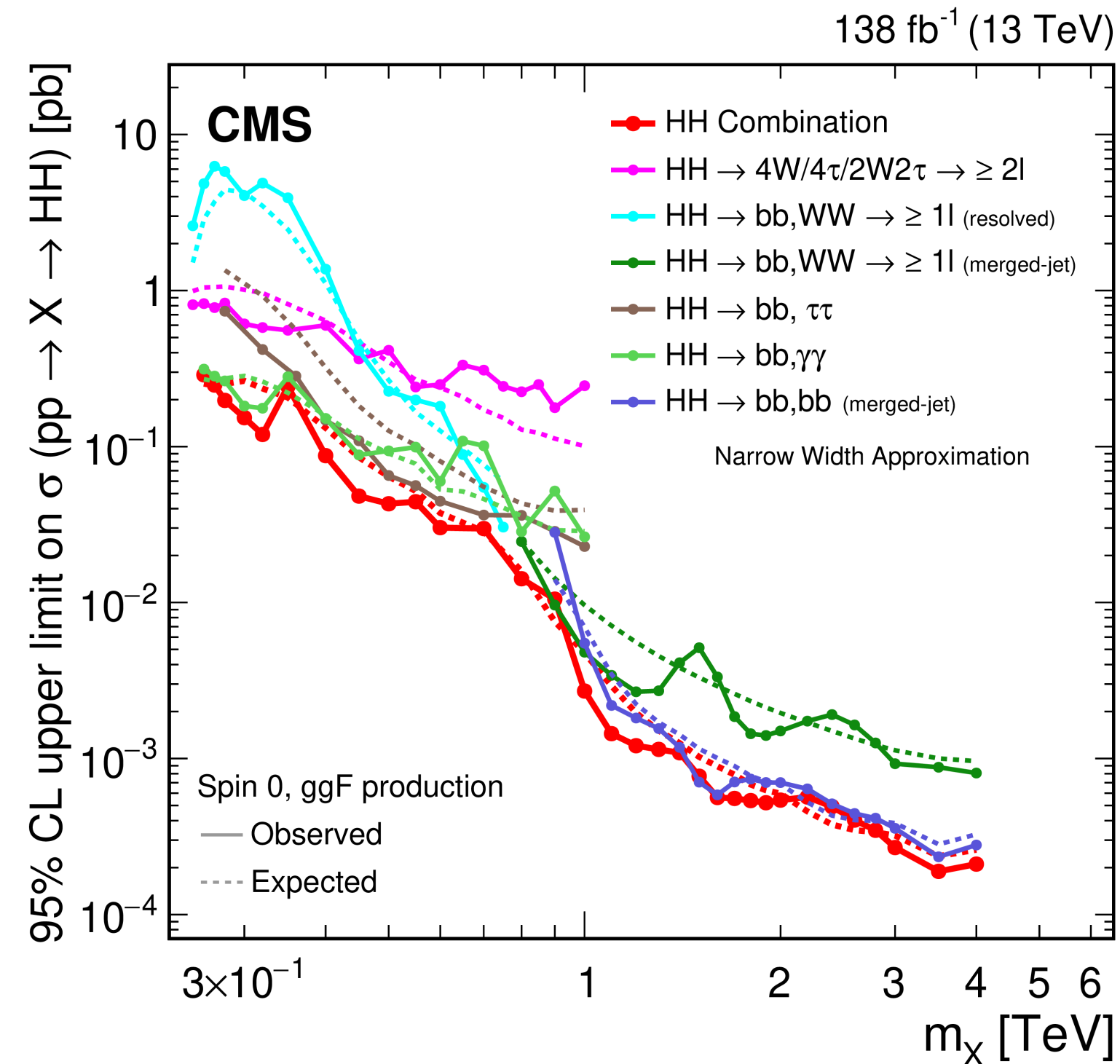
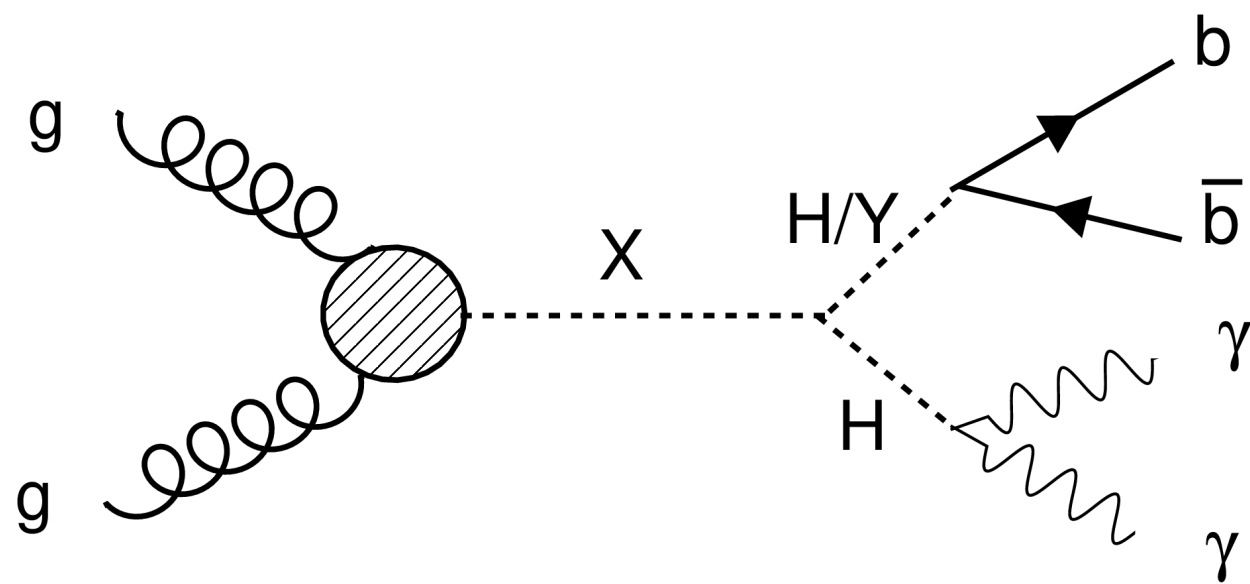


[arxiv:2504.00672](#)

- **Higgs self-coupling may provide a portal to new physics beyond it**
  - Vacuum stability, early universe evolution, ...
- **Double Higgs production is the way to directly probe Higgs self-couplings at the LHC**
  - B-jet tagging is a key
- LHC Run 2:  $-1.4 < \kappa_\lambda < 7.0$  (CMS);  $-1.2 < \kappa_\lambda < 7.2$  (ATLAS)
- HL-LHC:  $0.5 < \kappa_\lambda < 1.6$  (ATLAS+CMS)

# Resonant production of Higgs boson

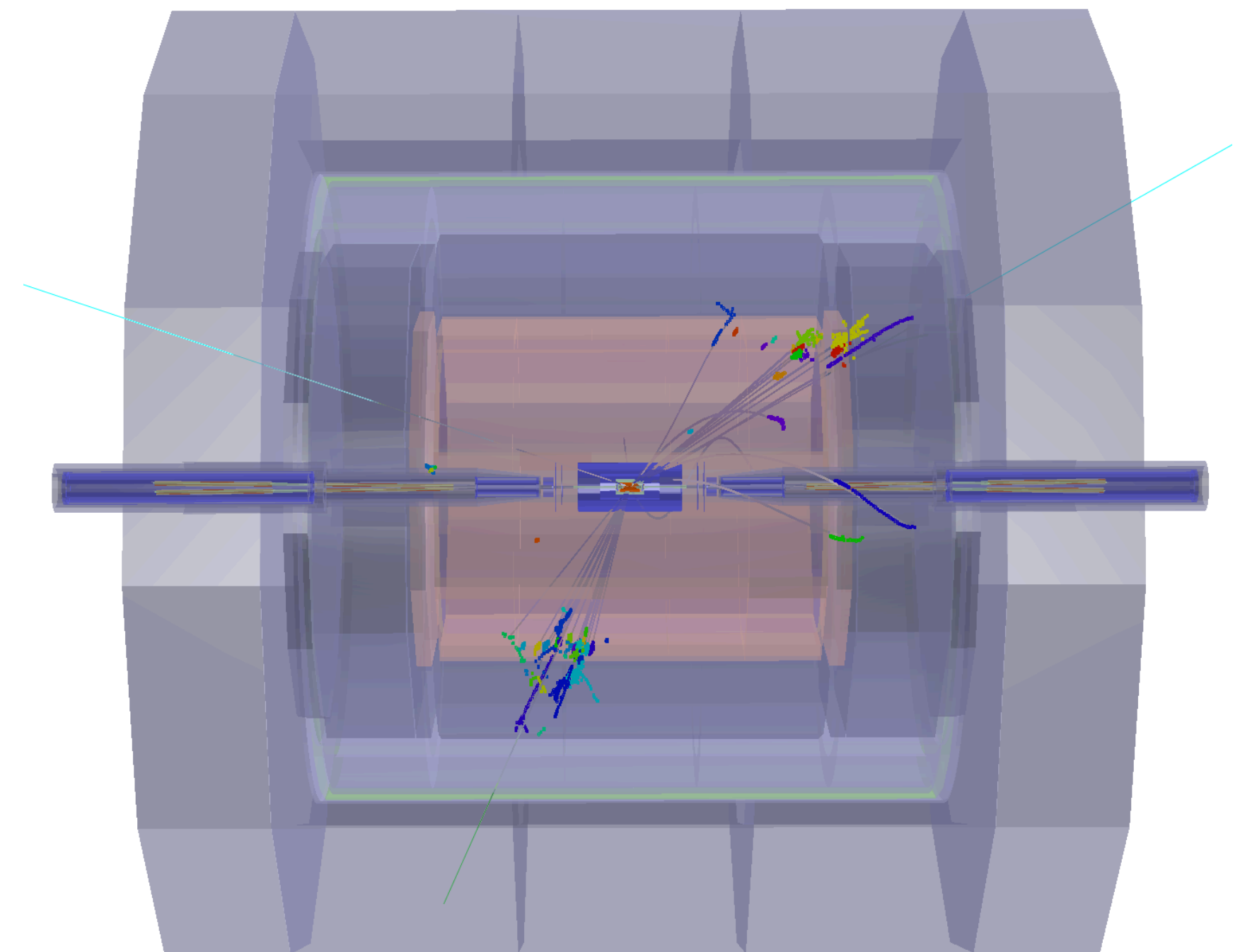
- Heavy resonance search channels including at least one Higgs boson plus another particle have formed an important part of the new physics search program at the LHC
- Combining searches channels at CMS, we obtain cross section limits on resonant production of Higgs boson and constraints on relevant new physics models





# Summary

- **Machine learning and quantum computing are becoming important tools for high energy physics**
- We show some applications in jet reconstruction, which can boost different physics studies (Higgs, QCD, BSM...)
- And there will probably be much more than what we know today



**Thank you!**