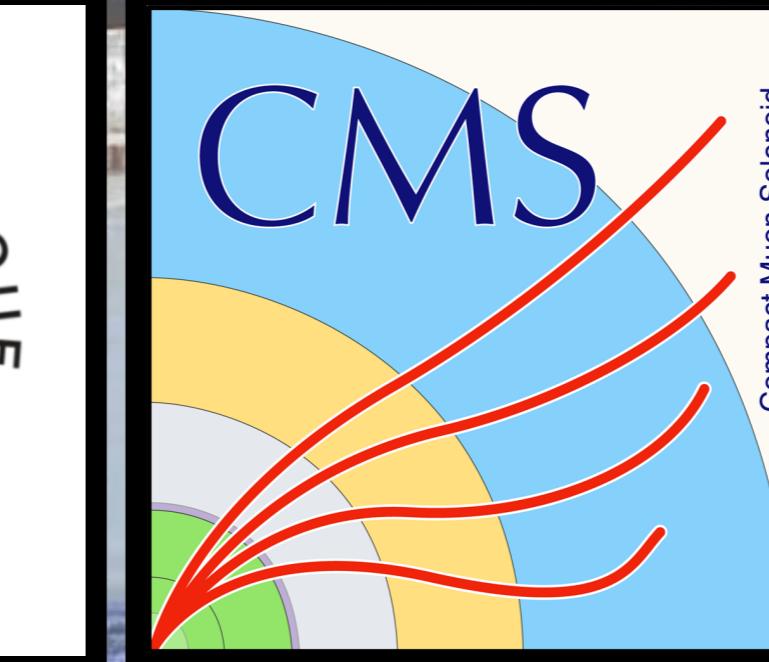
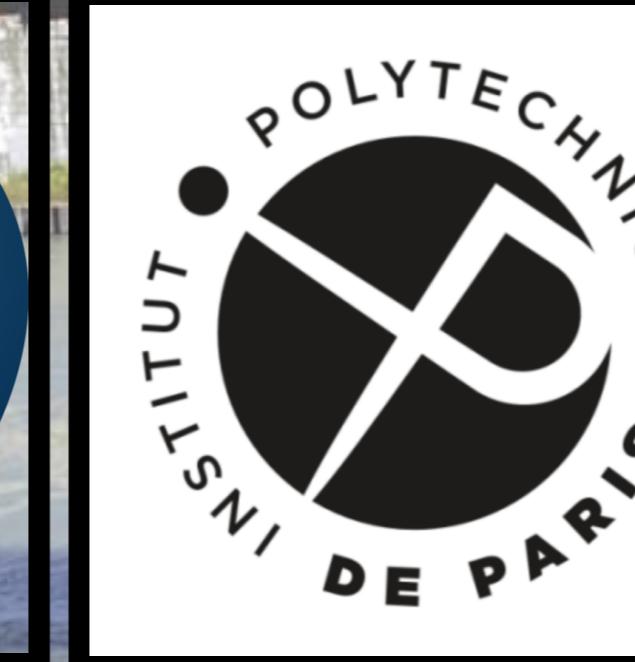
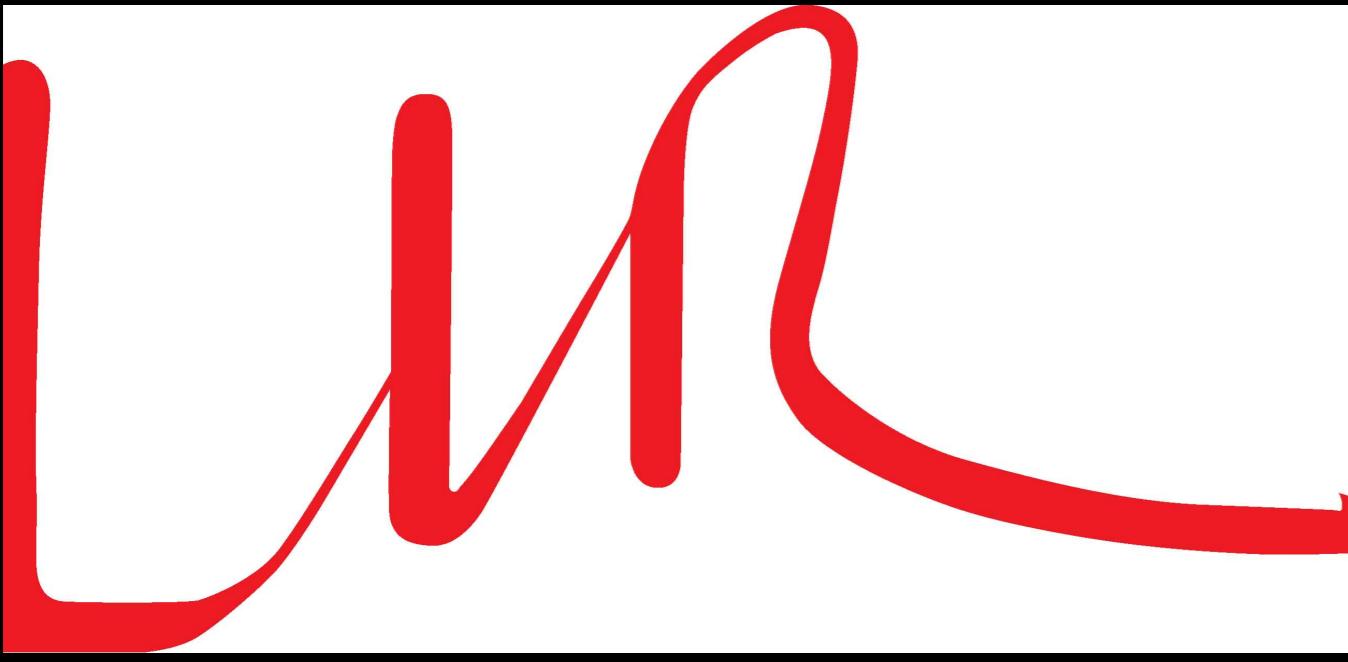


# Di-Higgs Searches @ CMS

Bruno Alves, for the CMS Collaboration



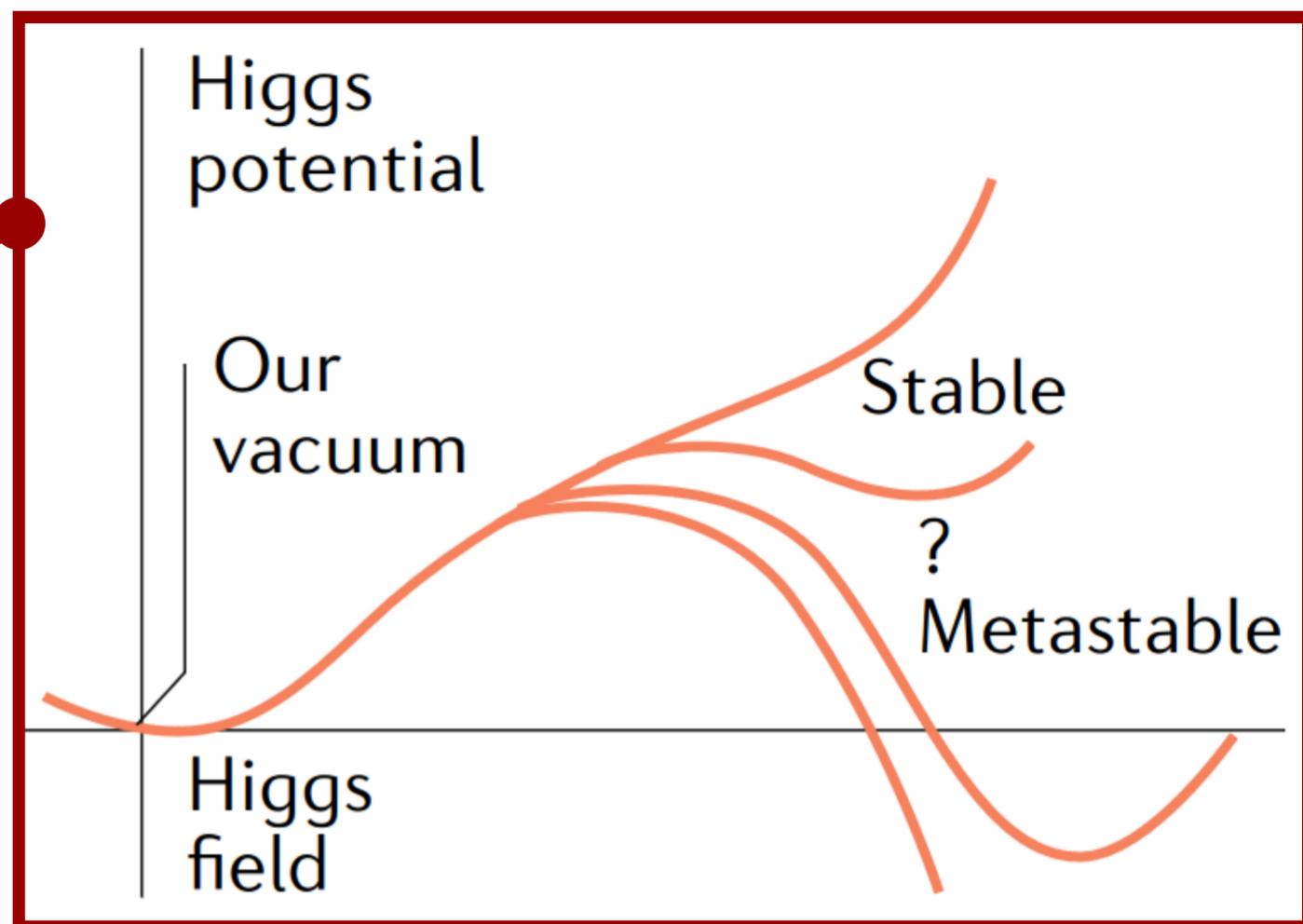
Indico  
9/04/2024  
DIS Grenoble

# Motivation: HH non-res. searches

Measurement of the Higgs boson self-interaction

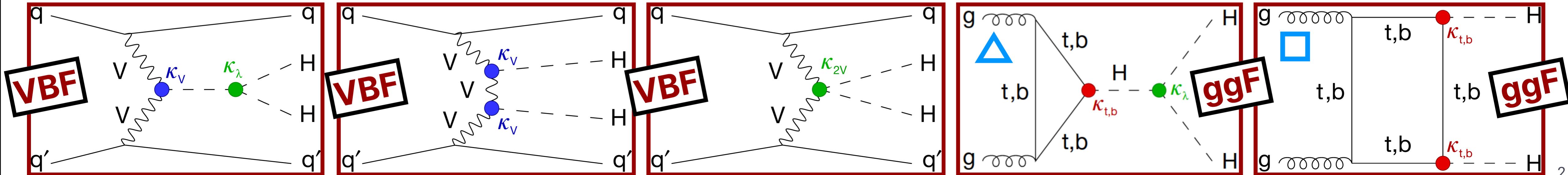
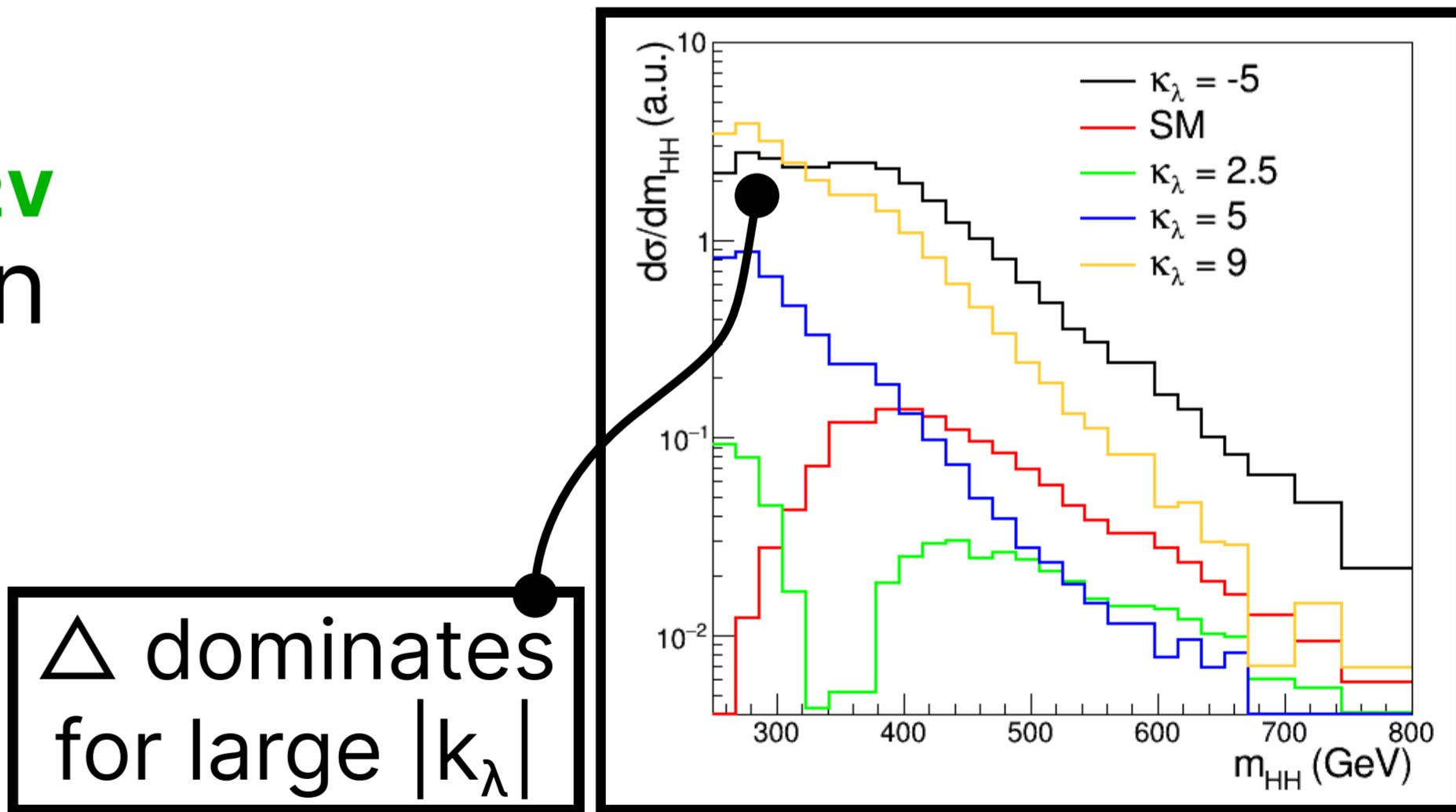
- test vacuum stability
- SM closure:  $\lambda_{HHH}$  is not free!
- deviations can explain baryogenesis

$$\lambda_{HHH}^{\text{SM}} = \frac{m_H^2}{2v^2} \simeq 0.13$$



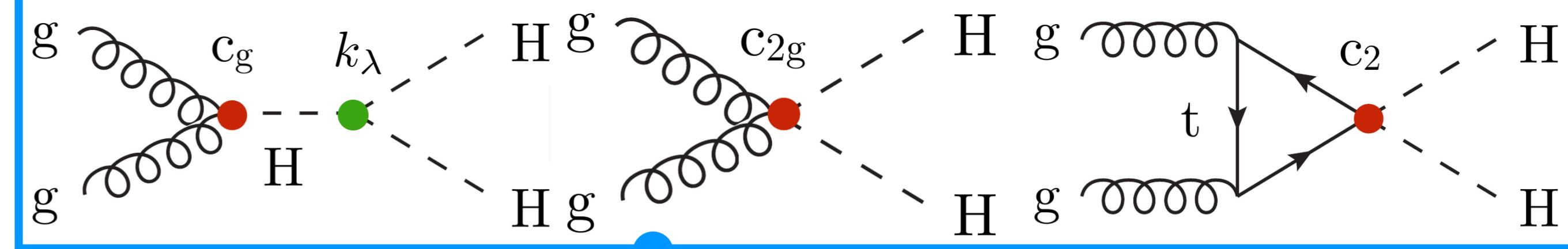
**Test BSM w/ anomalous couplings:**  $k_\lambda$ ,  $k_t$ ,  $k_v$ ,  $k_{2v}$

- deviations may lead to diffs. in HH production rates and kinematics
- small SM HH cross-section because of  $\Delta-\square$  interference



# Motivation: EFT

ggF



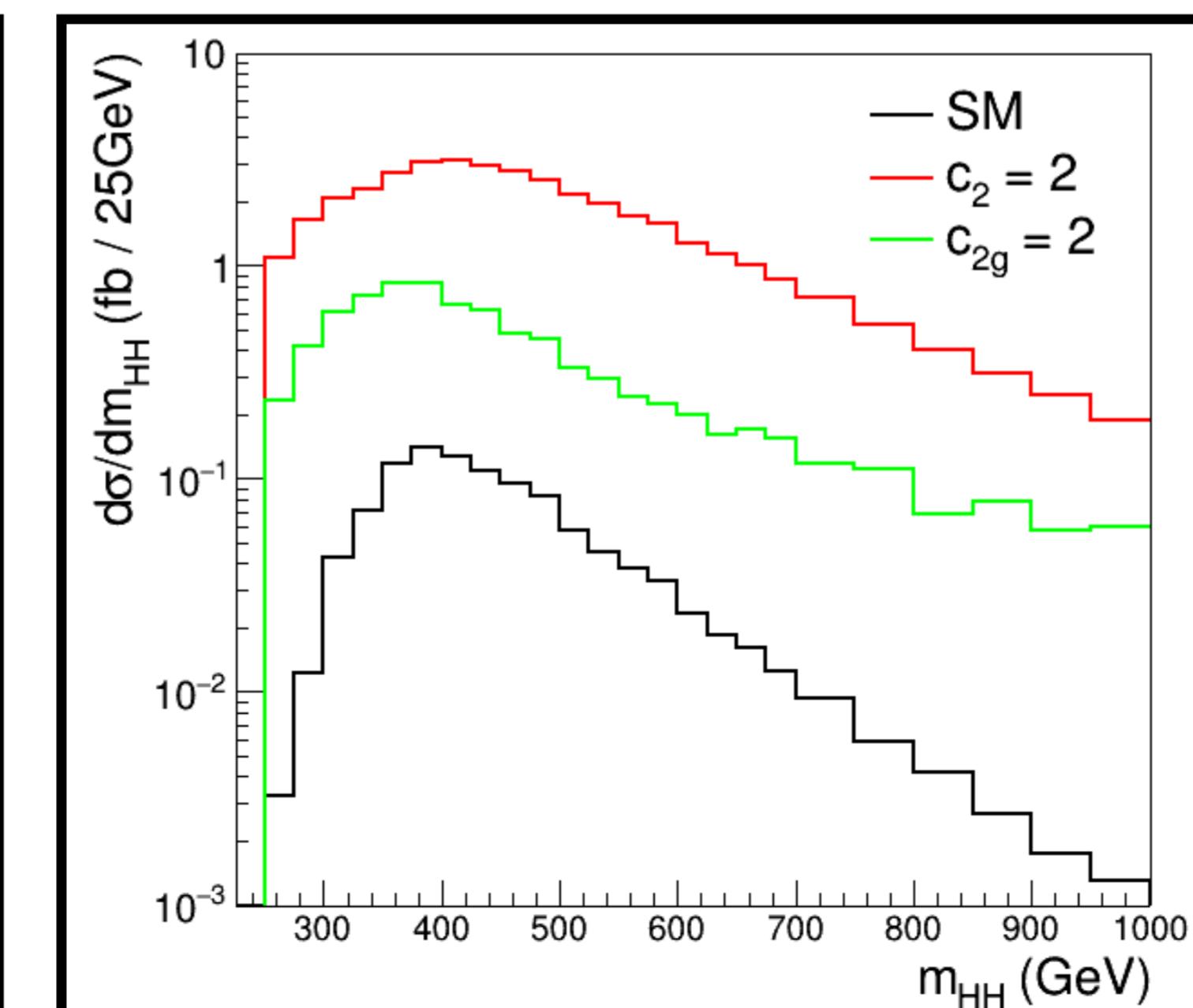
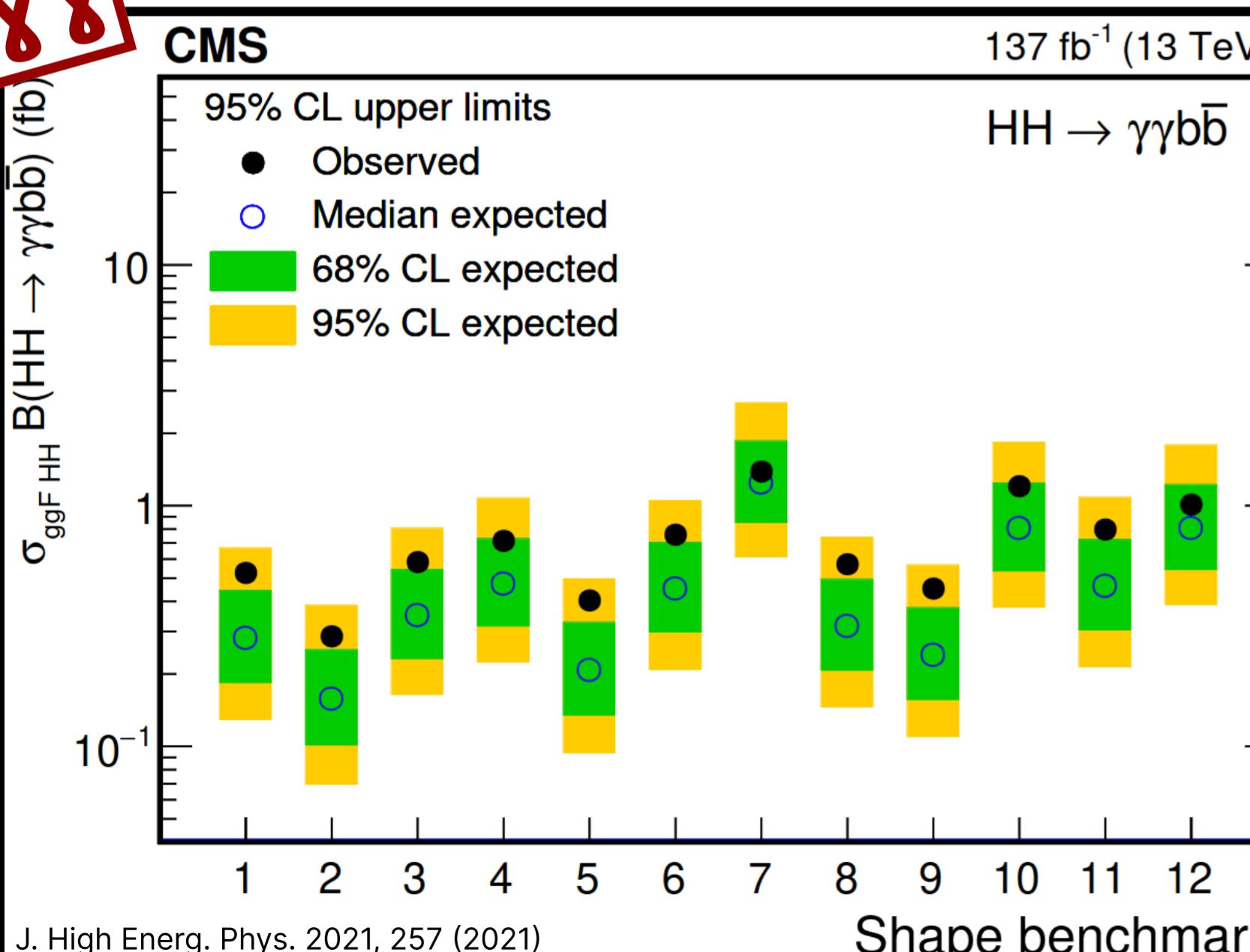
- Probe effects from resonances with very high mass at lower energy scales

- BSM effects on ggF HH production can be studied through EFT model with **three new couplings**

- Explore sensitivity to BSM EFT couplings with **EFT benchmark points**:

- based on test statistic measuring kinematics' similarity
- allow extrapolation between different points

bb̄γγ



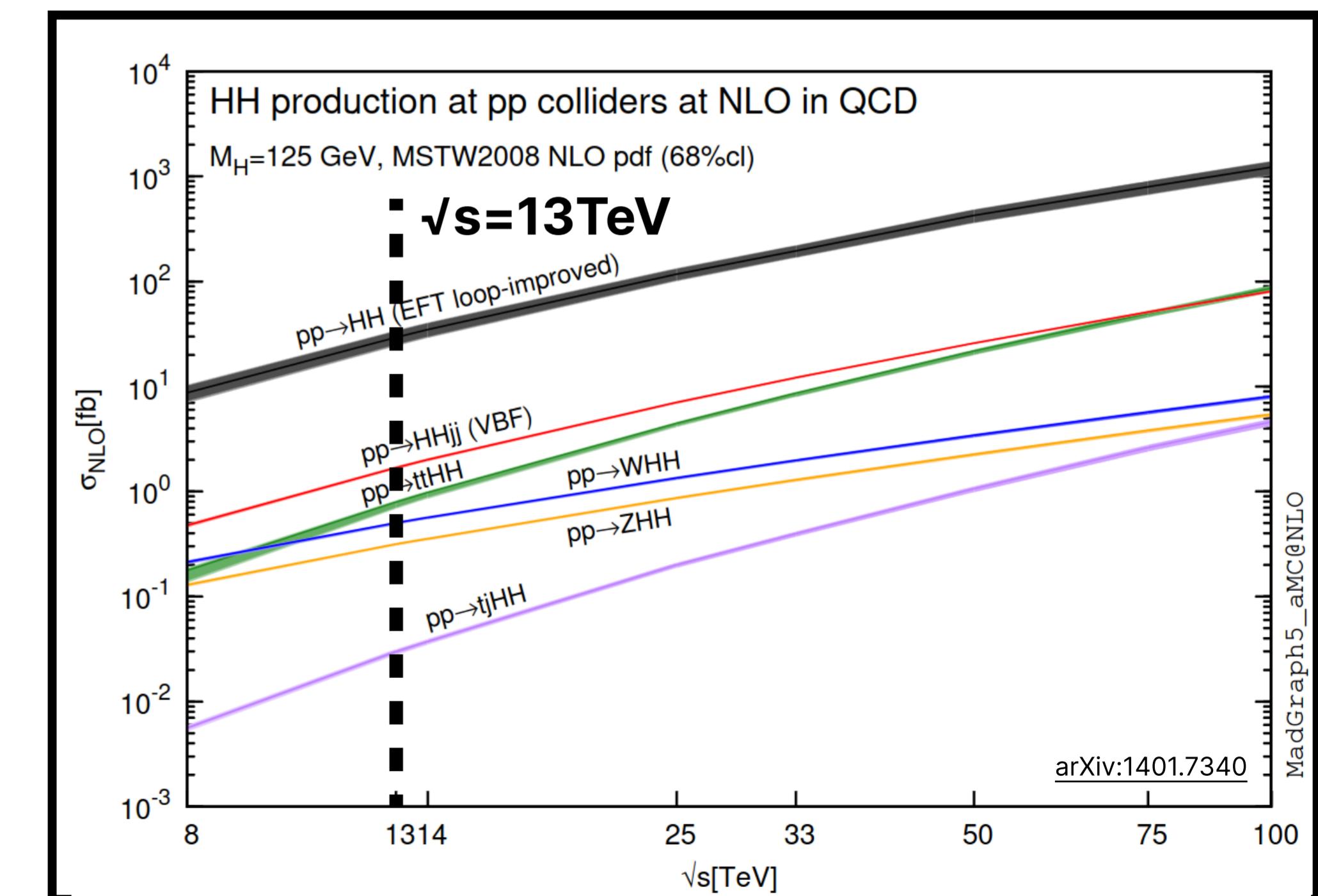
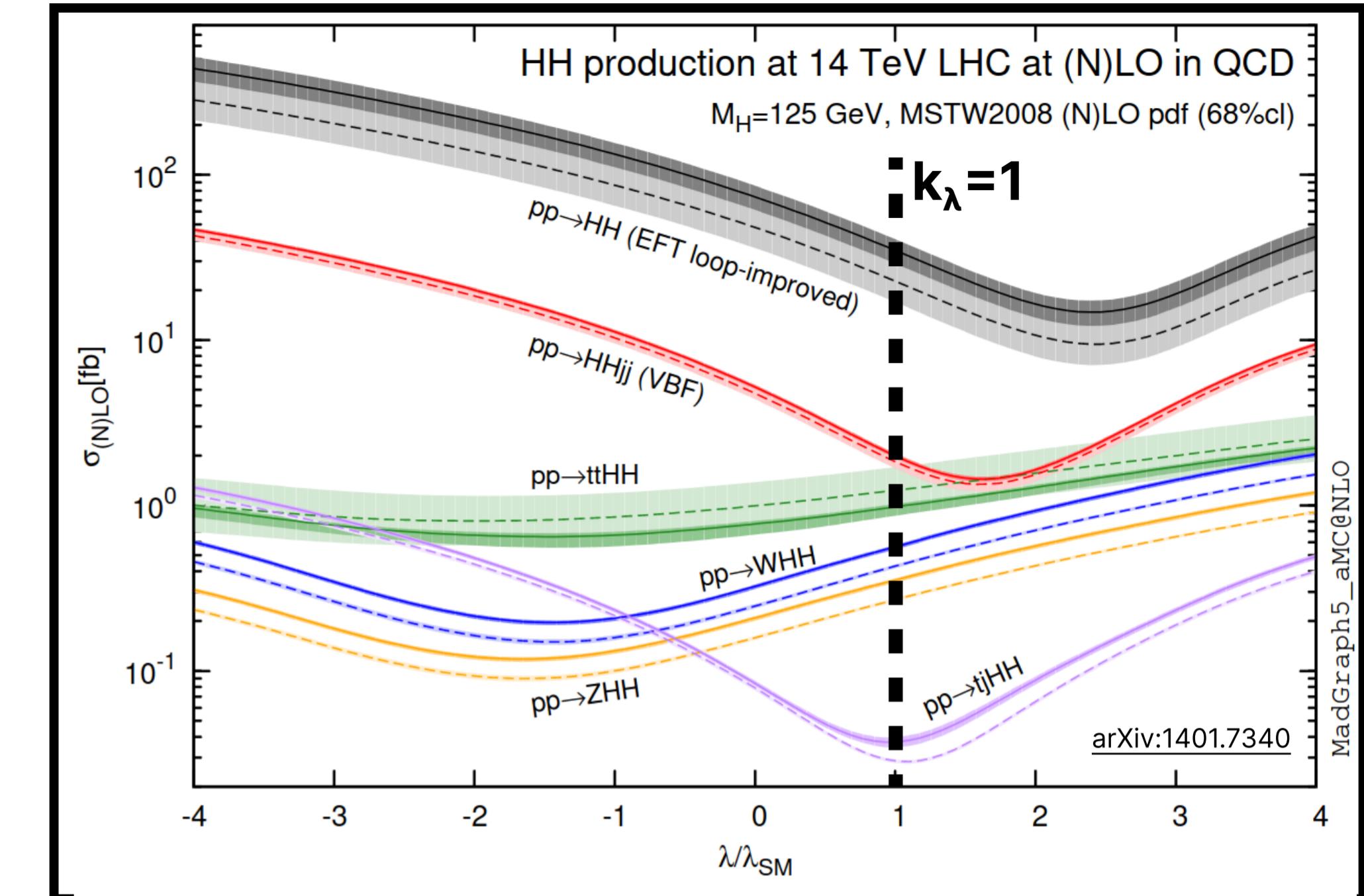
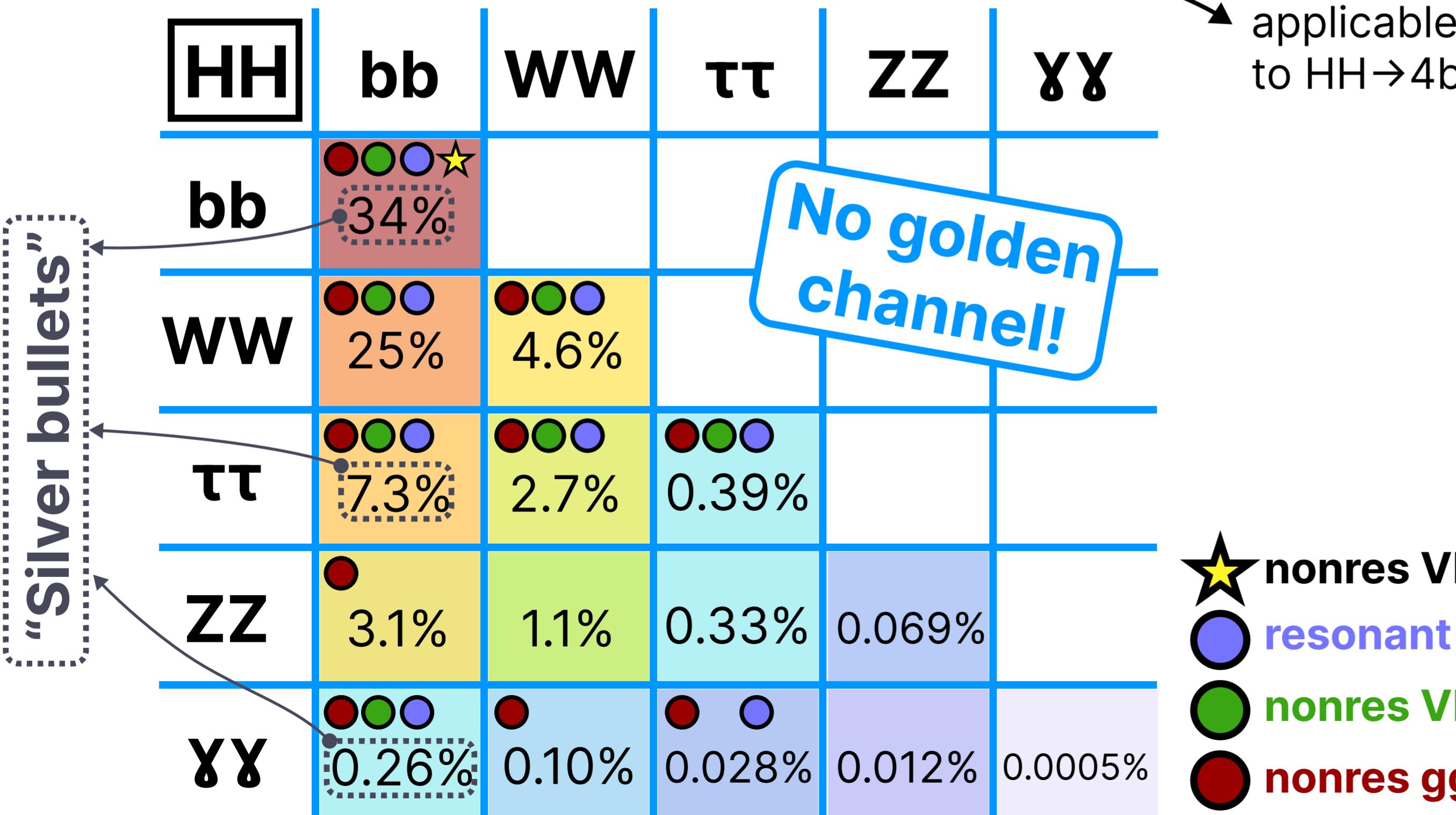
Benchmarks

	$\kappa_\lambda$	$\kappa_t$	$c_2$	$c_g$	$c_{2g}$
SM	1.0	1.0	0.0	0.0	0.0
1	7.5	1.0	-1.0	0.0	0.0
2	1.0	1.0	0.5	-0.8	0.6
3	1.0	1.0	-1.5	0.0	-0.8
4	-3.5	1.5	-3.0	0.0	0.0
5	1.0	1.0	0.0	0.8	-1
6	2.4	1.0	0.0	0.2	-0.2
7	5.0	1.0	0.0	0.2	-0.2
8	15.0	1.0	0.0	-1	1
9	1.0	1.0	1.0	-0.6	0.6
10	10.0	1.5	-1.0	0.0	0.0
11	2.4	1.0	0.0	1	-1
12	15.0	1.0	1.0	0.0	0.0
8a	1.0	1.0	0.5	$\frac{0.8}{3}$	0.0
1b	3.94	0.94	$-\frac{1}{3}$	0.75	-1
2b	6.84	0.61	$\frac{1}{3}$	0.0	1.0
3b	2.21	1.05	$-\frac{1}{3}$	0.75	-1.5
4b	2.79	0.61	$\frac{1}{3}$	-0.75	-0.5
5b	3.95	1.17	$-\frac{1}{3}$	0.25	1.5
6b	5.68	0.83	$\frac{1}{3}$	-0.75	-1.0
7b	-0.10	0.94	1.0	0.25	0.5

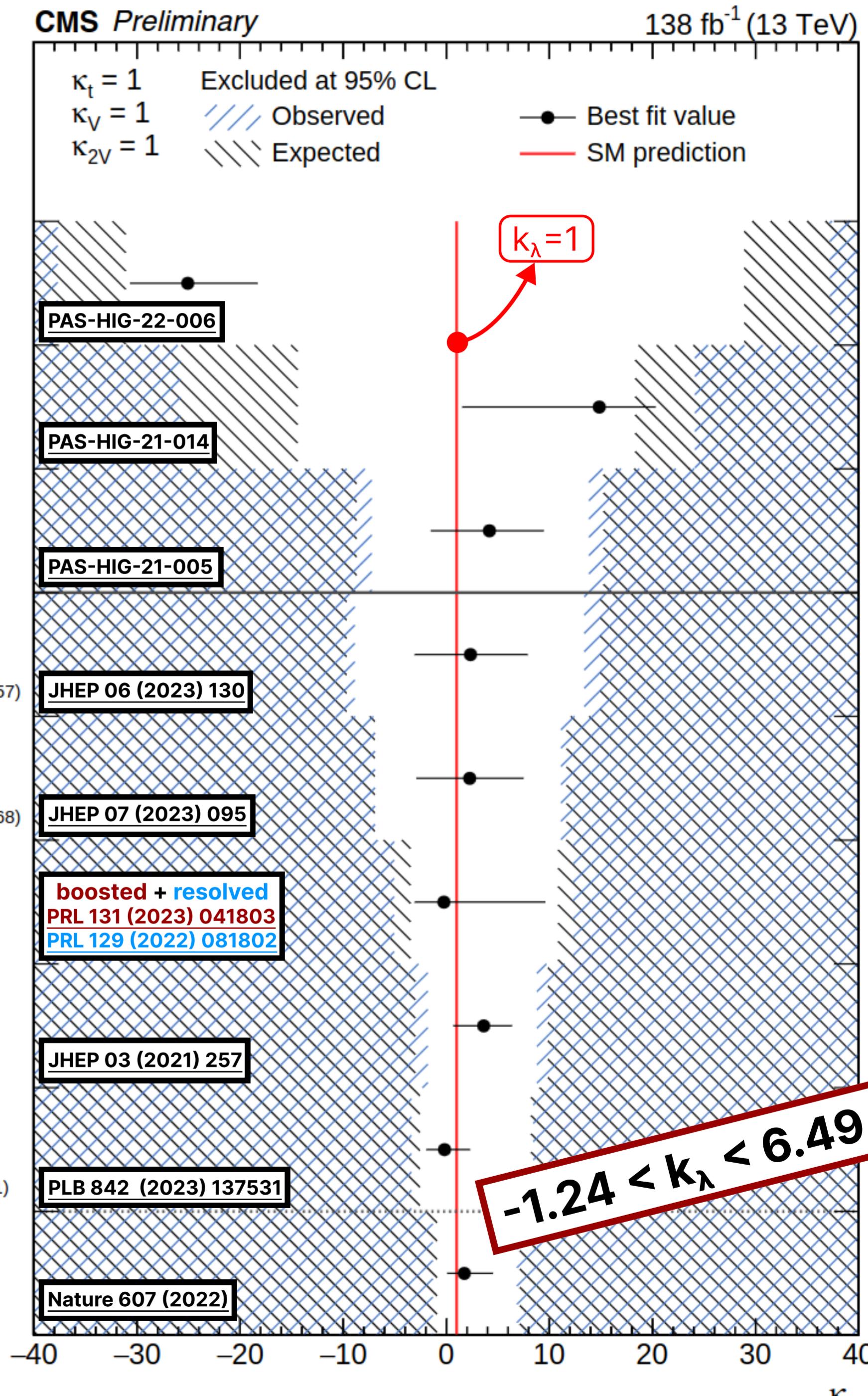
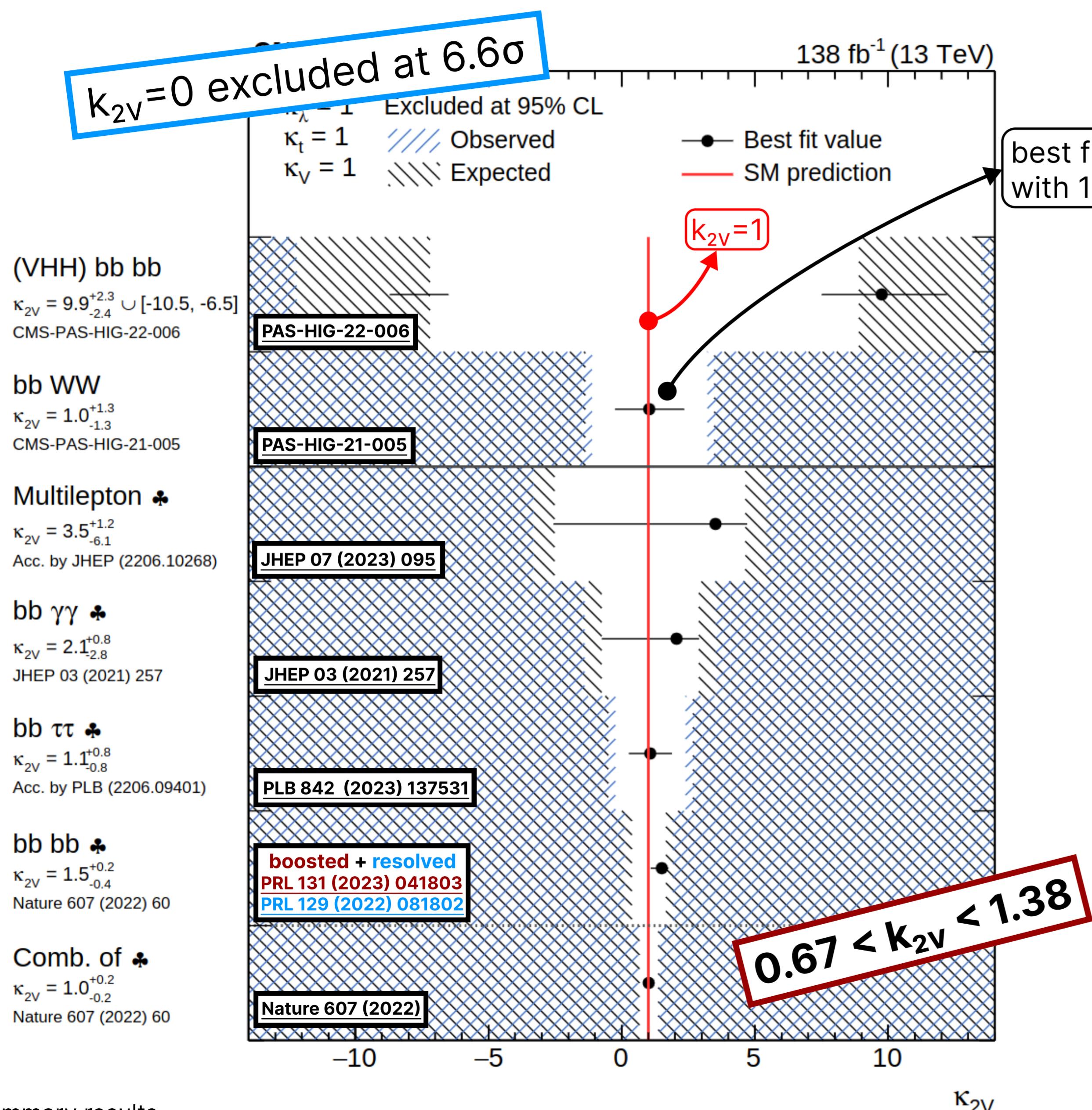
A. Carvalho et al, Capozi et al, Buchalla et al

# Production $\rightarrow$ HH $\rightarrow$ Decay

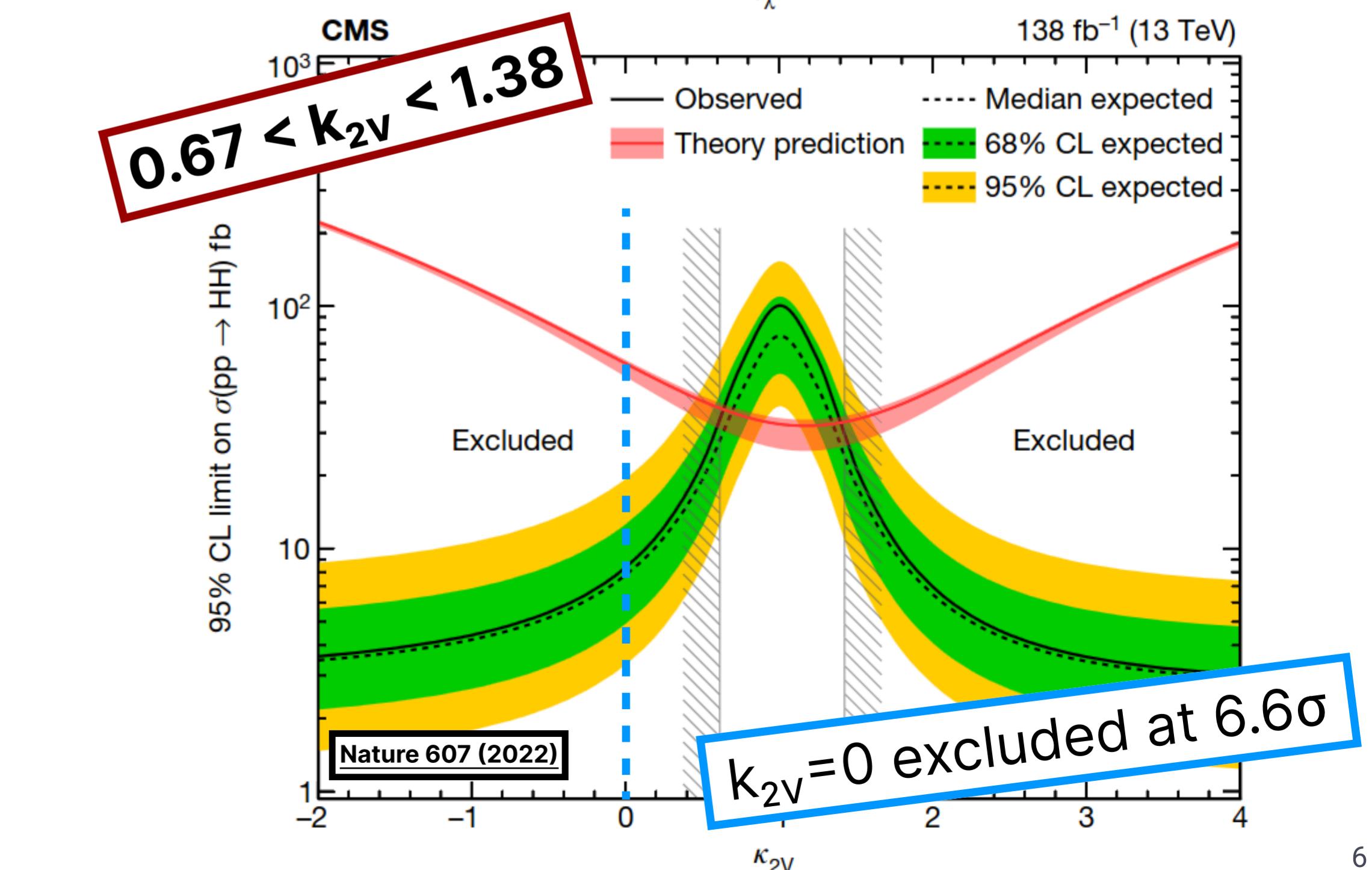
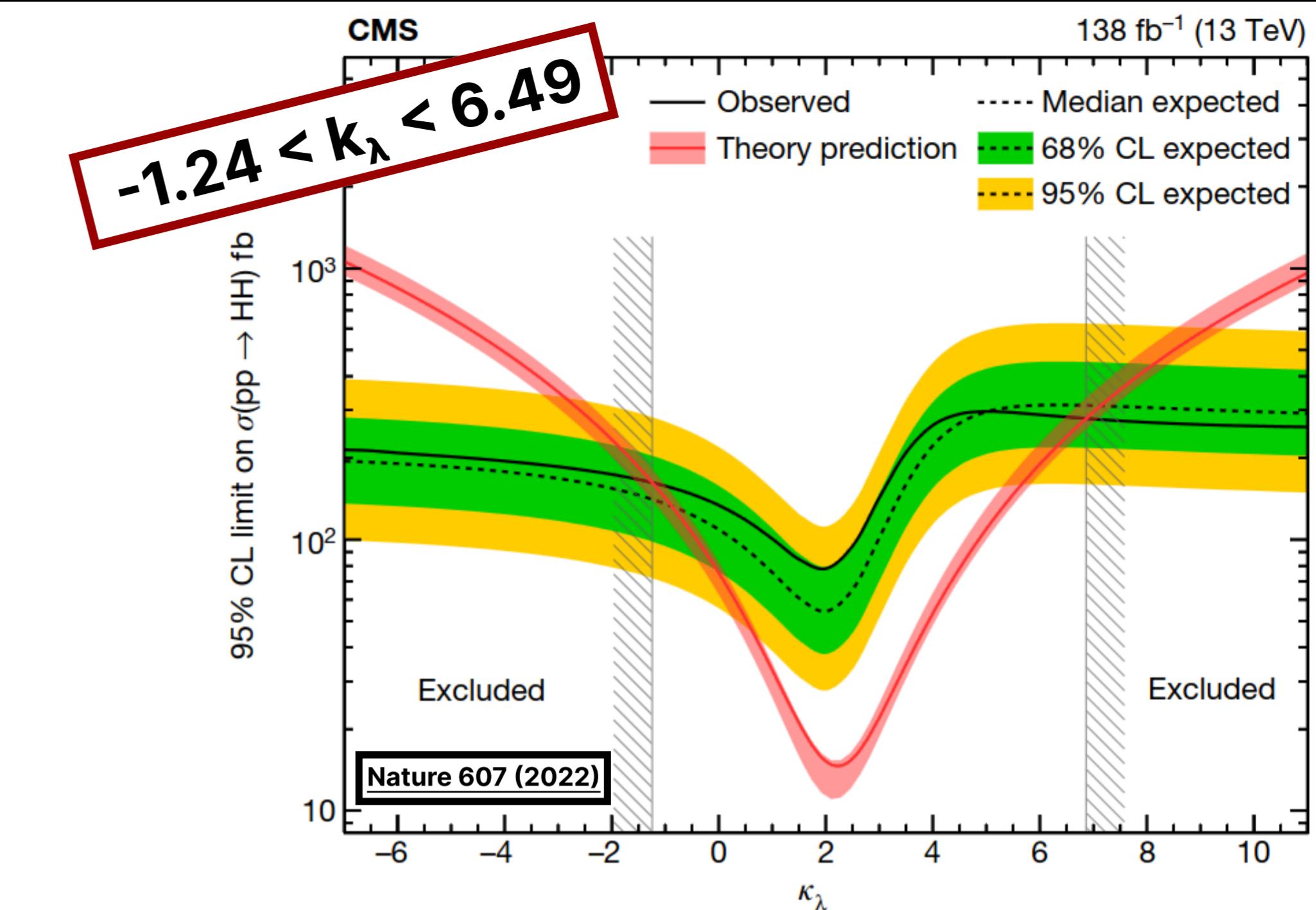
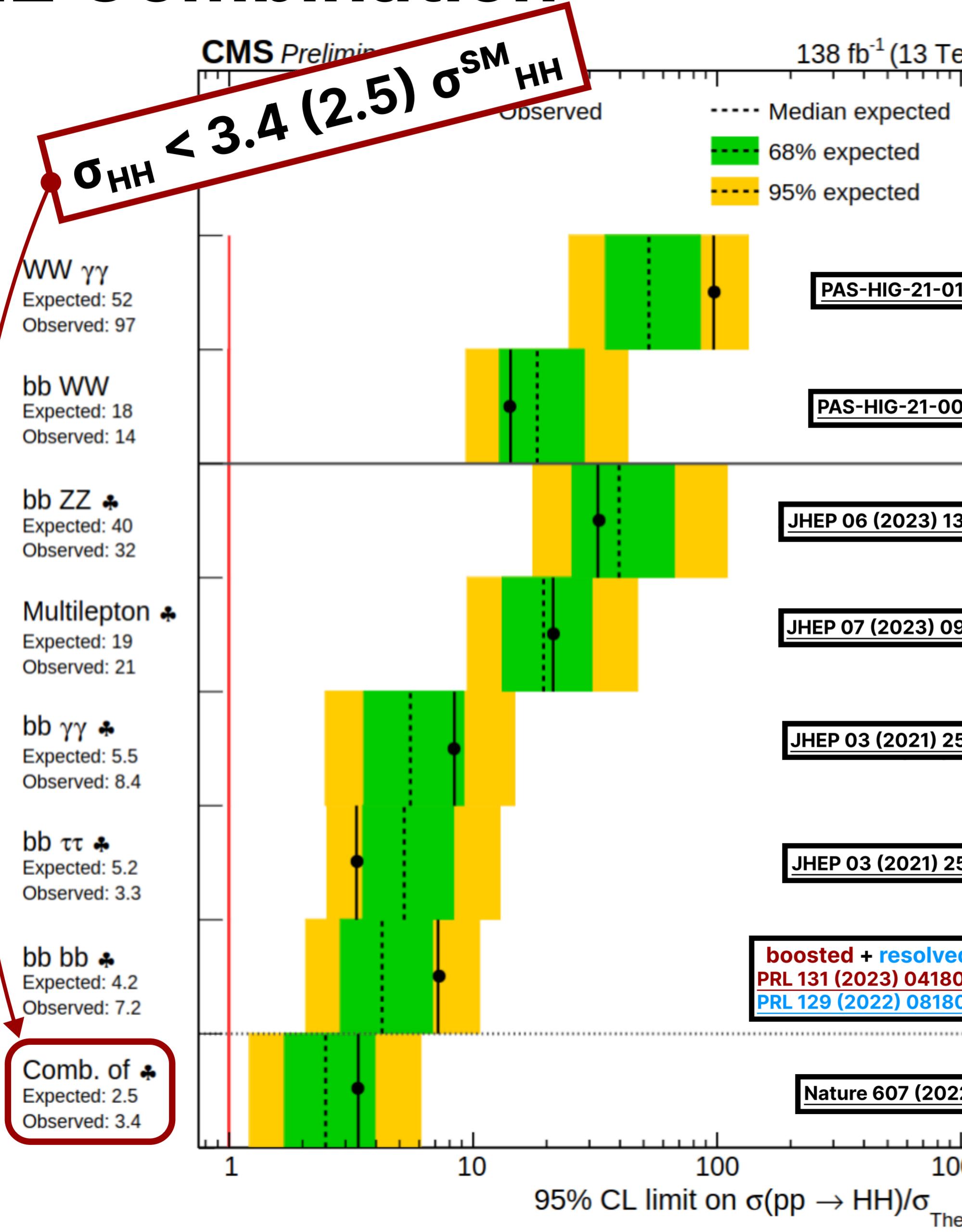
- **ggF, VBF** and **VHH** production
  - $\sim 1000$ x rarer than single-H production
- Large variety of HH channels
  - trade-off between selection purity and branching ratio
- **New:**  $\text{HH} \rightarrow \gamma\gamma\tau\tau$ ,  $\text{ZZ/ZH} \rightarrow 4b$ ,  $\text{H+HH}$



# Run2 Combination



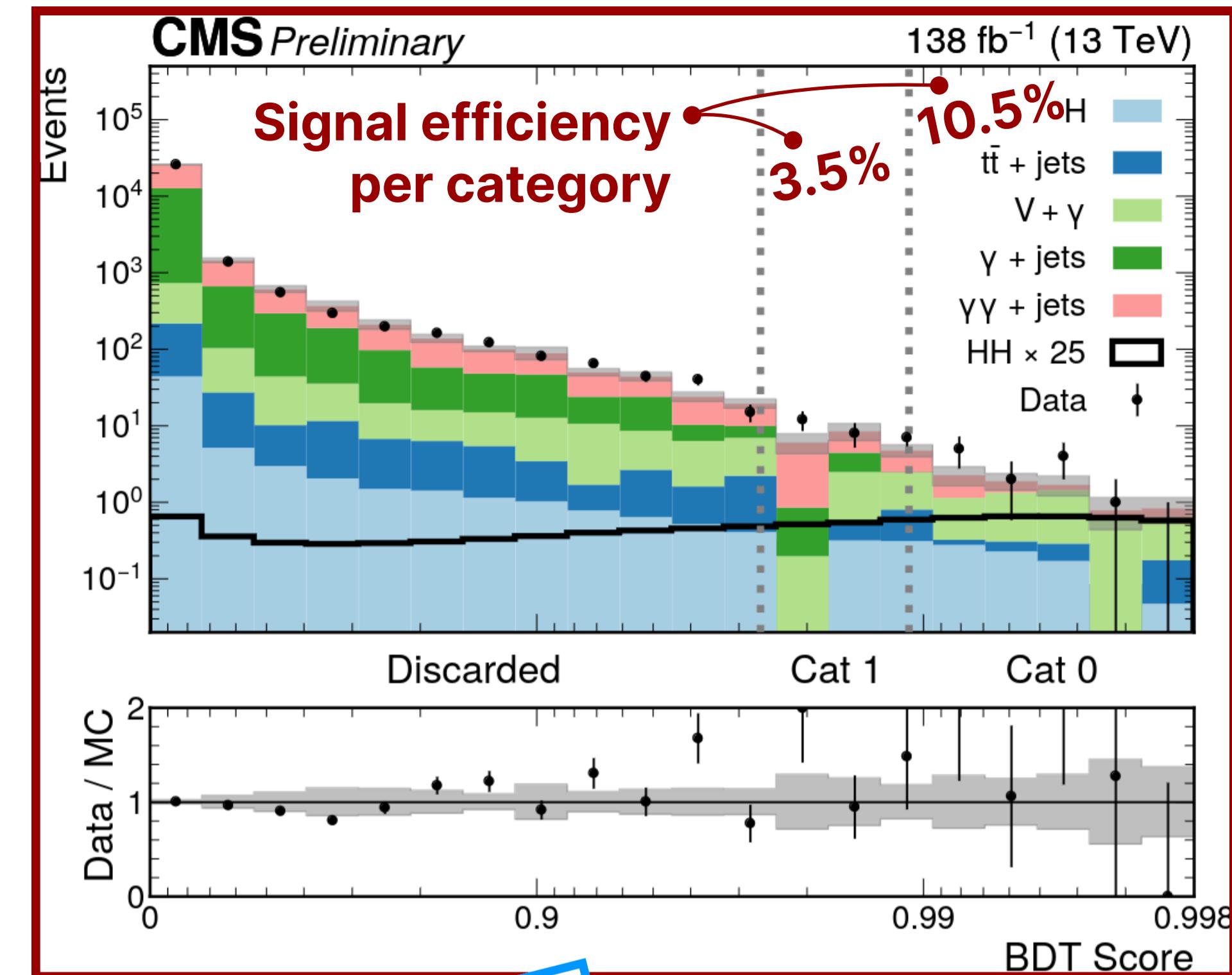
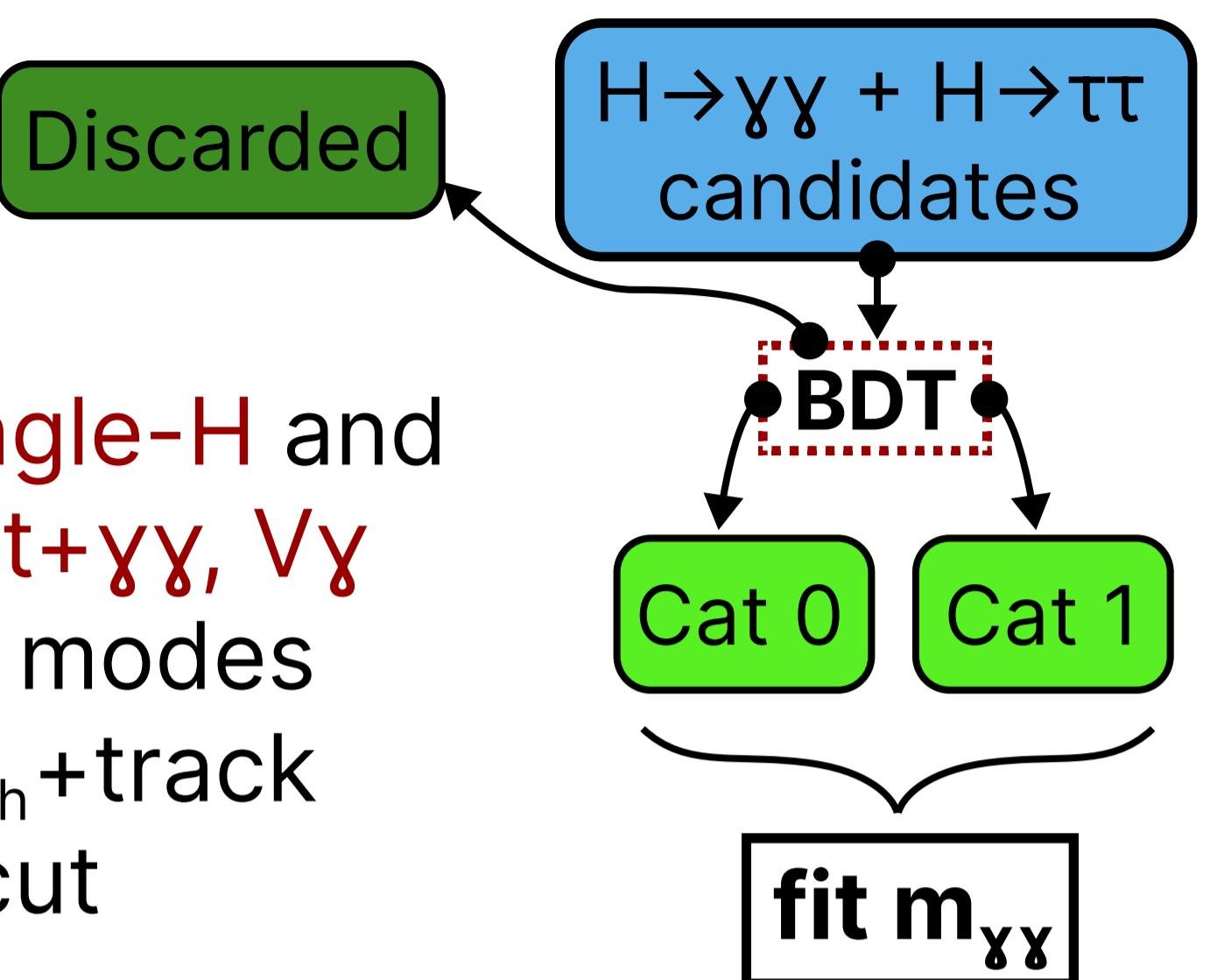
# Run2 Combination



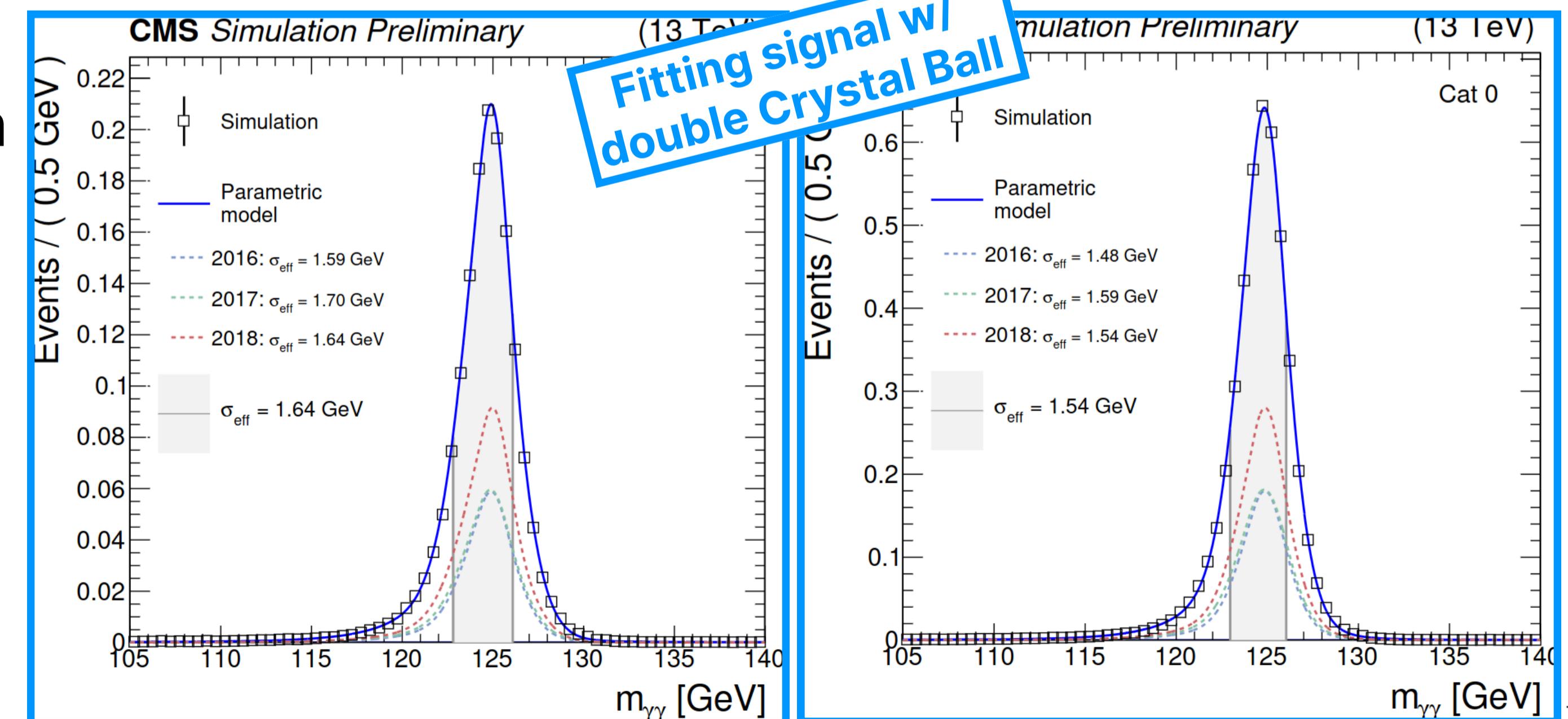
**$\text{HH} \rightarrow \gamma\gamma\tau\tau$**

**ggF**

- Dominant bkgs. are peaking single-H and **non-res.**  $\gamma + \text{jets}$ ,  $\gamma\gamma + \text{jets}$ ,  $t\bar{t} + \gamma$ ,  $t\bar{t} + \gamma\gamma$ ,  $V\gamma$
- $\tau$ 's reconstructed in all 6 decay modes
  - and consider single  $\tau_h$  and  $\tau_h + \text{track}$
  - apply DY veto with a mass cut
- A BDT uses kinematic features
  - enforces  $m_{\gamma\gamma}$  independence at 1<sup>st</sup> order



- Fit  $m_{\gamma\gamma}$  in signal-enriched categories
  - signal and  $H \rightarrow \gamma\gamma$  bkgd. from simulation
  - bkgd. continuum using the **discrete profiling method**
    - take the minimum likelihood of many analytic functions vs. signal strength: the envelope
    - penalize functions w/ more params



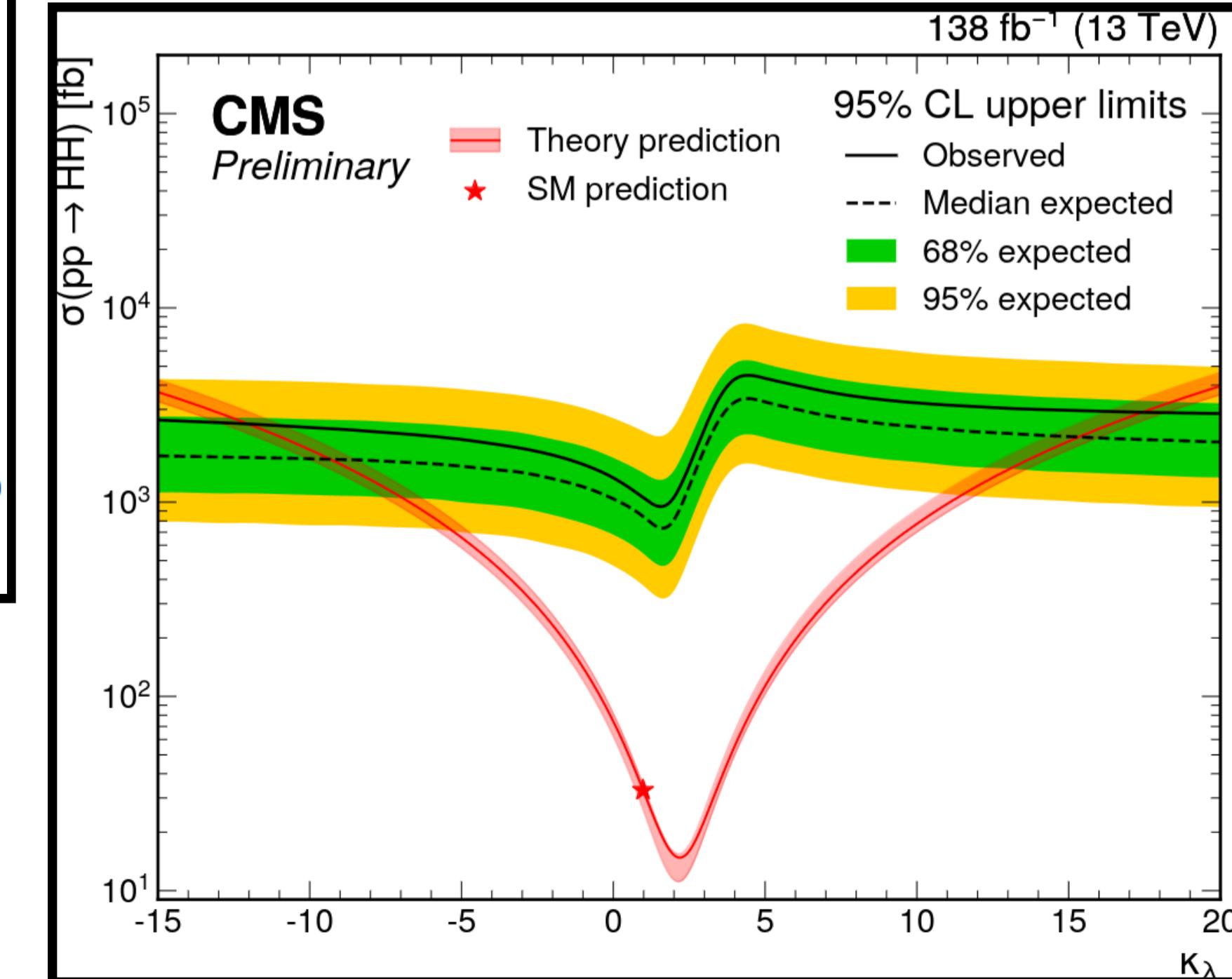
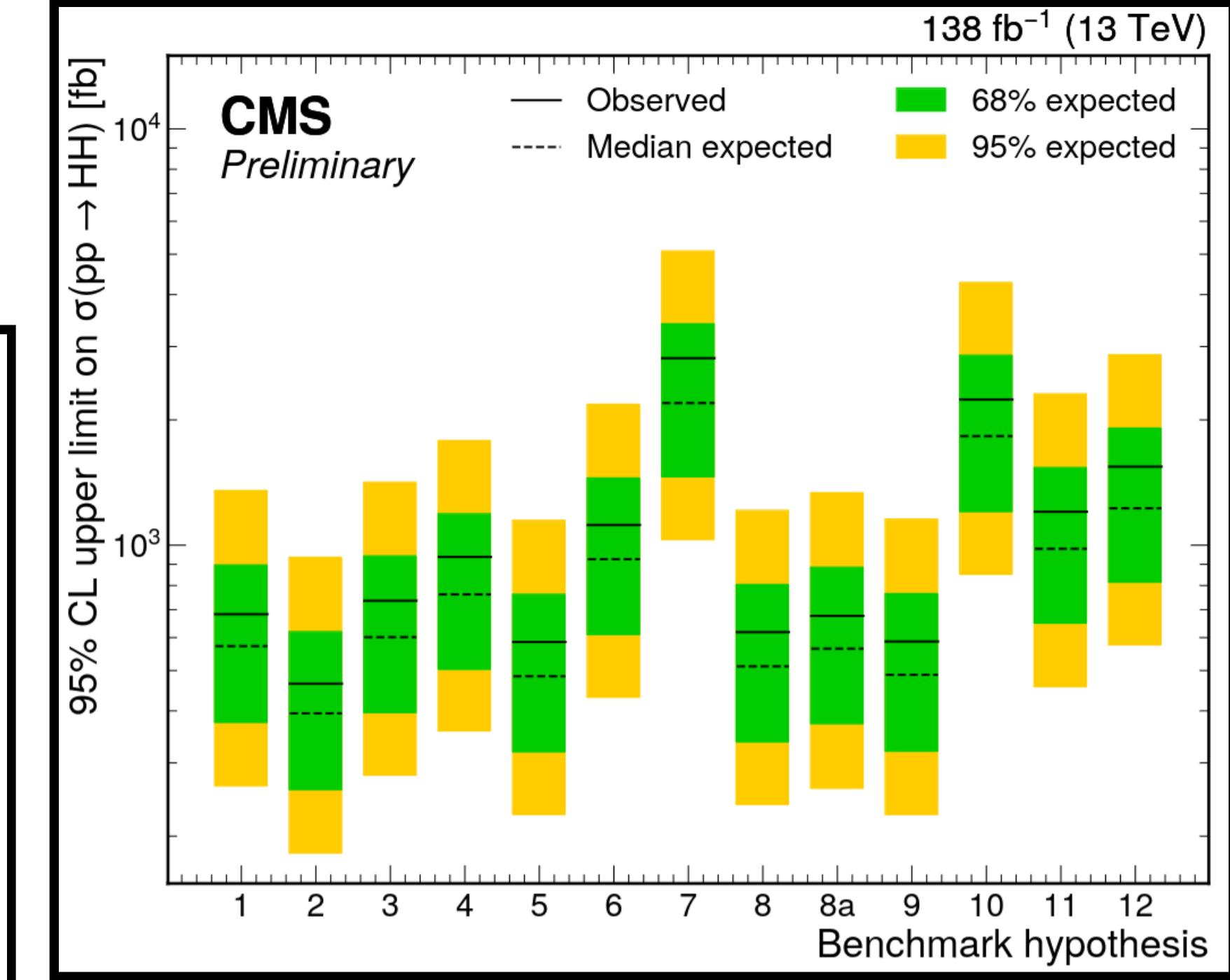
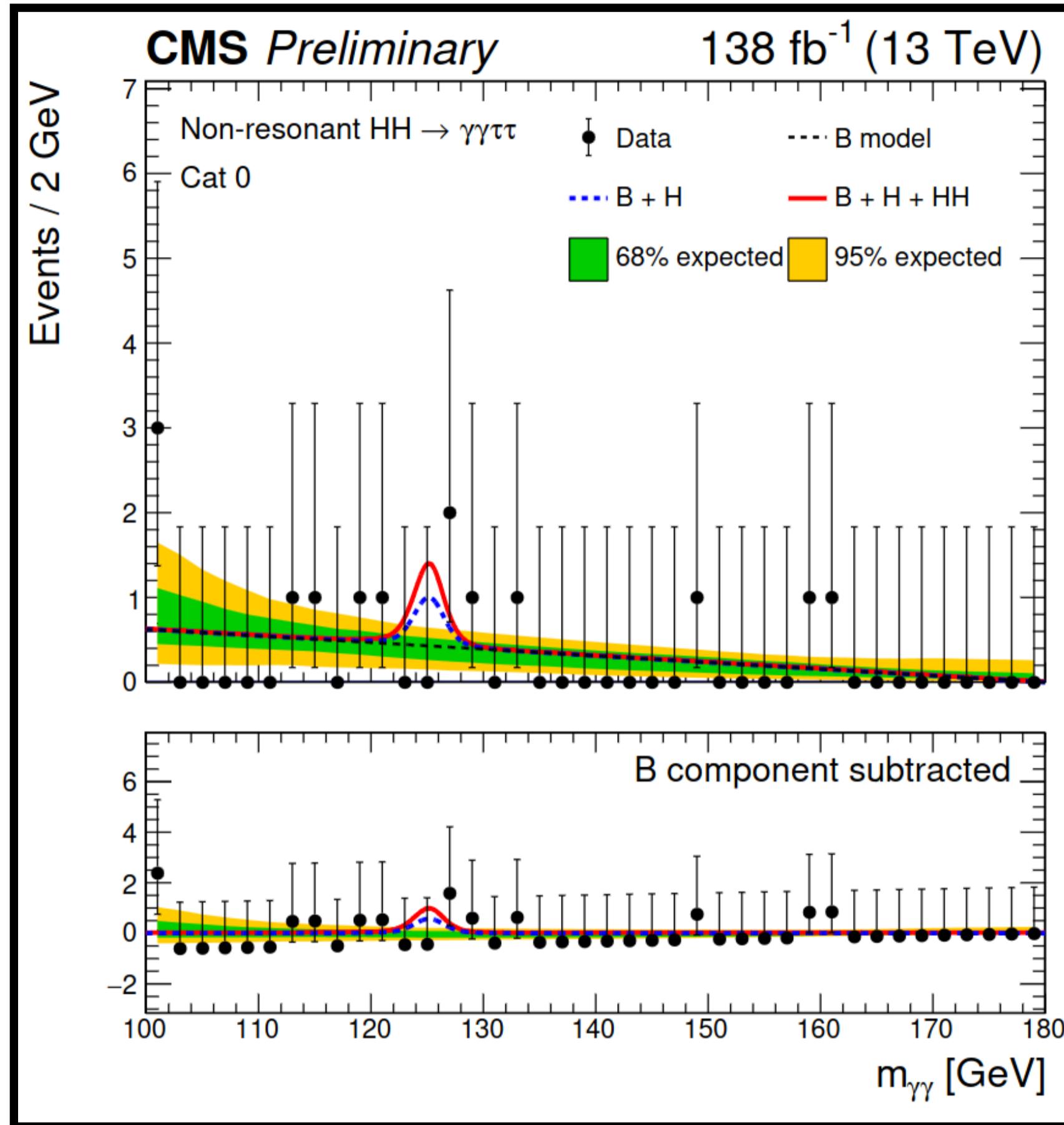
# $\text{HH} \rightarrow \gamma\gamma\tau\tau$

ggF

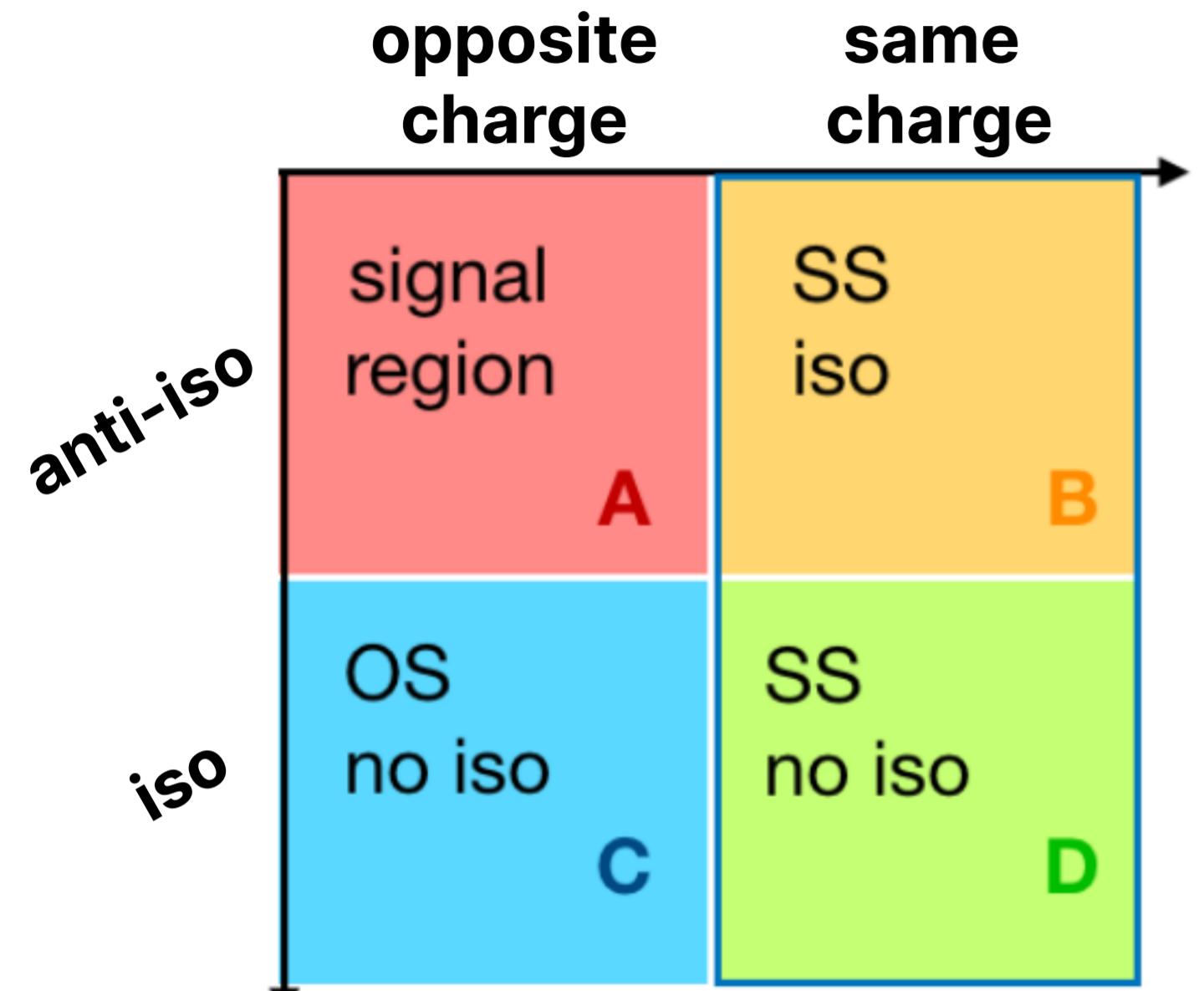
- Results expected to be added to a future HH combination
- The work also covers many resonant channels (not discussed here)

$$\sigma_{\text{HH}} < 33 \text{ (26)} \sigma_{\text{HH}}^{\text{SM}}$$

$$-13 \text{ (-11)} < k_\lambda < 18 \text{ (16)}$$



# Modelling the QCD background



shape      yield  
correction

$$A = \boxed{C} * \boxed{B/D}$$

$$A = \boxed{B} * \boxed{C/D}$$

$$N_{SR}^{4b} = \sum_{ibin} \left( \frac{N_{CR}^{4b}}{N_{CR}^{3b}} \right)_{ibin} \left( N_{SR}^{3b} \right)_{ibin}$$

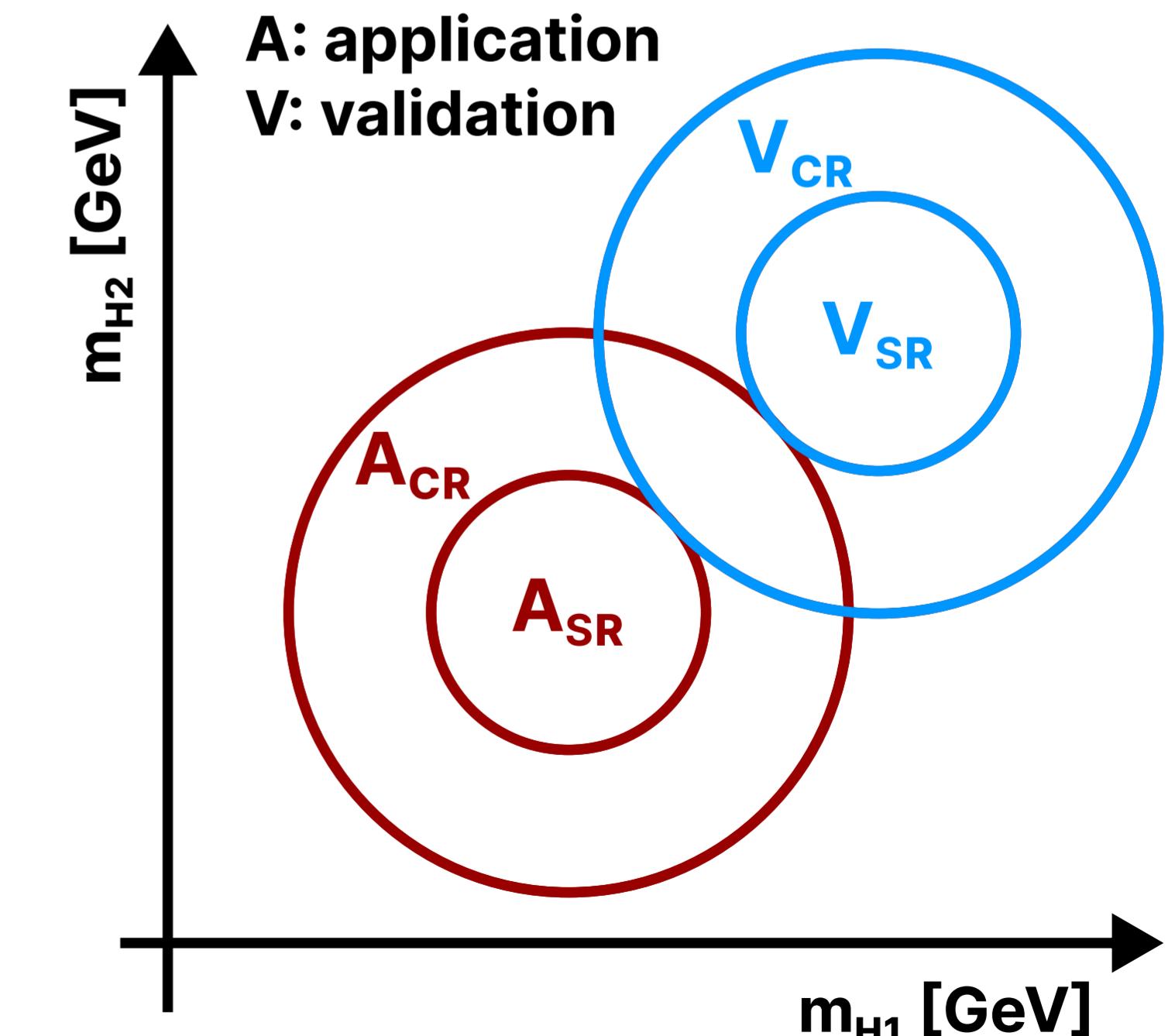
## $\text{HH} \rightarrow 4\text{b}$ (resolved)

- QCD data-driven estimation using 3/4 b's CRs
  - BDT reweighting fixes remaining diffs. in the 3b's SR
- ABCD-like method
  - Largest systematics come from the background model

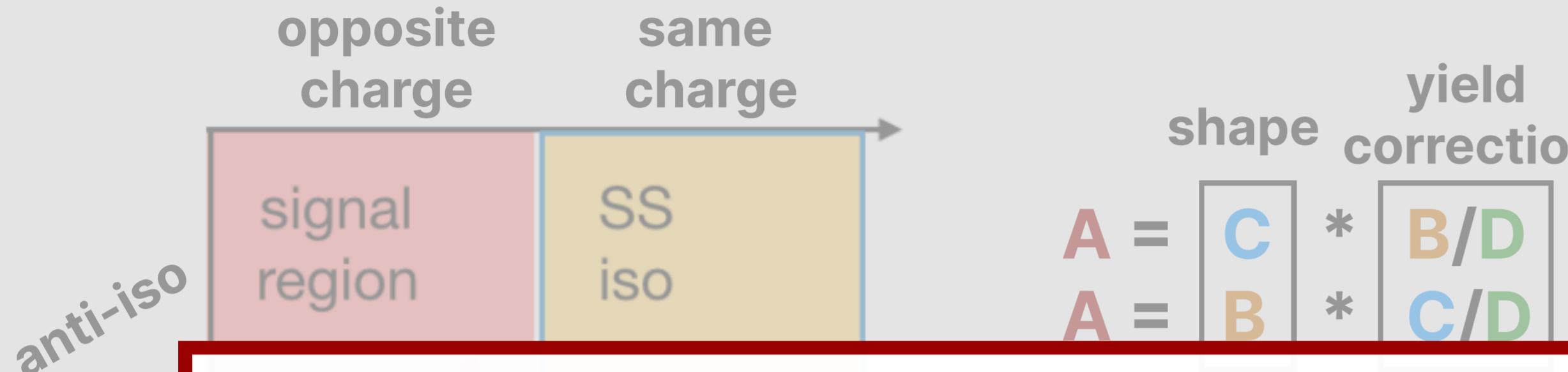
## $\text{HH} \rightarrow b\bar{b}\tau\tau$

Phys. Lett. B, 842, 137531

- QCD data-driven estimation using tau charge and isolation
- ABCD-like method
  - The shape uncertainty dominates (QCD dominates in the most sensitive  $\tau\tau$  channel)



# Modelling the QCD background



## $\text{HH} \rightarrow b\bar{b}\tau\bar{\tau}$

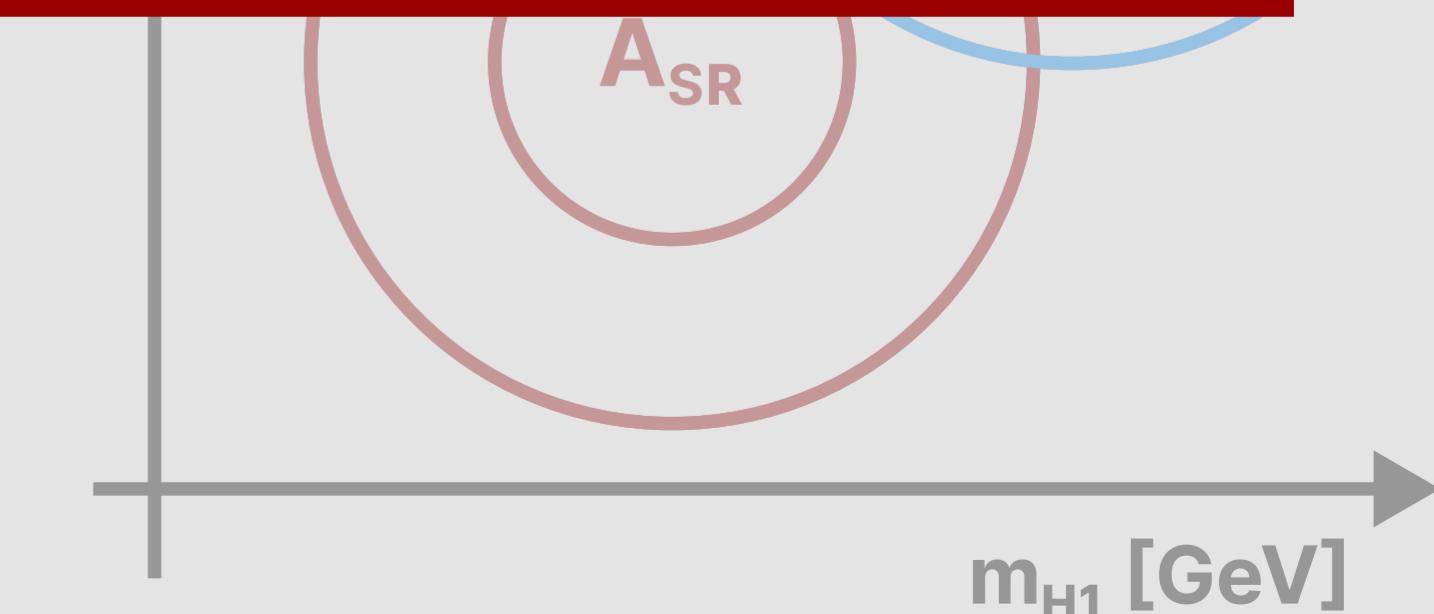
Phys. Lett. B, 842, 137531

- QCD data-driven estimation using tau charge and isolation
- ABCD-like method
  - The shape uncertainty dominates

Finite data in VRs imply an  
**“Inherent limitation on the capability to validate the performance of the background model”**

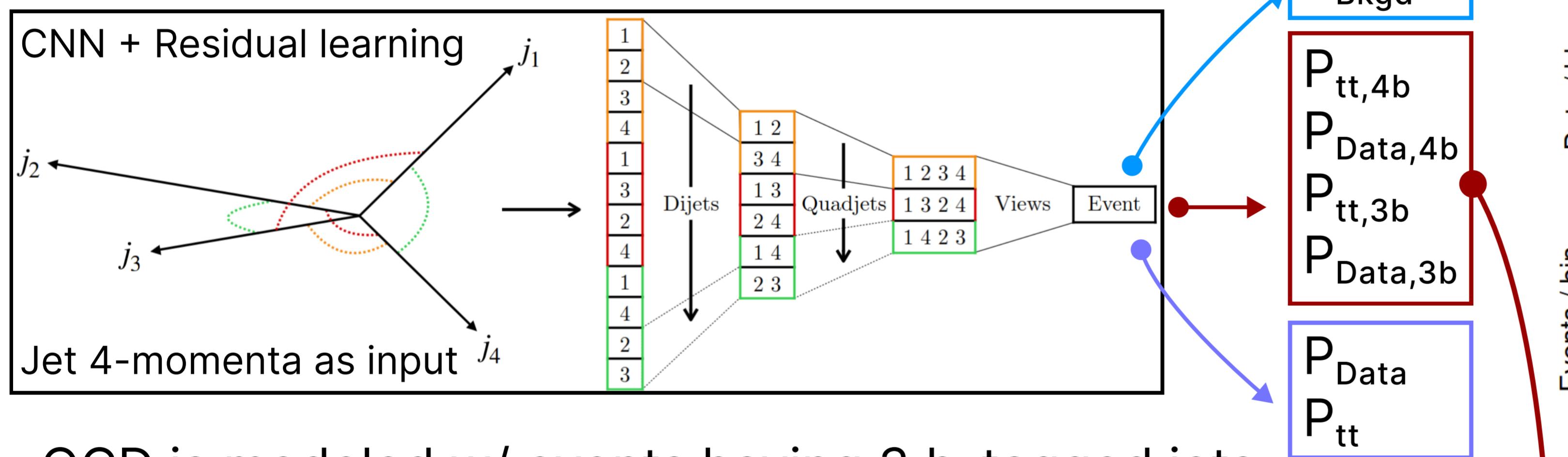
We would also like to  
**directly test the ABCD extrapolation in the SR**

- QCD data-driven estimation using 3/4 b's CRs
  - BDT reweighting fixes remaining diffs. in the 3b's SR
- ABCD-like method
  - Largest systematics come from the background model



# ZZ/ZH → 4b: Techniques for HH → 4b

- The same DNN multiclassifier architecture is used for:
  - signal vs bkgd. discrimination**
  - kinematic correction of the background model**
  - remove ttbar from synthetic dataset**

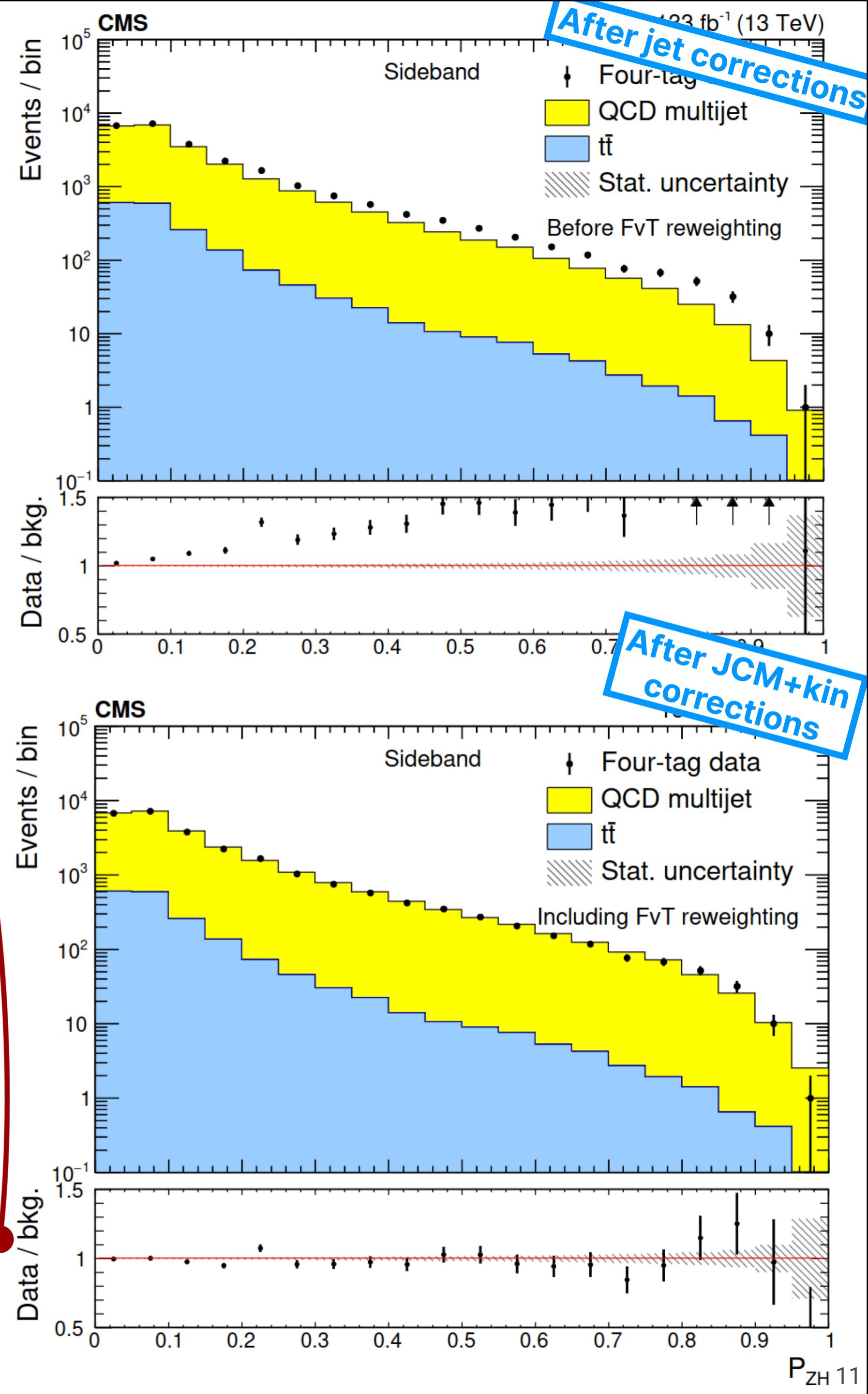


- QCD is modeled w/ events having 3 b-tagged jets
  - 17x more stats. (also loosening the b-tag requirement)
  - two sets of weights to describe the analysis 4b's bkgd.
    - account for additional jet activity**
    - correct kinematic differences**

$$w_{\text{JCM}} = \begin{cases} t \sum_{i=1}^n \binom{n}{i} f^i (1-f)^{n-i} (1+e/n^d) & (3+i) \text{ even} \\ t \sum_{i=1}^n \binom{n}{i} f^i (1-f)^{n-i} & (3+i) \text{ odd,} \end{cases}$$

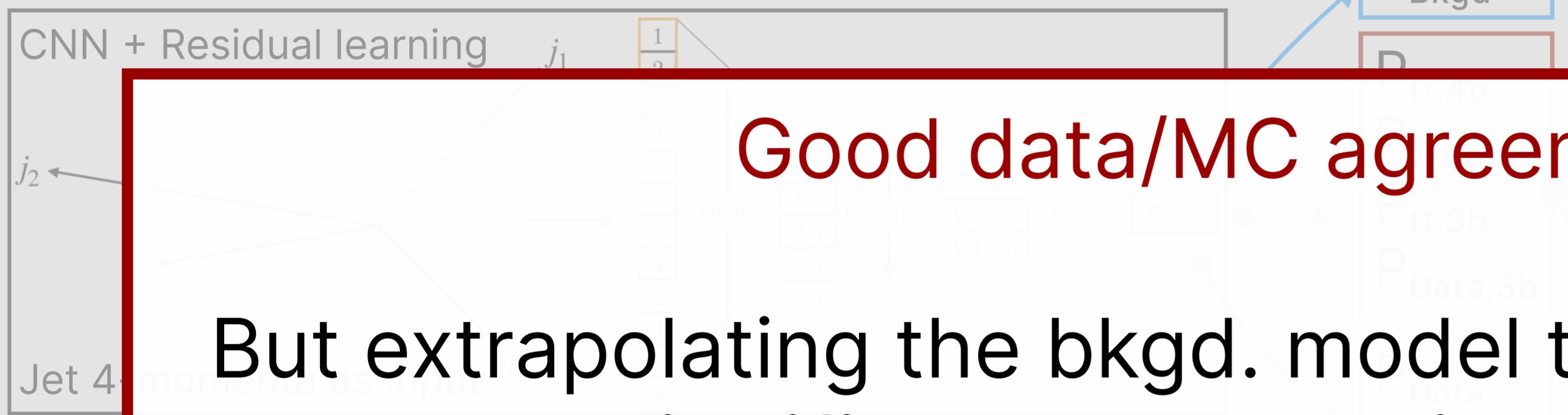
$$w_{\text{FvT}} = \frac{P(M_{4\text{b}})}{P(D_{3\text{b}})} \equiv \frac{P(D_{4\text{b}}) - P(\bar{t}t_{4\text{b}})}{P(D_{3\text{b}})}$$

see backup for explanation



# ZZ/ZH $\rightarrow 4b$ : Techniques for HH $\rightarrow 4b$

- The same DNN multiclassifier architecture is used for:
  - signal vs bkgd. discrimination
  - kinematic correction of the background model
  - remove ttbar from synthetic dataset



Good data/MC agreement.

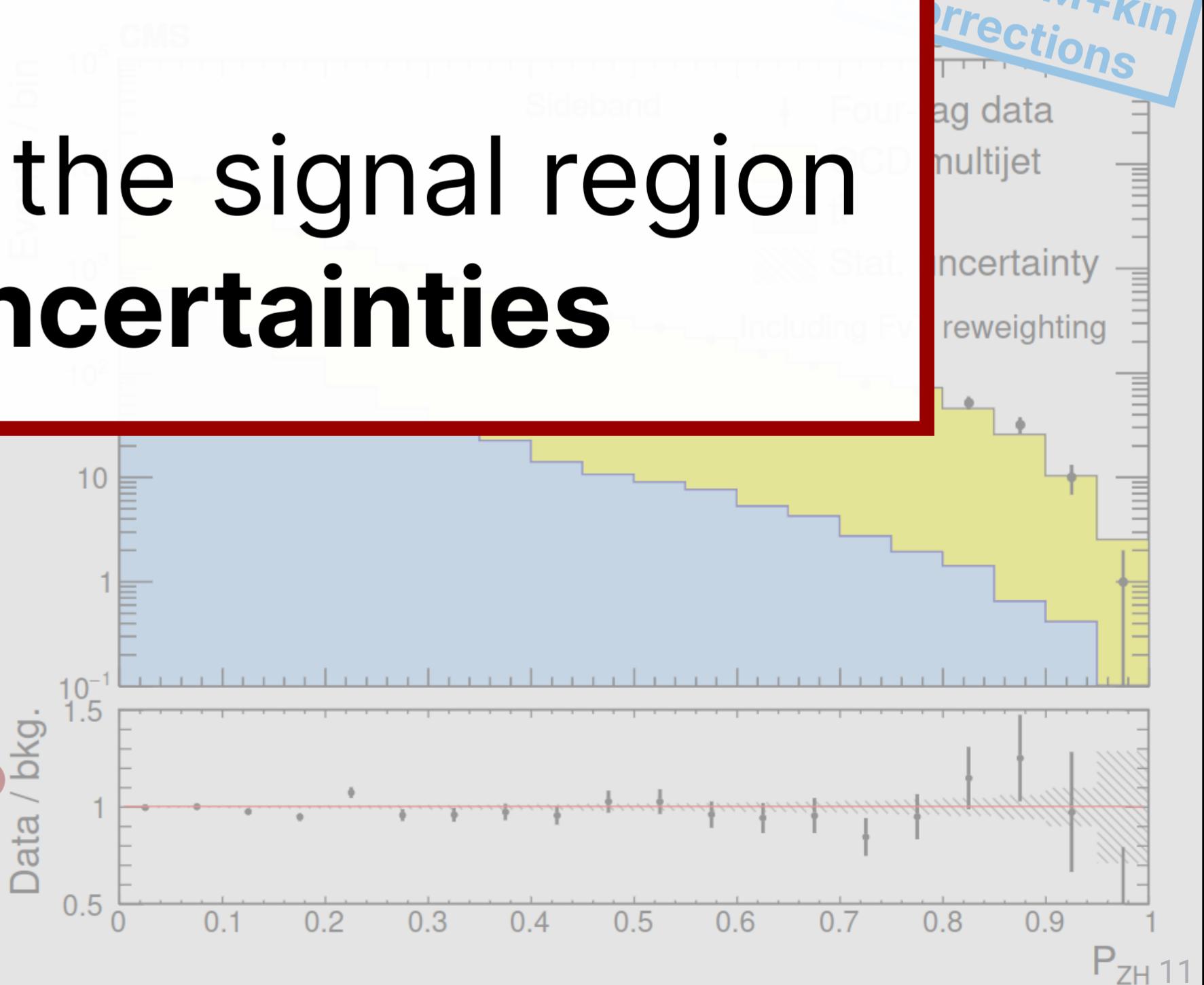
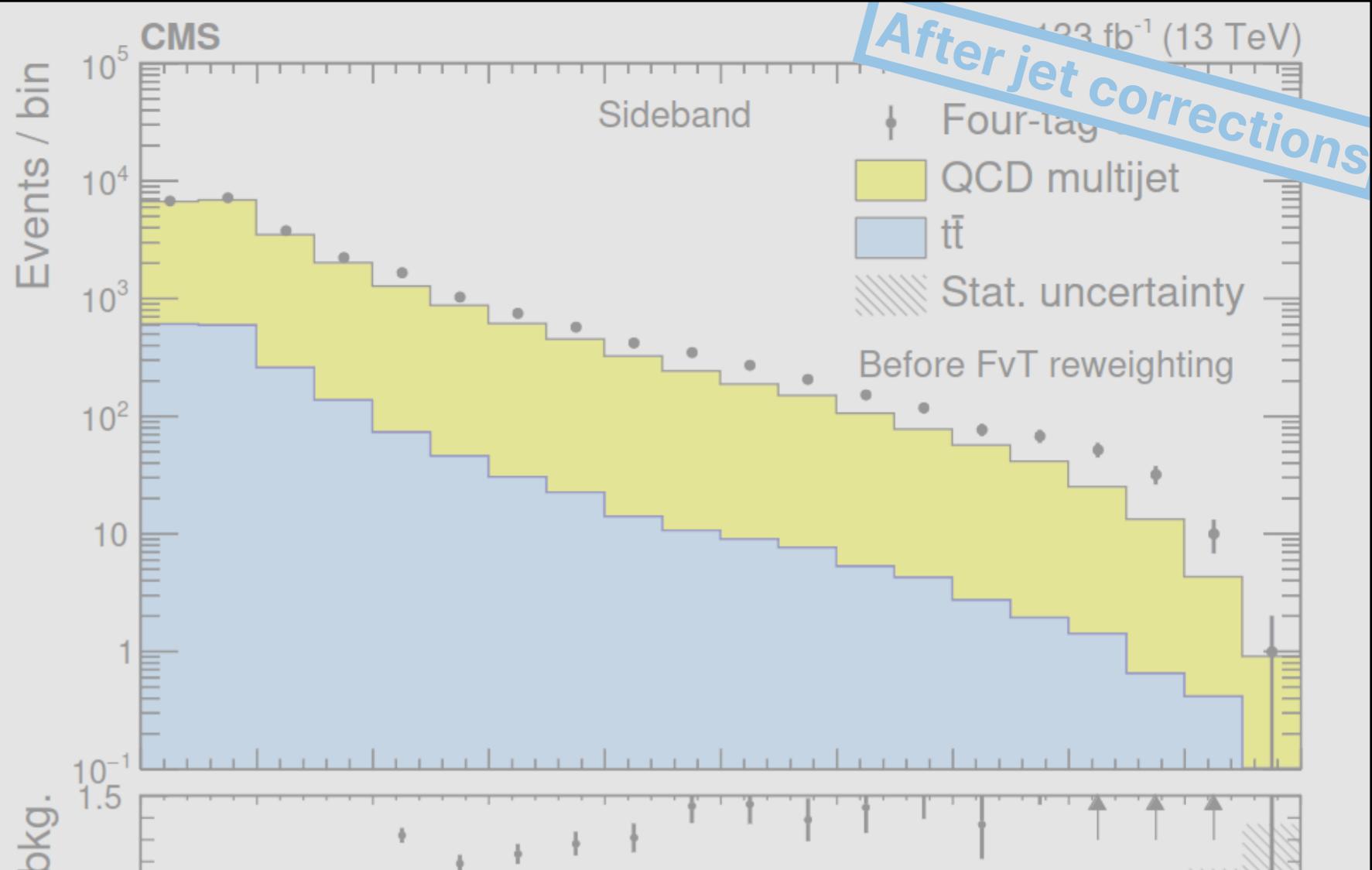
But extrapolating the bkgd. model to the signal region adds **significant systematics uncertainties**

- QC
  - 1/x more stats. (also loosening the b-tag requirement)
  - two sets of weights to describe the analysis 4b's bkgd.
    - i. account for additional jet activity
    - ii. correct kinematic differences

$$w_{\text{JCM}} = \begin{cases} t \sum_{i=1}^n \binom{n}{i} f^i (1-f)^{n-i} (1+e/n^d) & (3+i) \text{ even} \\ t \sum_{i=1}^n \binom{n}{i} f^i (1-f)^{n-i} & (3+i) \text{ odd,} \end{cases}$$

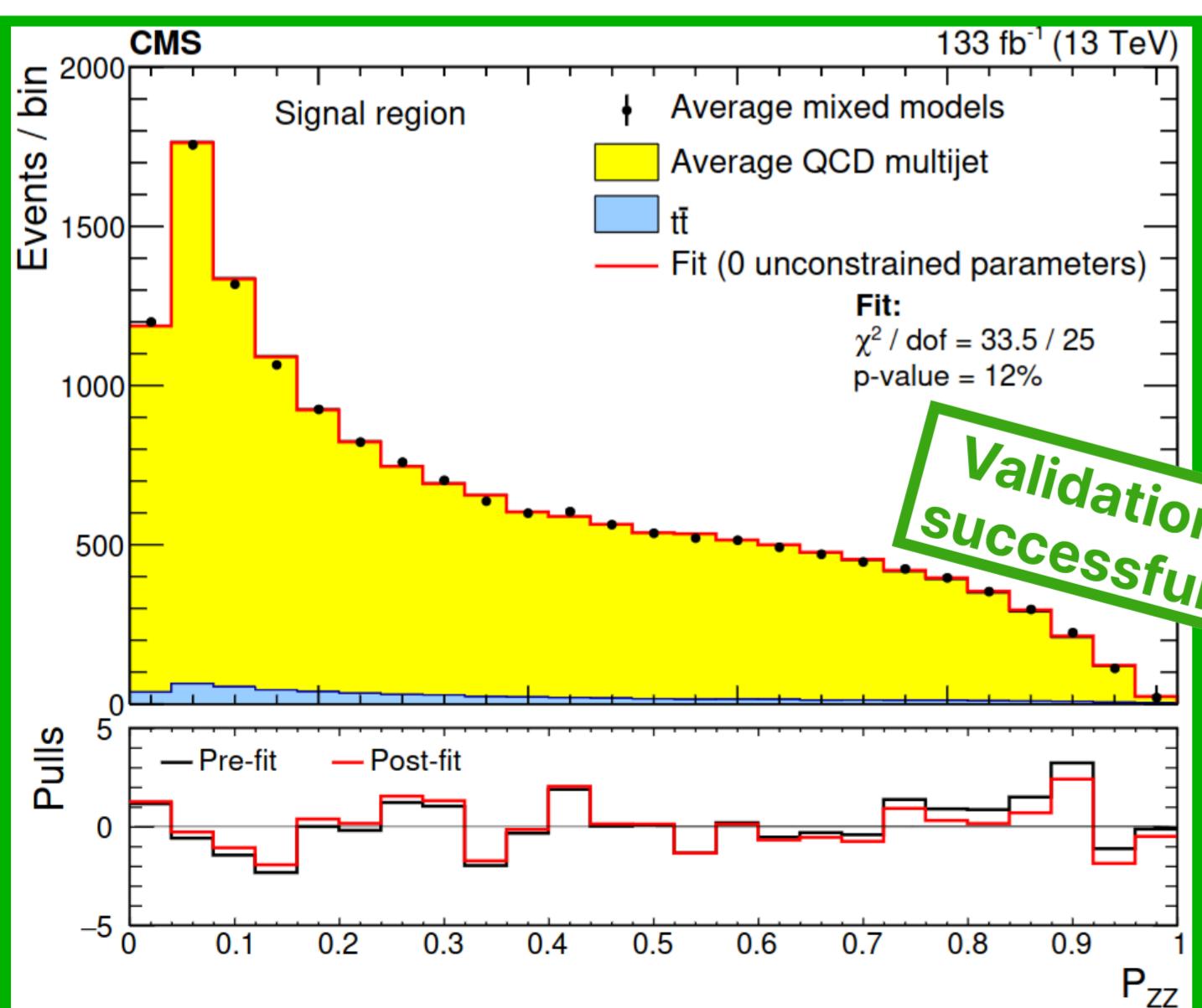
$$w_{\text{FvT}} = \frac{P(M_{4b})}{P(D_{3b})} \equiv \frac{P(D_{4b}) - P(t\bar{t}_{4b})}{P(D_{3b})}$$

see backup for explanation

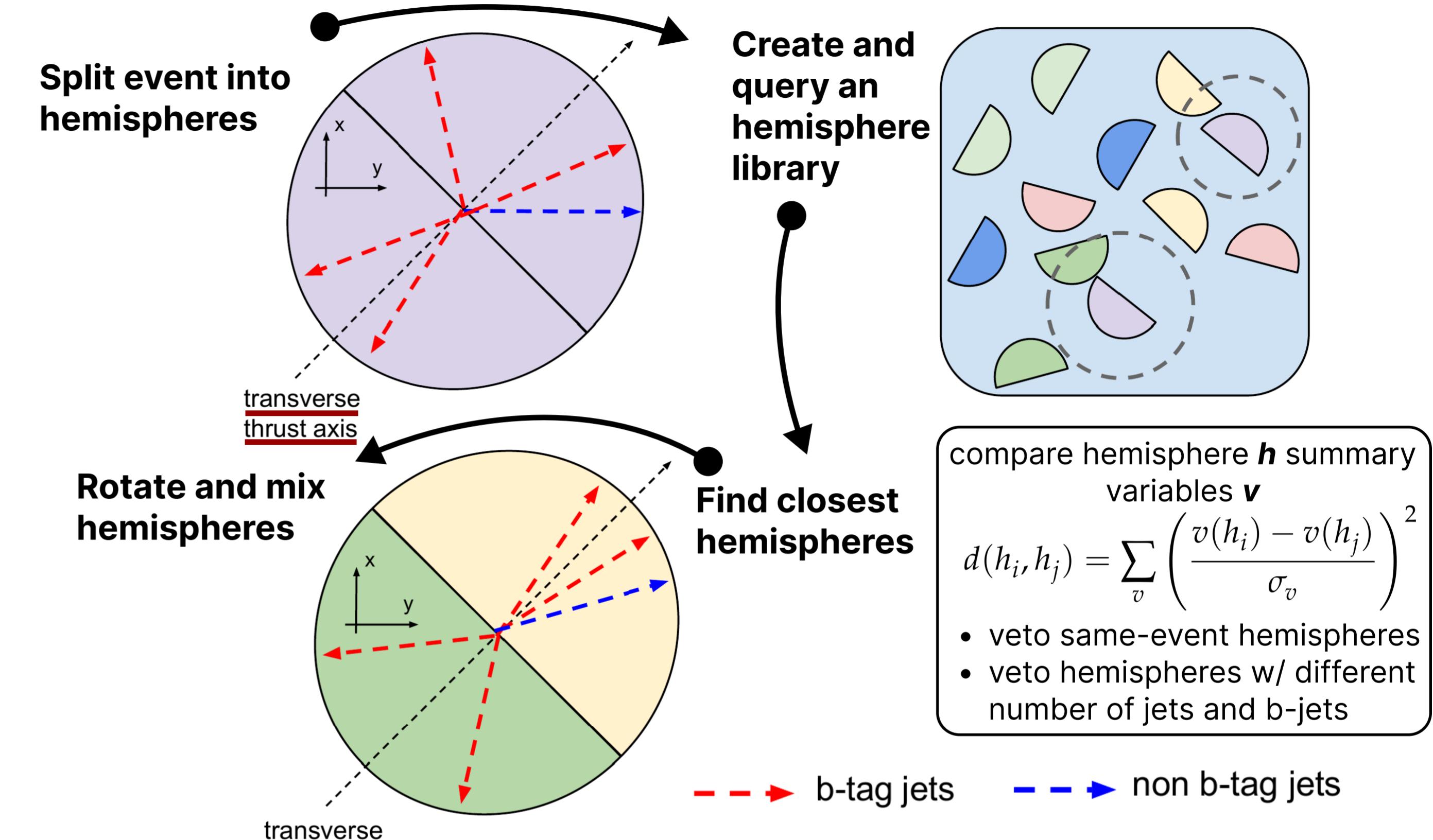


# Hemisphere Mixing

- Generate instead **synthetic data**
  - use it for bkgd. validation
  - technique introduced in an older CMS  $\text{HH} \rightarrow 4\text{b}$  analysis
- **HM improvements (new!)**
  - a. use 3-tagged data in the mixing: more stats, less signal contam.
  - b. Avoid mixing QCD hemispheres with  $t\bar{t}$  hemispheres



**Thrust axis:** the axis where the sum of the absolute values of the projections of the  $p_T$  of the jets is maximal



compare hemisphere  $h$  summary variables  $v$

$$d(h_i, h_j) = \sum_v \left( \frac{v(h_i) - v(h_j)}{\sigma_v} \right)^2$$

- veto same-event hemispheres
- veto hemispheres w/ different number of jets and b-jets

## Systematics

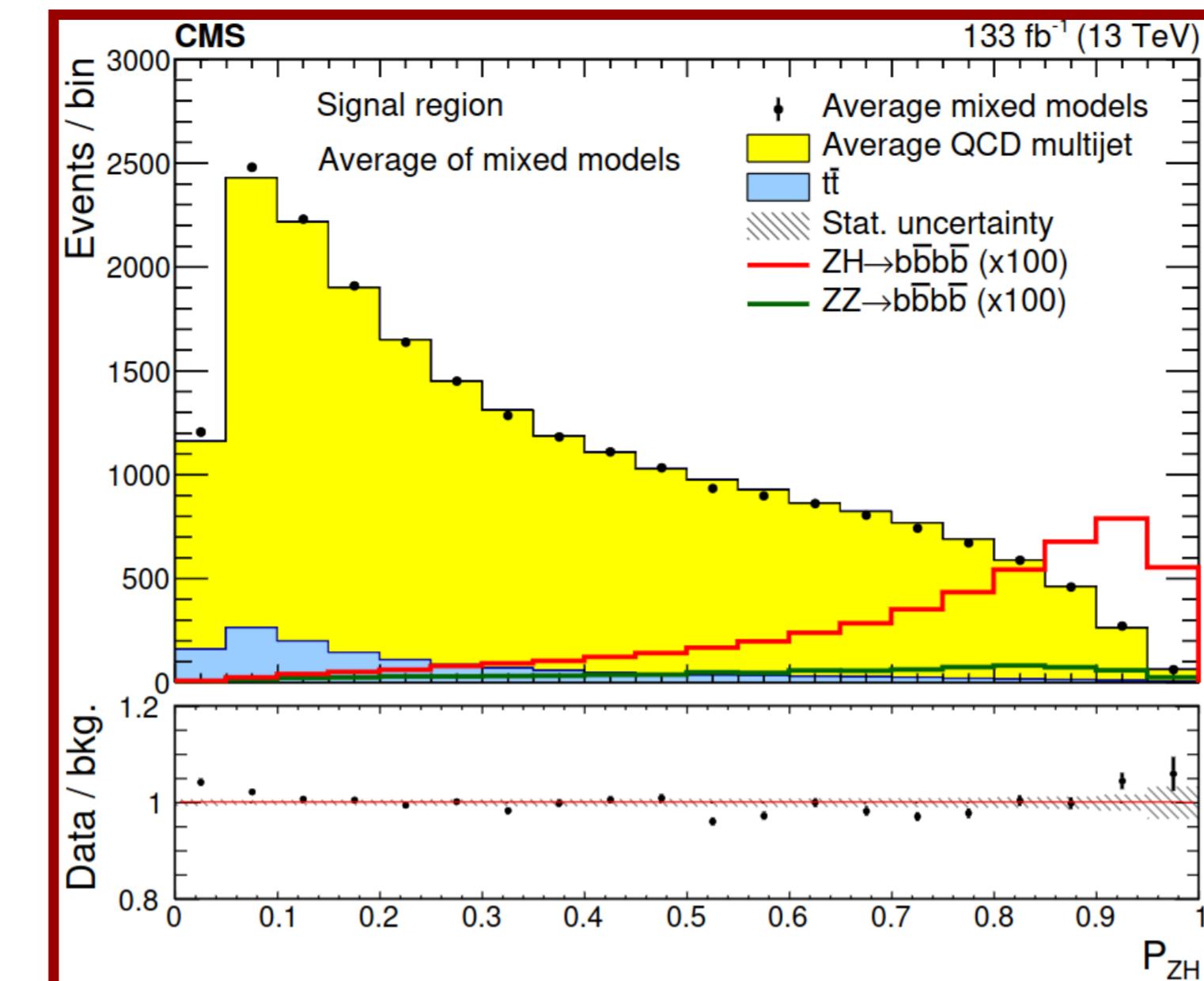
1. diffs. btw. **mixed models** and their averages
2. diffs. btw. bkgd. model and mixed models
3. check fit robustness against the addition of an unconstrained signal template

each mixed model is a subsample of the mixed data to match the stats of the 4-tagged SR

Not suffering from low stats!

# ZZ/ZH $\rightarrow$ 4b: Results and Prospects

- Fit validated using a mixed model replacing the 4-tag dataset
- Combined fit in ZZ and ZH
  - Similar sensitivities despite different xsecs
  - ZH has more efficiency and less background



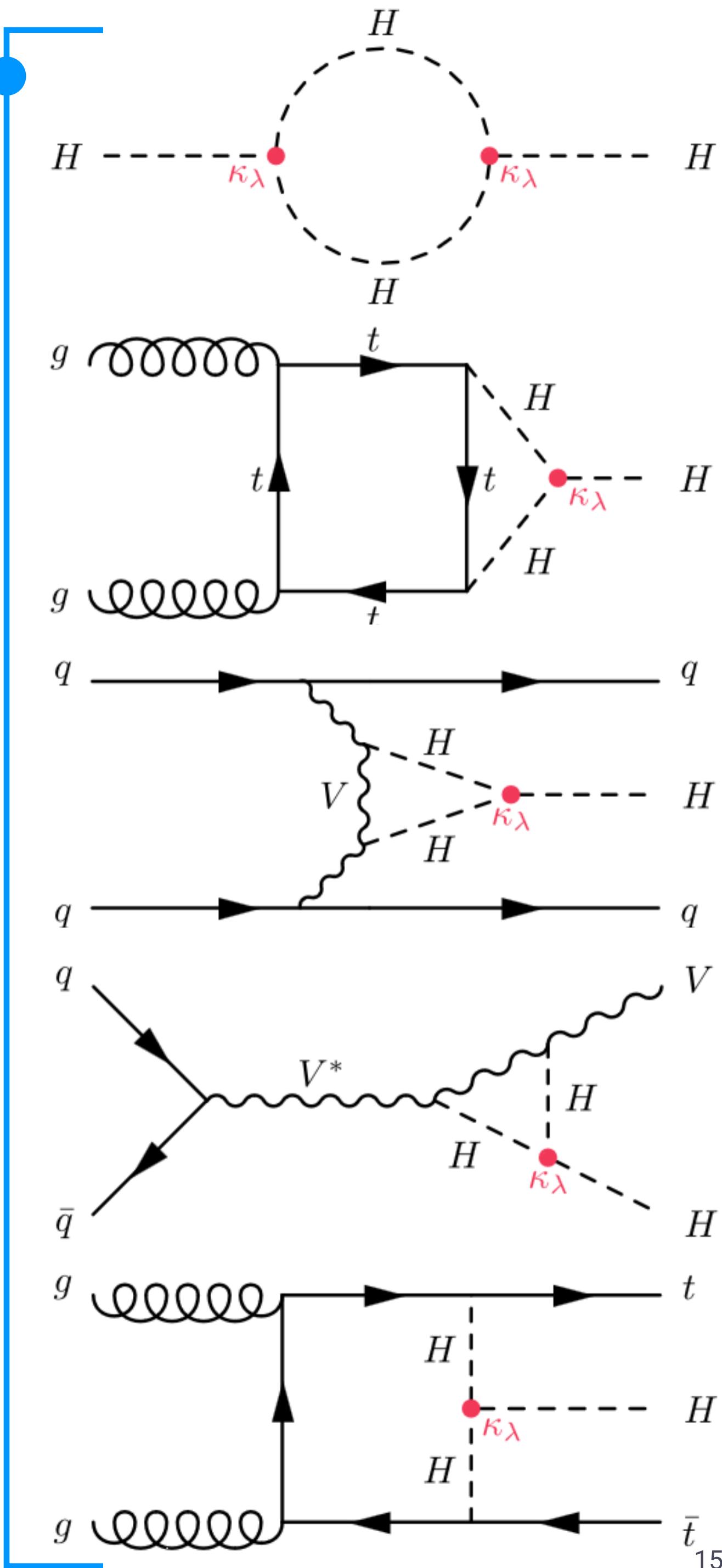
	ZZ	ZH
$\mu(\text{exp.})$	$1^{+1.9}_{-1.7} (1^{+1.4}_{-1.3})$	$1^{+1.5}_{-1.4} (1^{+1.1}_{-1.1})$
$\mu(\text{obs.})$	$0.0^{+2.0}_{-1.7}$	$2.2^{+0.9}_{-0.8}$
Limit(exp.)	3.8 (2.8)	2.9 (2.3)
Limit(obs.)	3.8	5.0

**ZH will likely be observed first!**

**Importantly:** We now have a principled and precise way of measuring the most important systematics directly in the SR.

# Combination H+HH

- We can exploit **NLO corrections** to single-H which depend on  $k_\lambda$ 
  - largest sensitivity is present in VH and ttH processes (up to 10%)
- **Complementarity**
  - Single-H provides stronger constraints on H couplings to fermions and vector bosons
  - HH is more sensitive to  $\mathbf{k}_\lambda$
- Main challenge: **estimate and efficiently remove overlaps** between signal region of different analysis
  - additional selections are applied and/or
  - the least sensitive category/analysis is removed
  - example:  $\text{HH} \rightarrow \text{bbZZ}$  is removed in favour of  $\text{H} \rightarrow \text{ZZ} \rightarrow 4l$
- The modelling of systematics in HH processes is generally simpler due to their limited statistics



# Combination H+HH

Setting couplings to SM

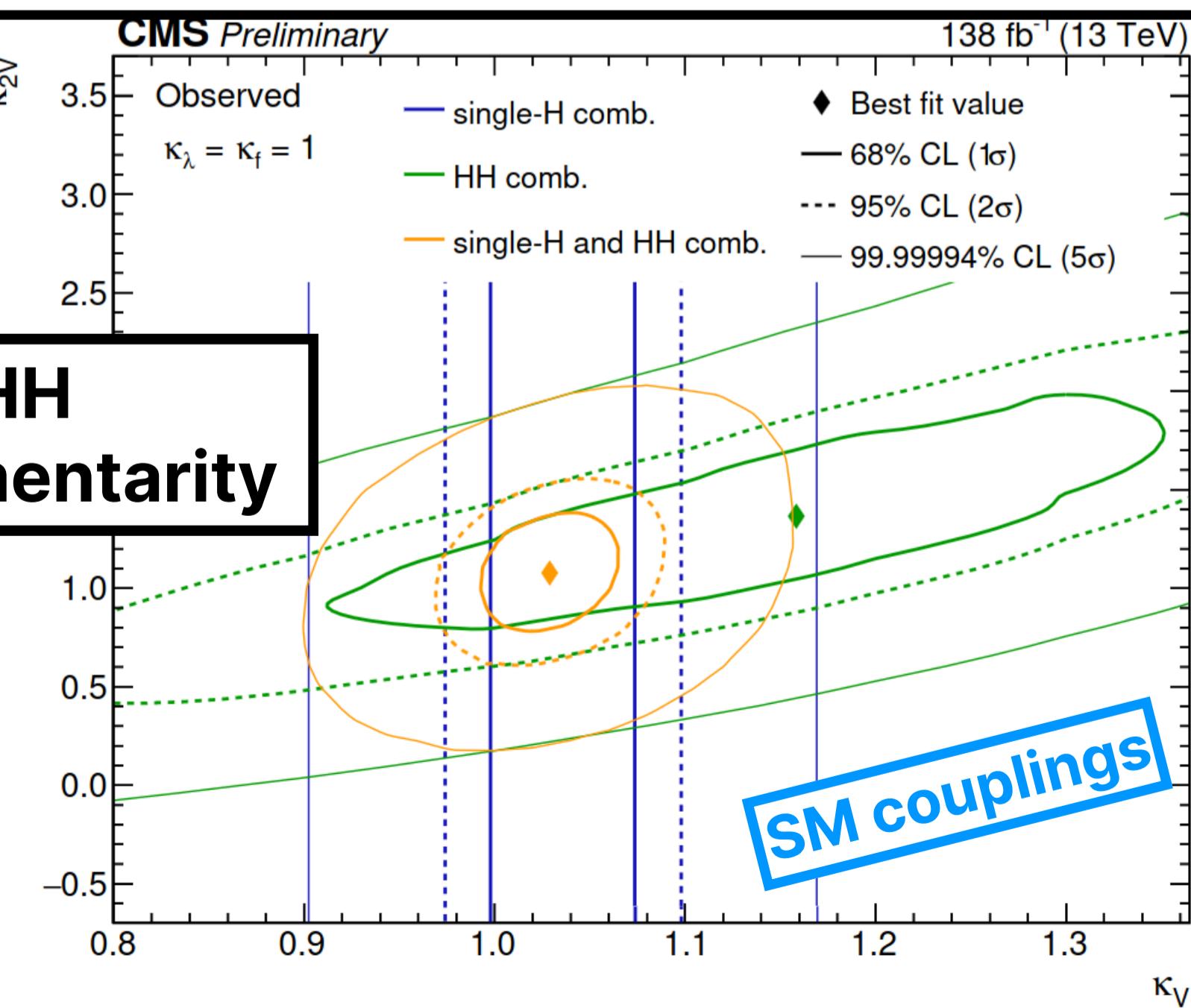
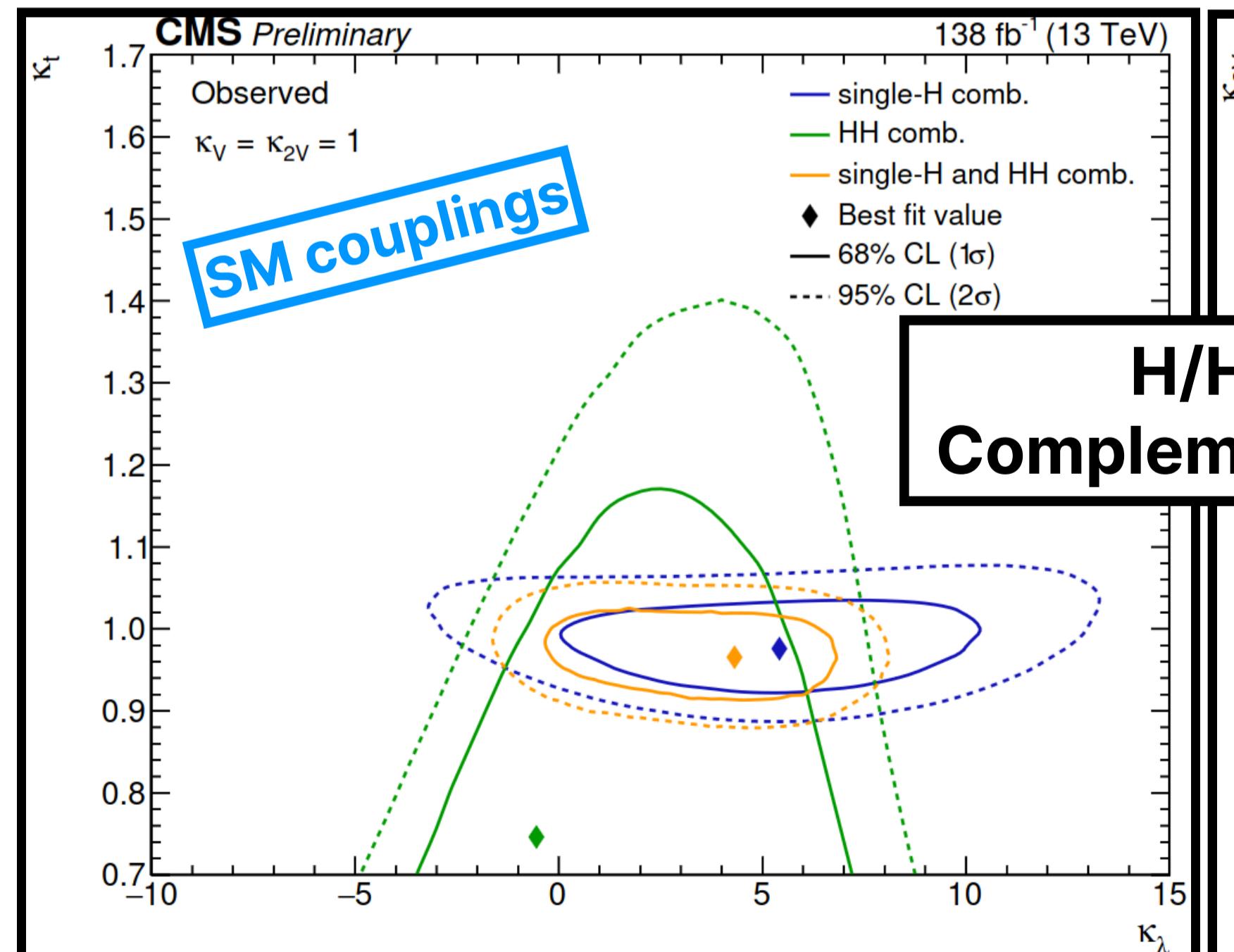
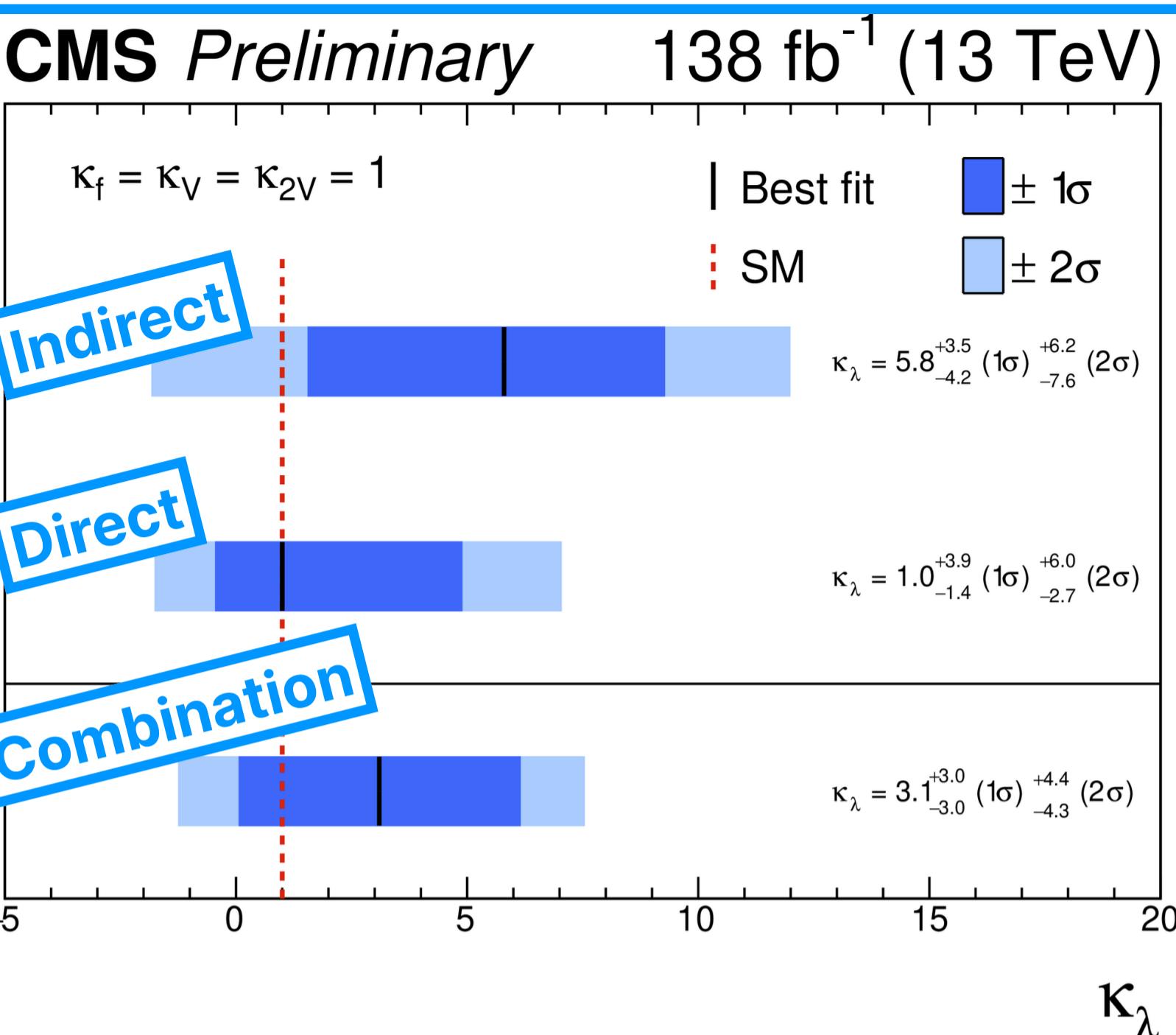
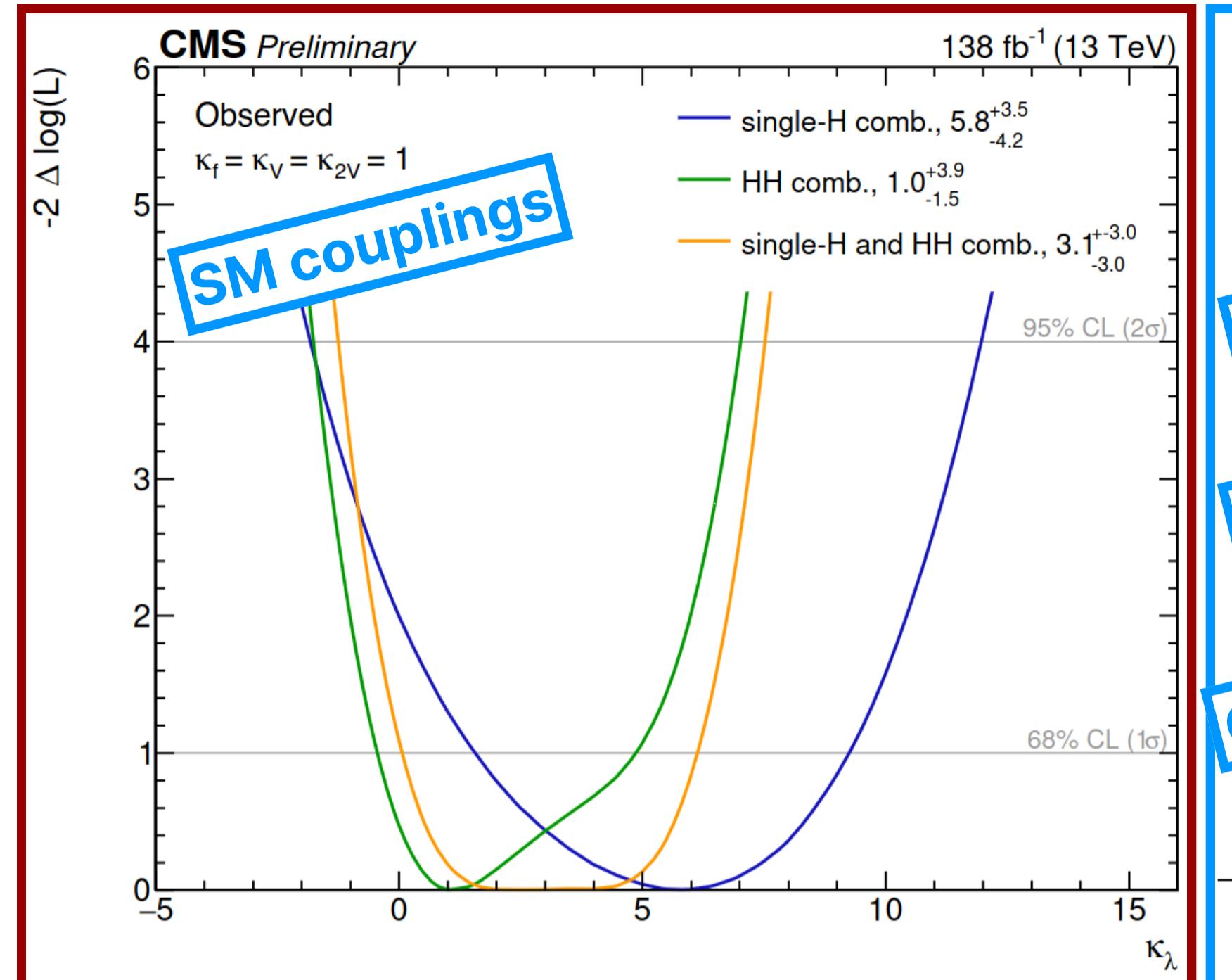
- $-1.2(-2.0) < \kappa_\lambda < 7.5 (7.7)$

Let couplings free in the fit

- $-1.4(-2.3) < \kappa_\lambda < 7.8 (7.8)$

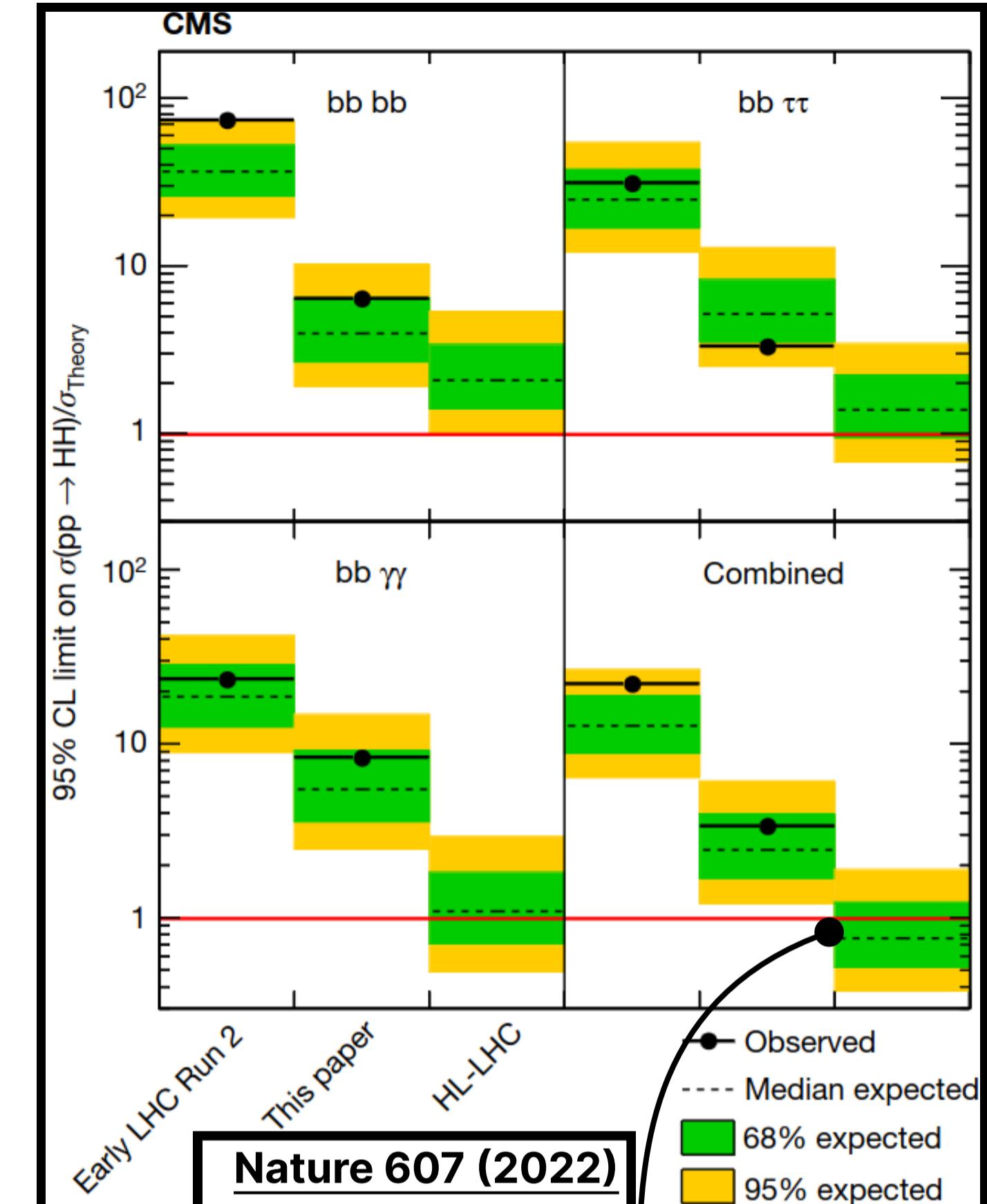
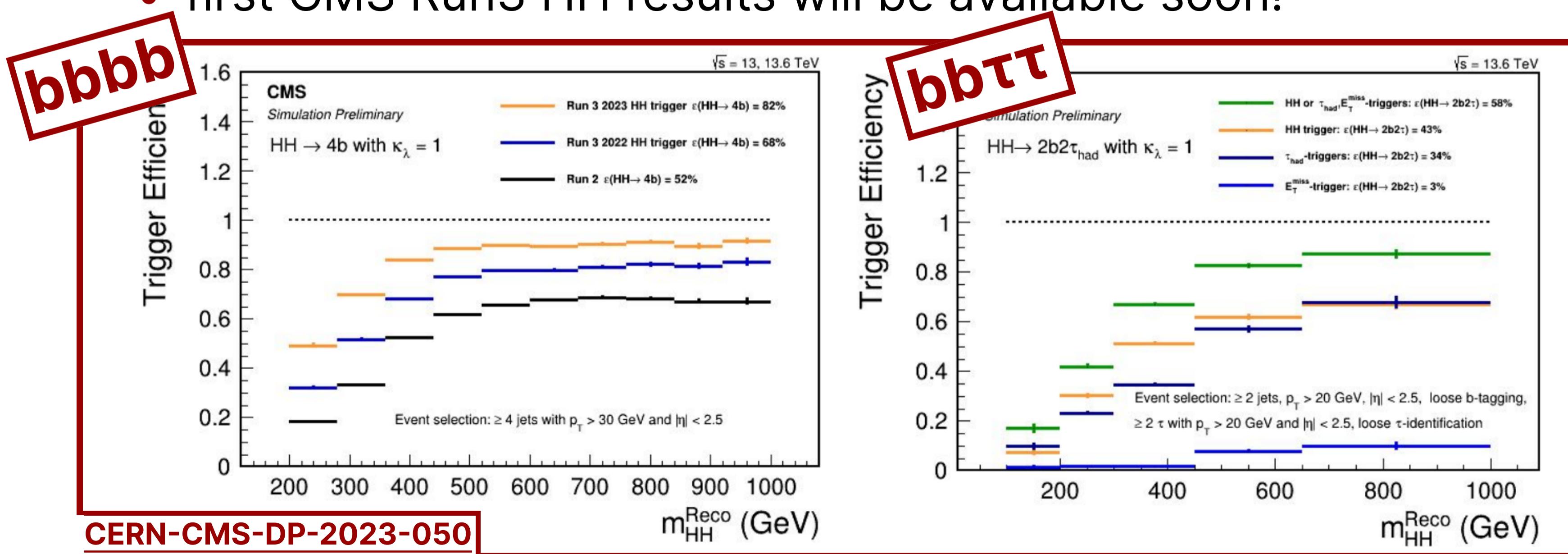
- Similar sensitivity to  $\kappa_\lambda$  as in the ATLAS H+HH comb.

- $\kappa_{2V}=0$  excluded at  $>5\sigma$



# Run 3 and beyond

- $k_\lambda$  and **EFT** will be further constrained in the near future
  - **new HH decay** channels are being explored
  - stats are still a limiting factor
    - but ggF theory uncert. may become important in the future
  - we are **close to SM HH sensitivity and  $k_{2v}=0$  was excluded**
- Run 3 is an opportunity for improvement before the HL-LHC
  - **improved trigger strategy** will boost HH searches
  - **improved taggers**: transformers, PNet for  $\tau$ -jets, ...
  - several analysis might benefit from synthetic datasets
  - first CMS Run3 HH results will be available soon!

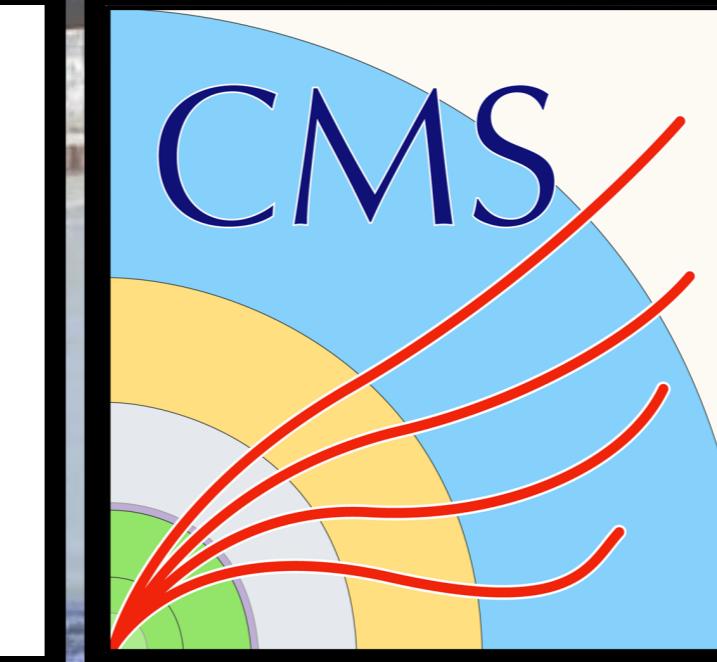
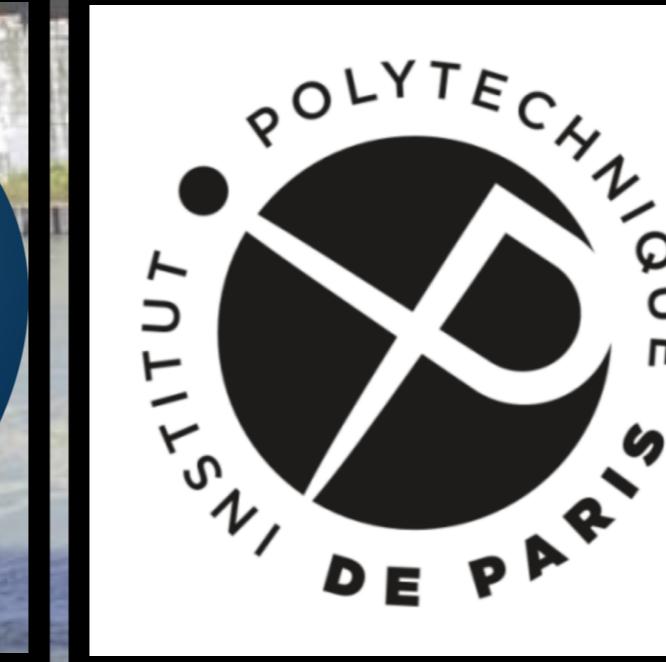


We are getting close...

PNet b-tagging at trigger level

Data parking allows lower HT thresholds

# Backup



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# ZZ/ZH $\rightarrow$ 4b: Weight for additional jet activity

- Requiring 3 b-tagged jets biases the SR (which has 4)
- A “Jet Combinatorial Model” weight is introduced to improve the background description of the four-tag sample
  - introduce “anti-b-tagged” jets: jets not passing the looser b-tagging requirement
  - considers **all combinations** of anti-b-tagged jets where at least one is treated like a real b-jet (so 3+1 b-jets, or 3+2, or 3+3, or 3+...)

$$w_{\text{JCM}} = \begin{cases} t \sum_{i=1}^n \binom{n}{i} f^i (1-f)^{n-i} & \text{if } \text{even} \\ t \sum_{i=1}^n \binom{n}{i} f^i (1-f)^{n-i} \frac{(3+i)}{(3+i)} & \text{if } \text{odd}, \end{cases}$$

number of anti-b-tagged jets

“pair-enhancement” factor

even odd,

normalization to account for the looser b-tag requirement of the 3-tag sample

constant per-jet “transfer factor”

dilution of “pair-enhancement” with increasing jet multiplicity

number of b-tagged jets in the event

The parameters  $t$ ,  $f$ ,  $e$  and  $d$  are obtained with a fit to the distributions of jet and b-jet multiplicities

# Motivation: EFT

- Given the absence of clear NP signals: EFTs!
- Look for effects from an unknown high-energy theory in a model independent way

Taylor expanding the SM in  $(E, vev)/\Lambda$ :

$$\mathcal{L}_{\text{SMEFT}} = \mathcal{L}_{\text{SM}} + \sum_i \frac{c_i^{(5)}}{\Lambda_i} \mathcal{O}_i^{(5)} + \sum_i \frac{c_i^{(6)}}{\Lambda_i^2} \mathcal{O}_i^{(6)} + \sum_i \frac{c_i^{(7)}}{\Lambda_i^3} \mathcal{O}_i^{(7)} + \sum_i \frac{c_i^{(8)}}{\Lambda_i^4} \mathcal{O}_i^{(8)} + \dots$$

19 parameters

1 operator type  
(Weinberg operator)

Majorana  $\nu$  masses  
( $m_\nu$  small  $\rightarrow \Lambda_i$  high)

violate L,  
some violate B,  
high suppression

2499 parameters with  $\Delta L = \Delta B = 0$   
+ O(300) with  $\Delta L = \Delta B = 1$   $\longrightarrow$  proton decay  $\rightarrow \Lambda_i$  high

76 assuming  $U(3)^5$  flavor symmetry

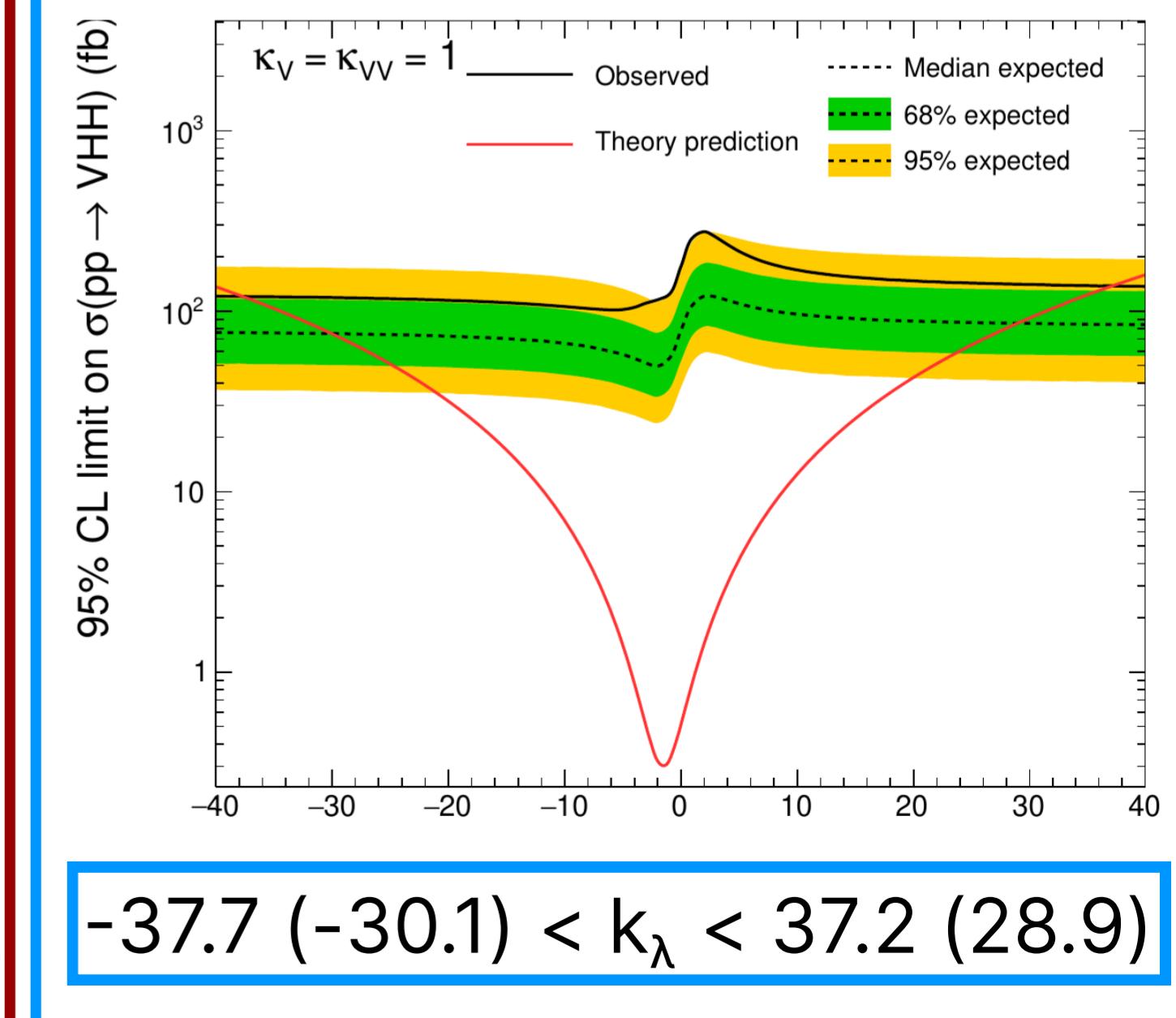
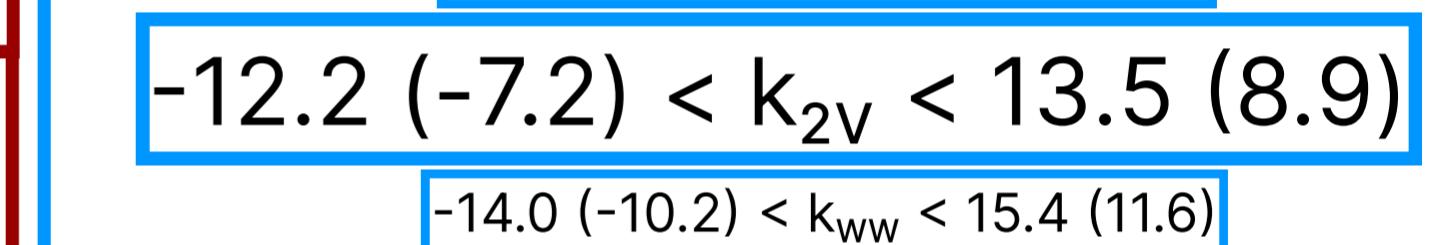
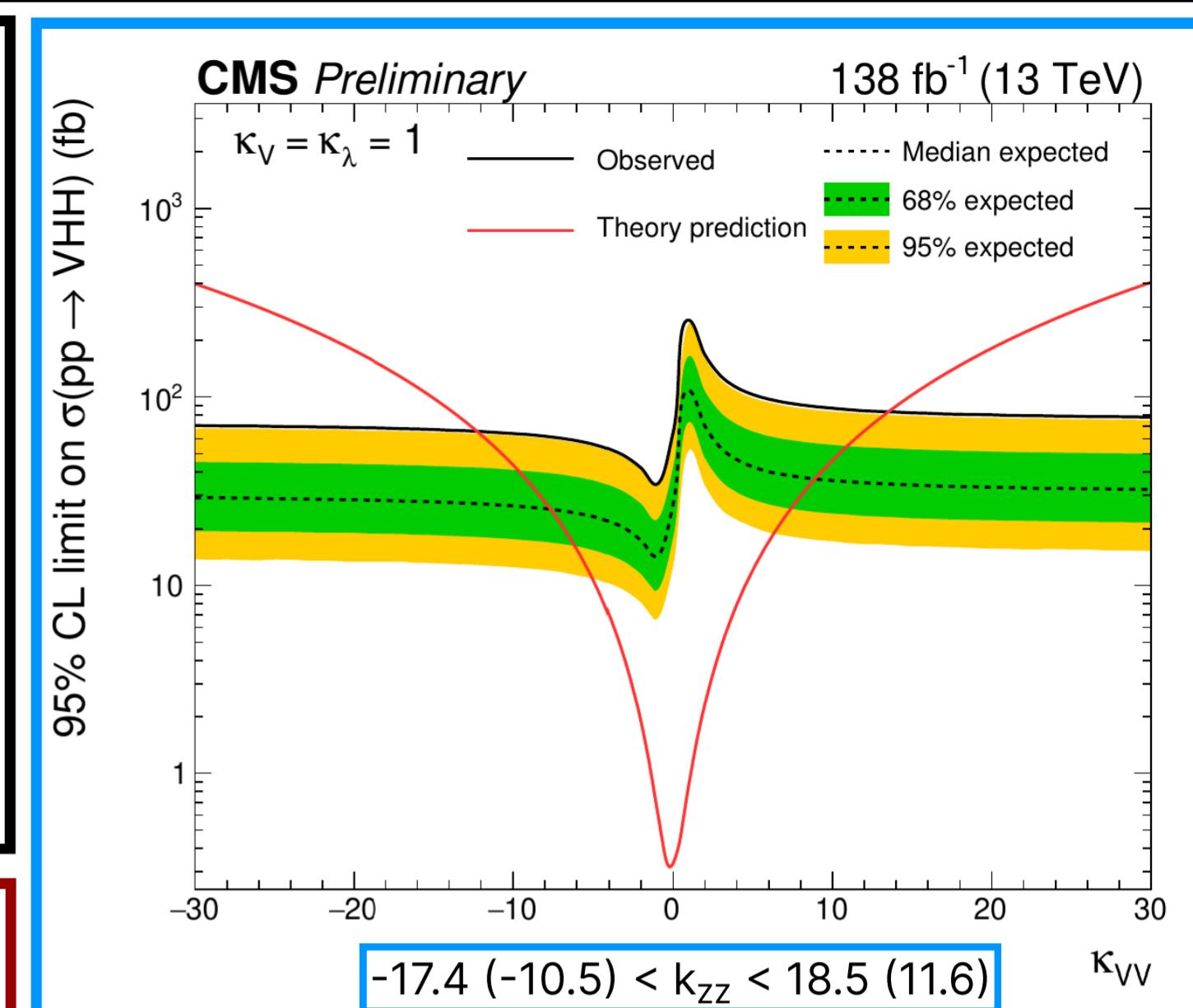
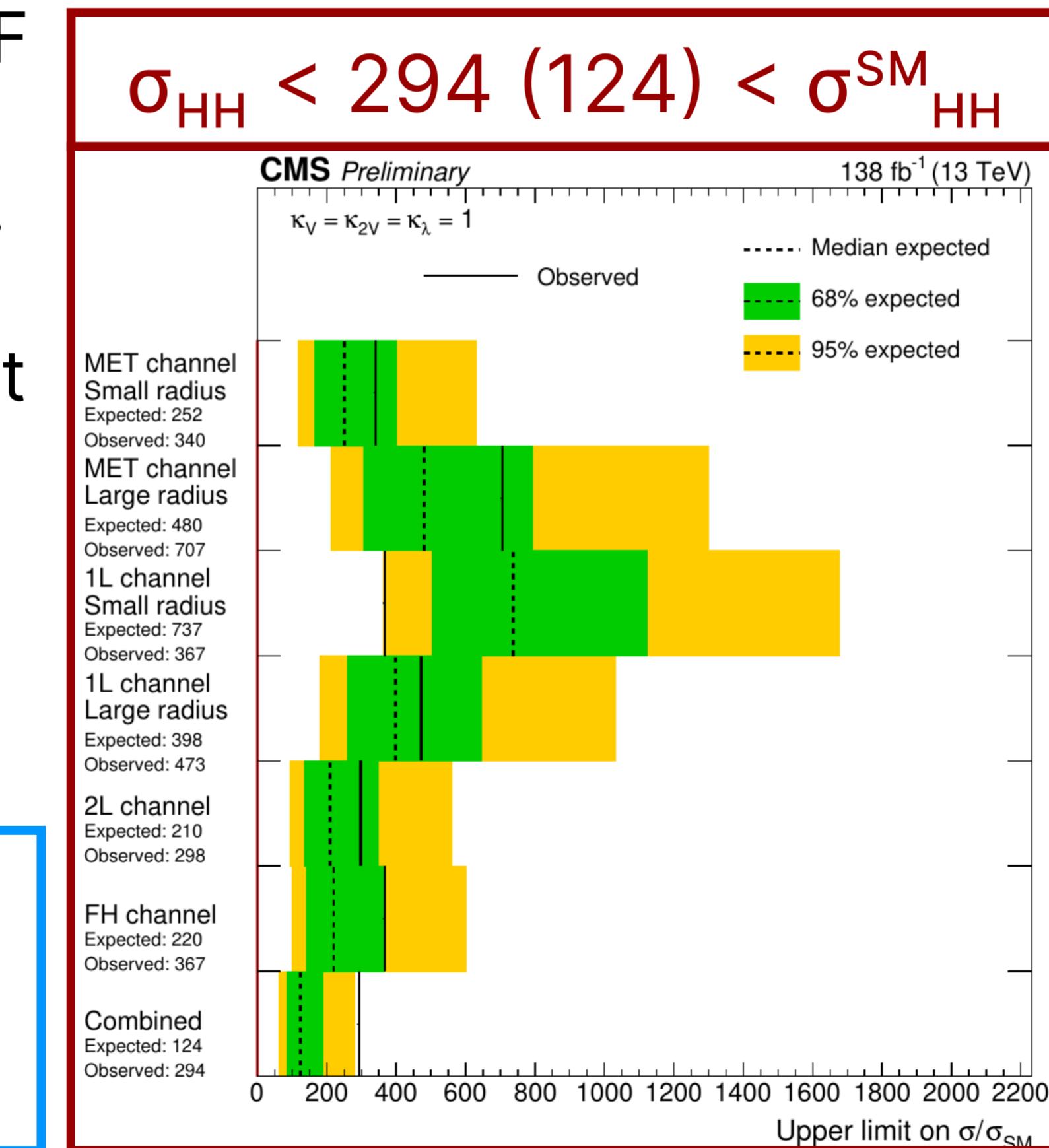
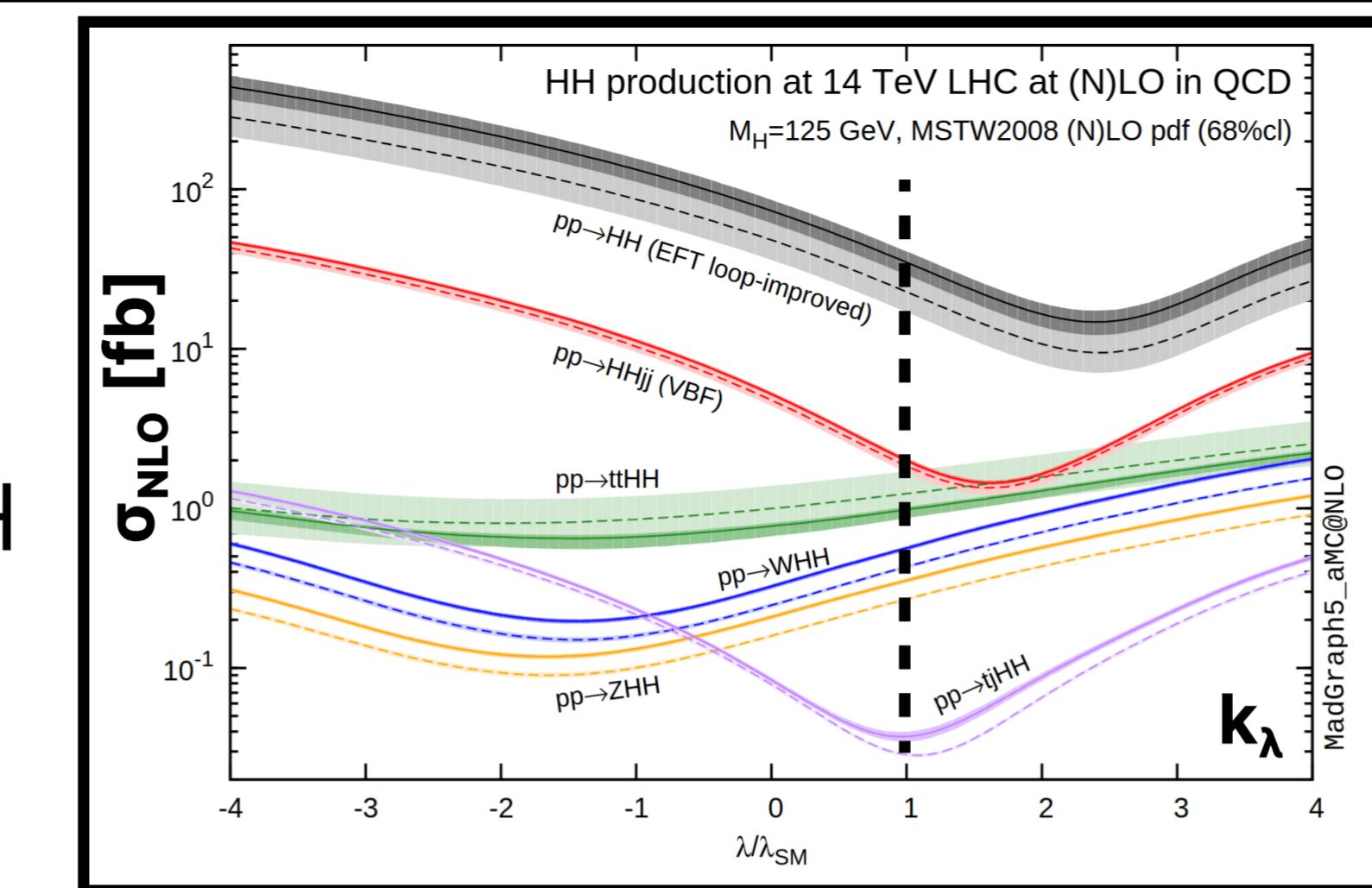
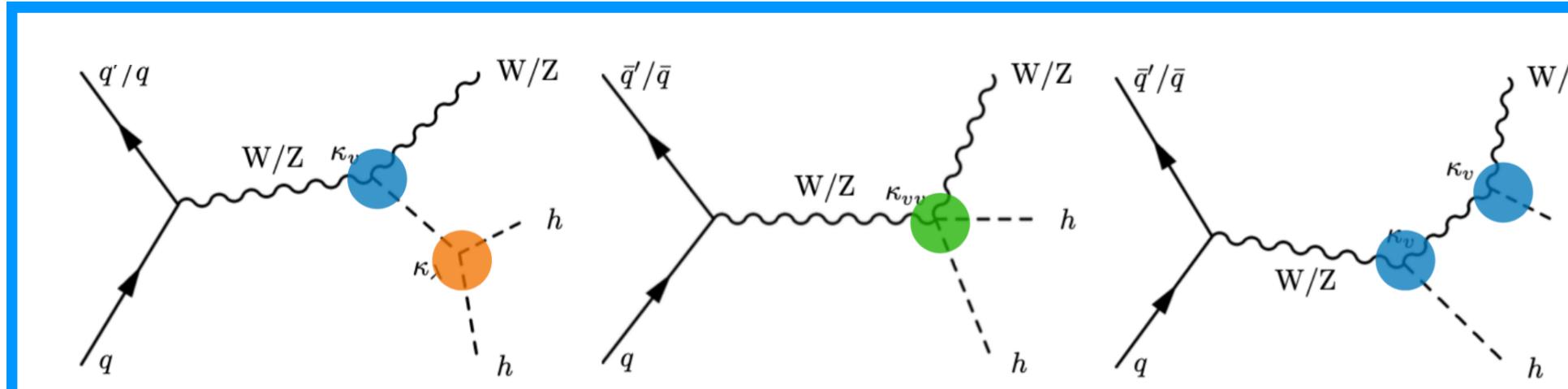
further suppressed

Credit: B. Moser

# VHH ( $\rightarrow 4b$ )

- VHH for the 1<sup>st</sup> time at CMS**
  - ~110 events expected** (before H decay to b's, without selection)
- Complementary to ggF and VBF!**
  - especially for  $4 < k_\lambda < 7$
  - because xsec comparable to ggF and VBF HH
- 4 channels:** 0/1/2 leptons and invis.
- 59 categories:** resolv./boosted,  $m_{\text{HH}}$ , #b-jets, signal- and tt-enhancement
- BDT and NN classifiers are used as signal vs bkg. discriminants
  - BDT defines regions sensitive to anomalous  $k_\lambda$  or  $k_{2V}$  hypoth.

## VHH LO

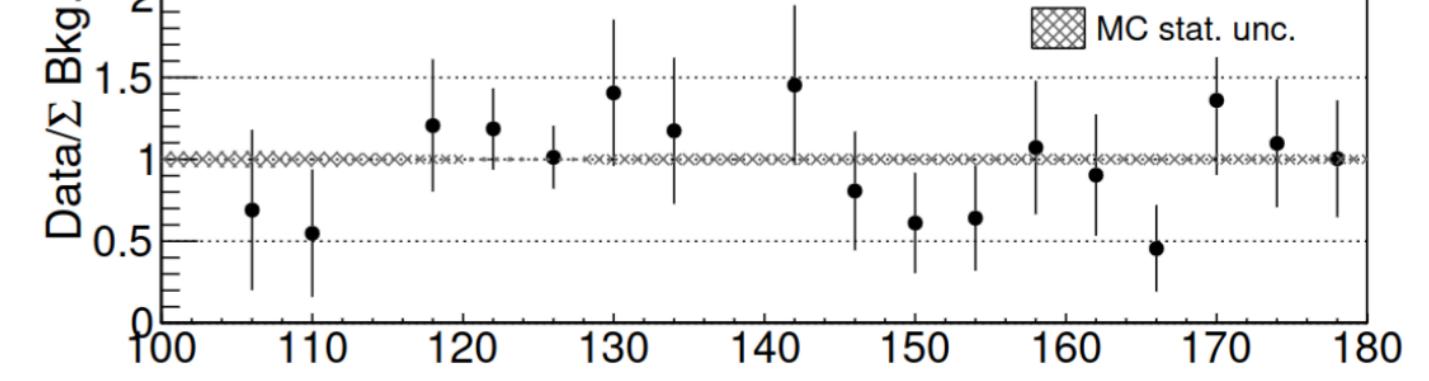
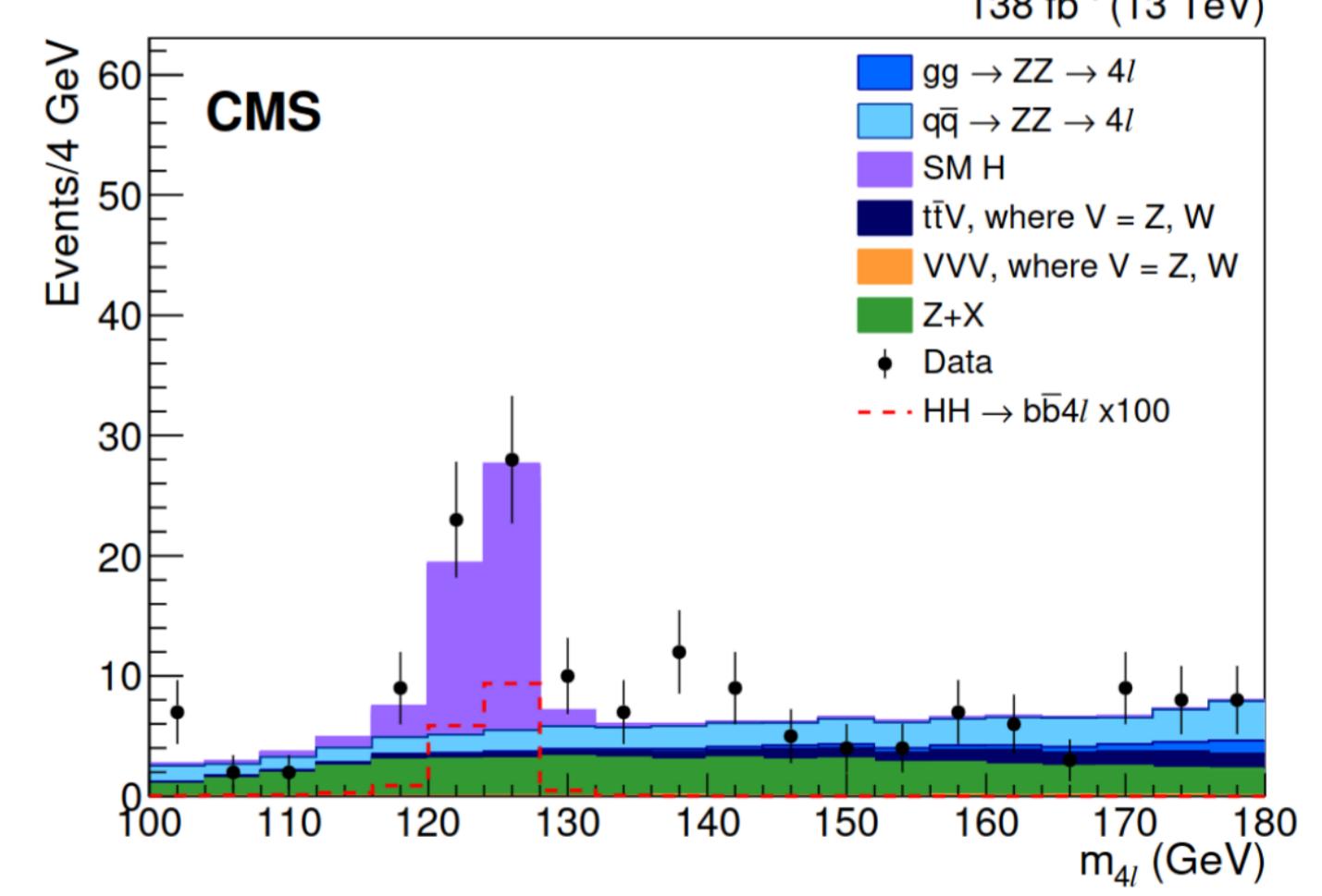
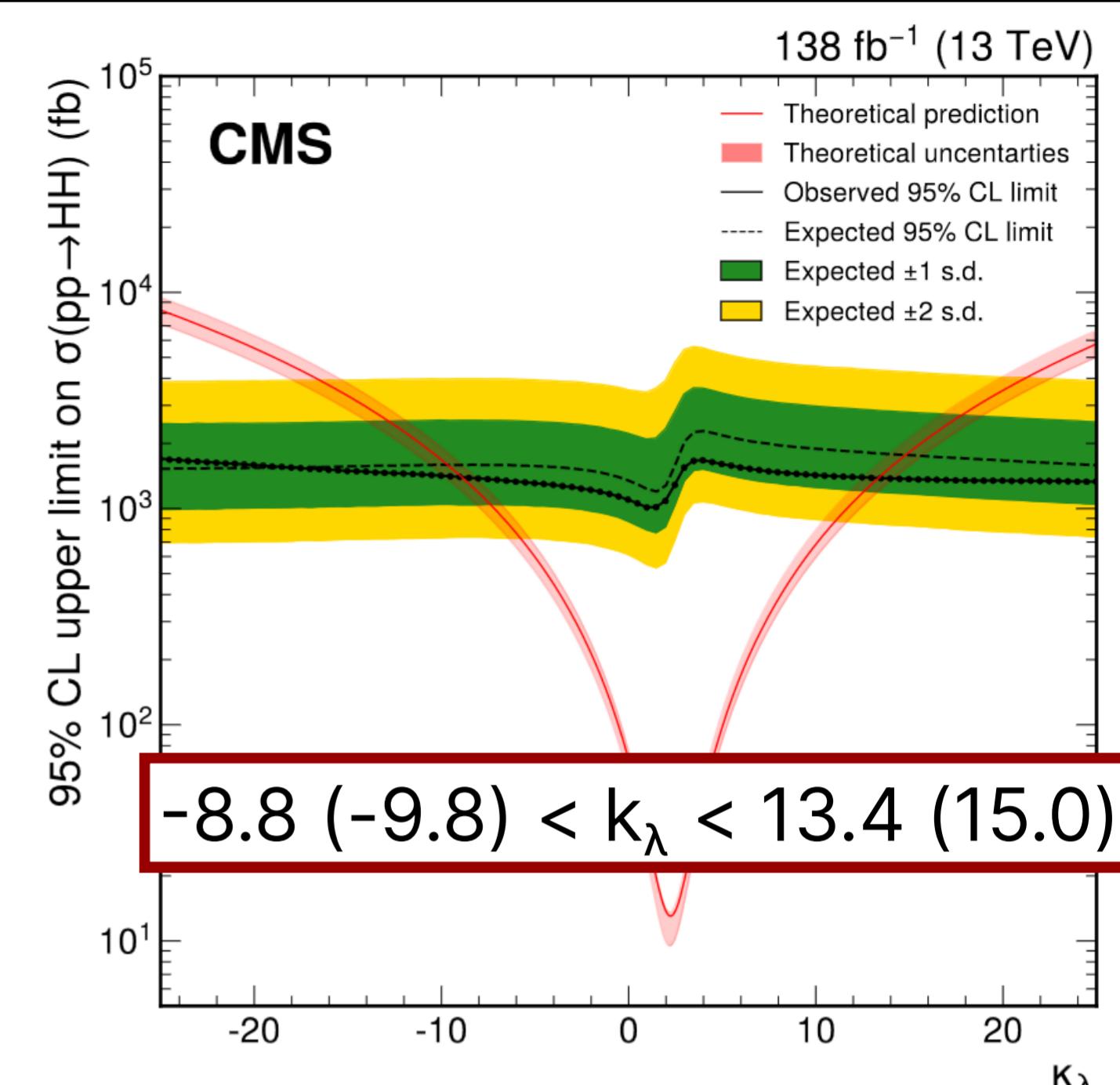
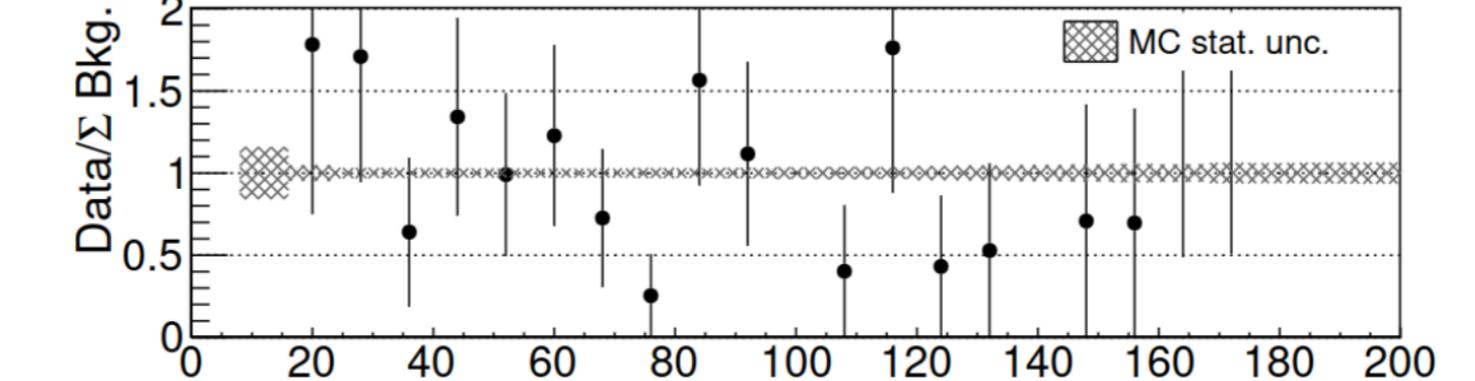
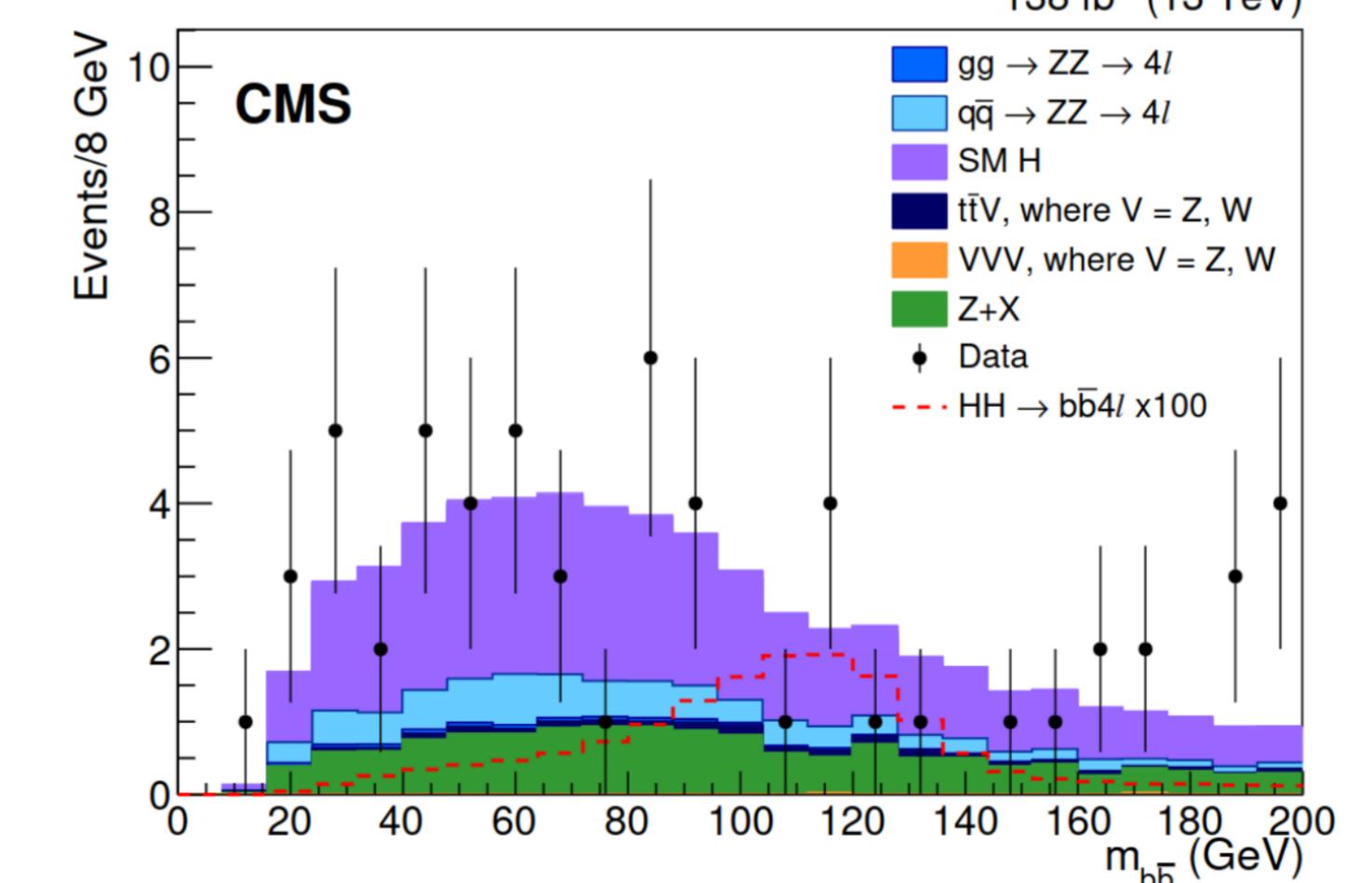
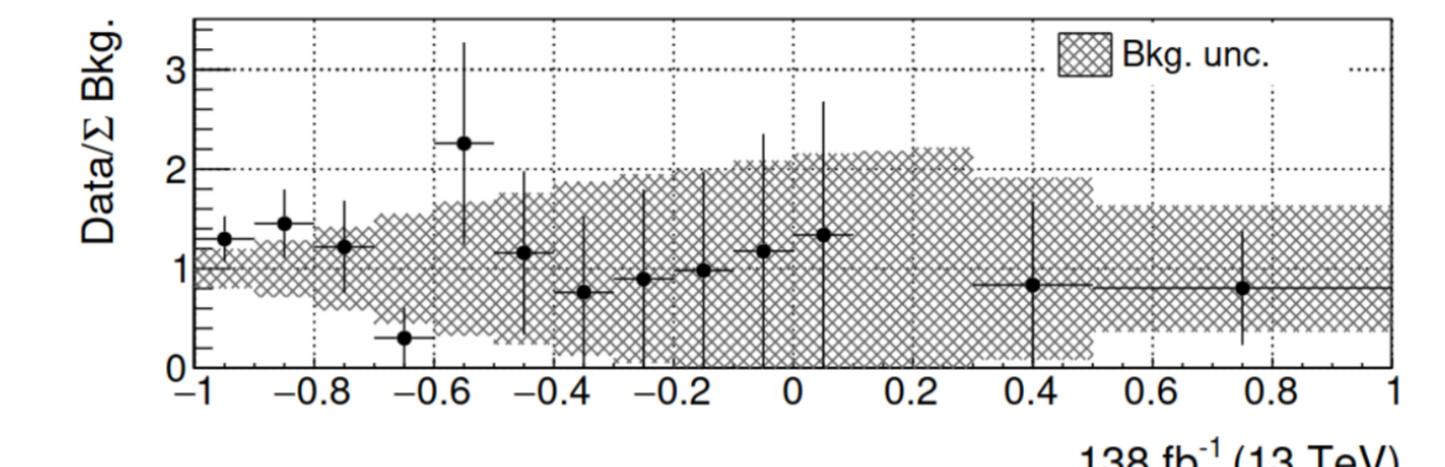
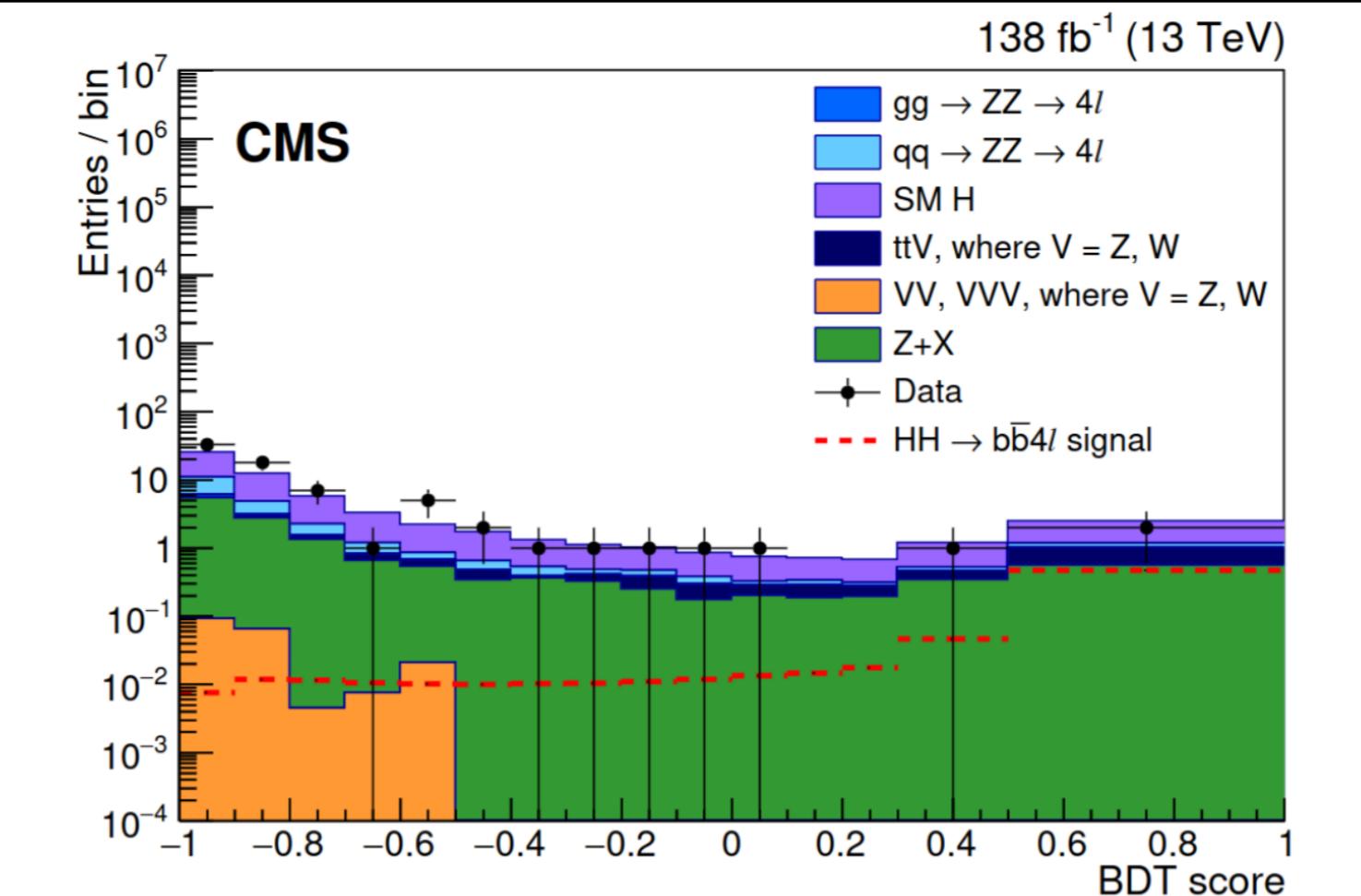


# $\text{HH} \rightarrow \text{bbZZ}$

ggF

- $\text{H} \rightarrow \text{ZZ}$  taken from [HIG-19-001](#)
- Select 2 extra jets w/ highest DeepCSV score
- “Fake” non-prompt leptons estimated from data
  - sources:  $e \rightarrow \gamma$  conversion, misrec. jets, HF decays
  - measure fake rate in  $Z + 1l + 2\text{jets}$  region
  - apply fake rate in  $Z + 2l + 2\text{jets}$  region
- Signal vs bkg. discrimination w/ BDT being fed full b-tagger distribution of jets
  - year- and channel-dependent training

$$\sigma_{\text{HH}} < 32.4 \text{ (39.6)} < \sigma^{\text{SM}}_{\text{HH}}$$



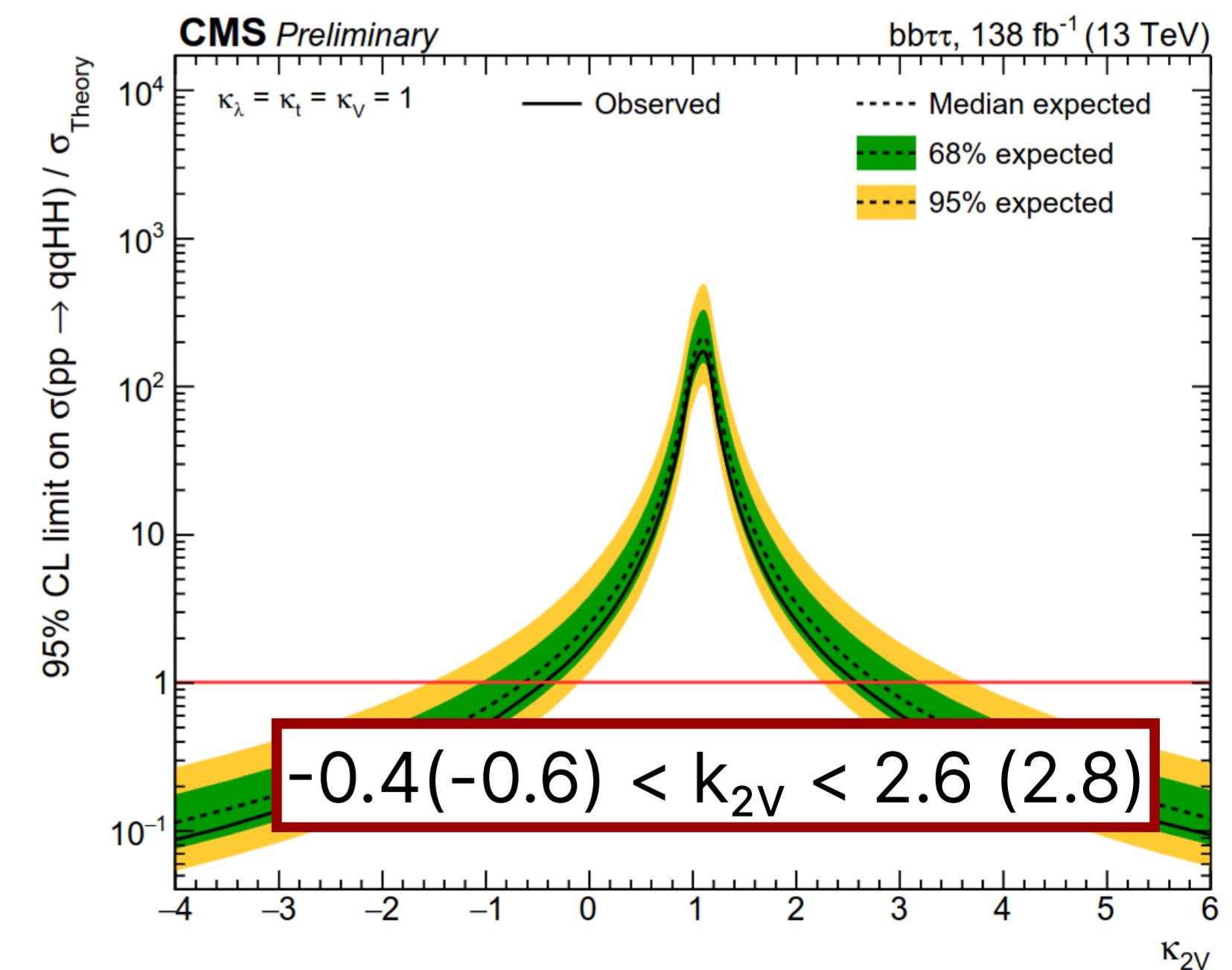
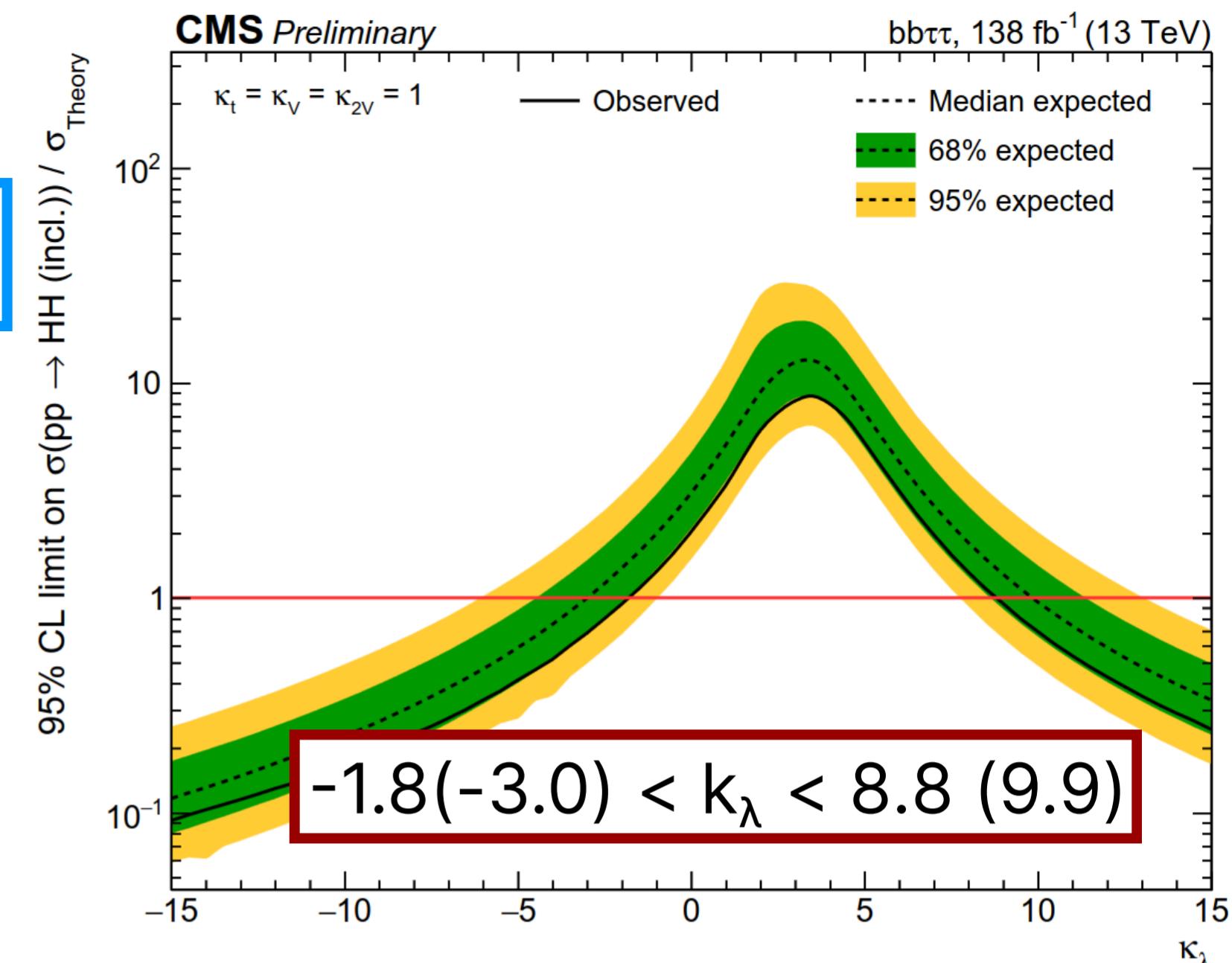
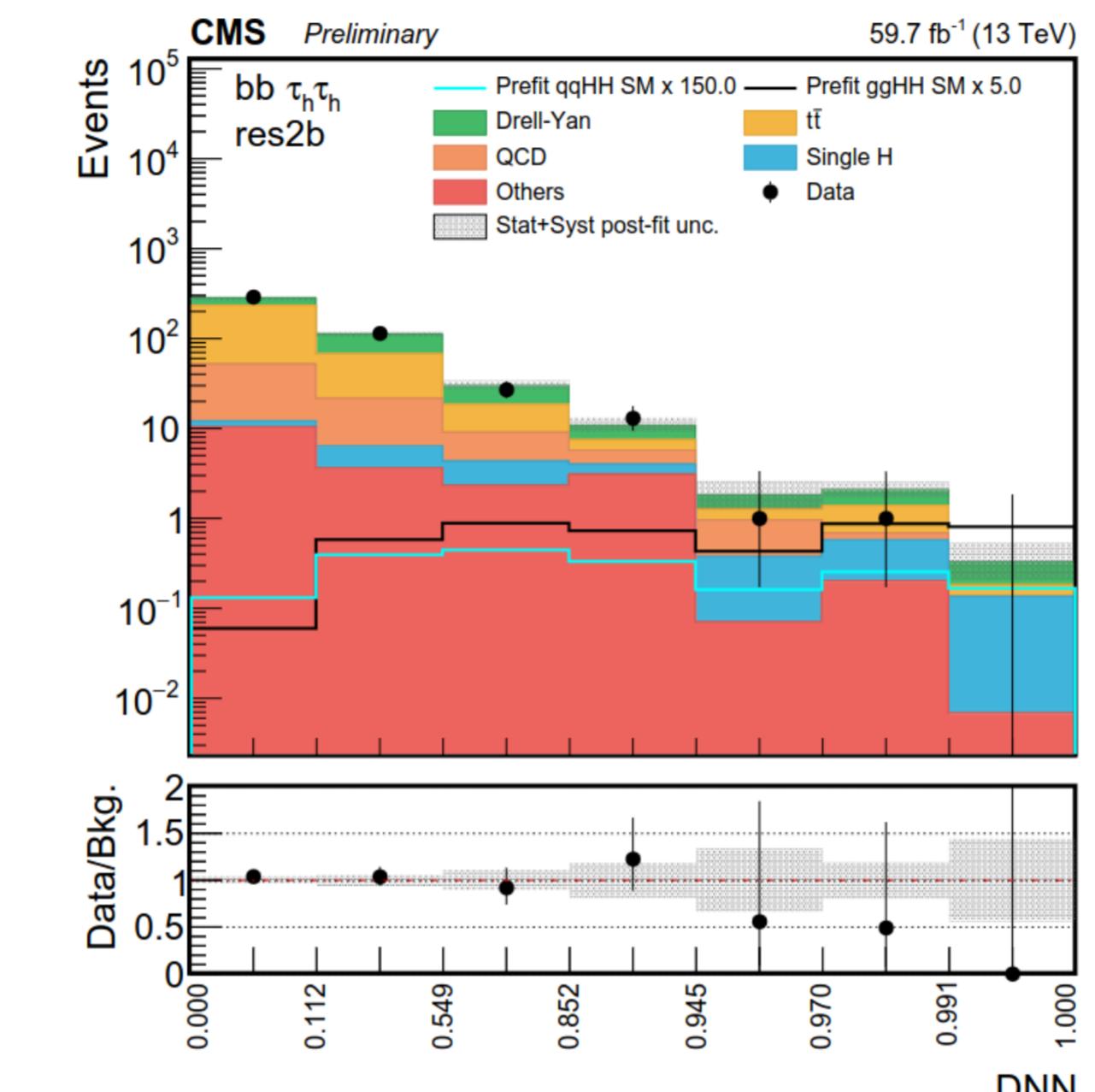
# $\text{HH} \rightarrow \text{bb}\tau\tau$

ggF VBF

- ID with DeepJet and DeepTau
- 3 channels based on  $\tau$  DM
- Categories: resolved, boosted and VBF-like
- Multi-classification approach to increase analysis sensitivity in the VBF category
  - 2 signal + 3 bkg. classes
- Fit the DNN score
  - most important features: DeepJet scores, inv. masses and many kinematic variables
  - two discriminators to enable inference on the entire dataset
  - ten networks per discriminator trained with 10-fold stratified cross-validation

$$\sigma_{\text{HH}} < 3.3 \text{ (5.2)} < \sigma_{\text{HH}}^{\text{SM}}$$

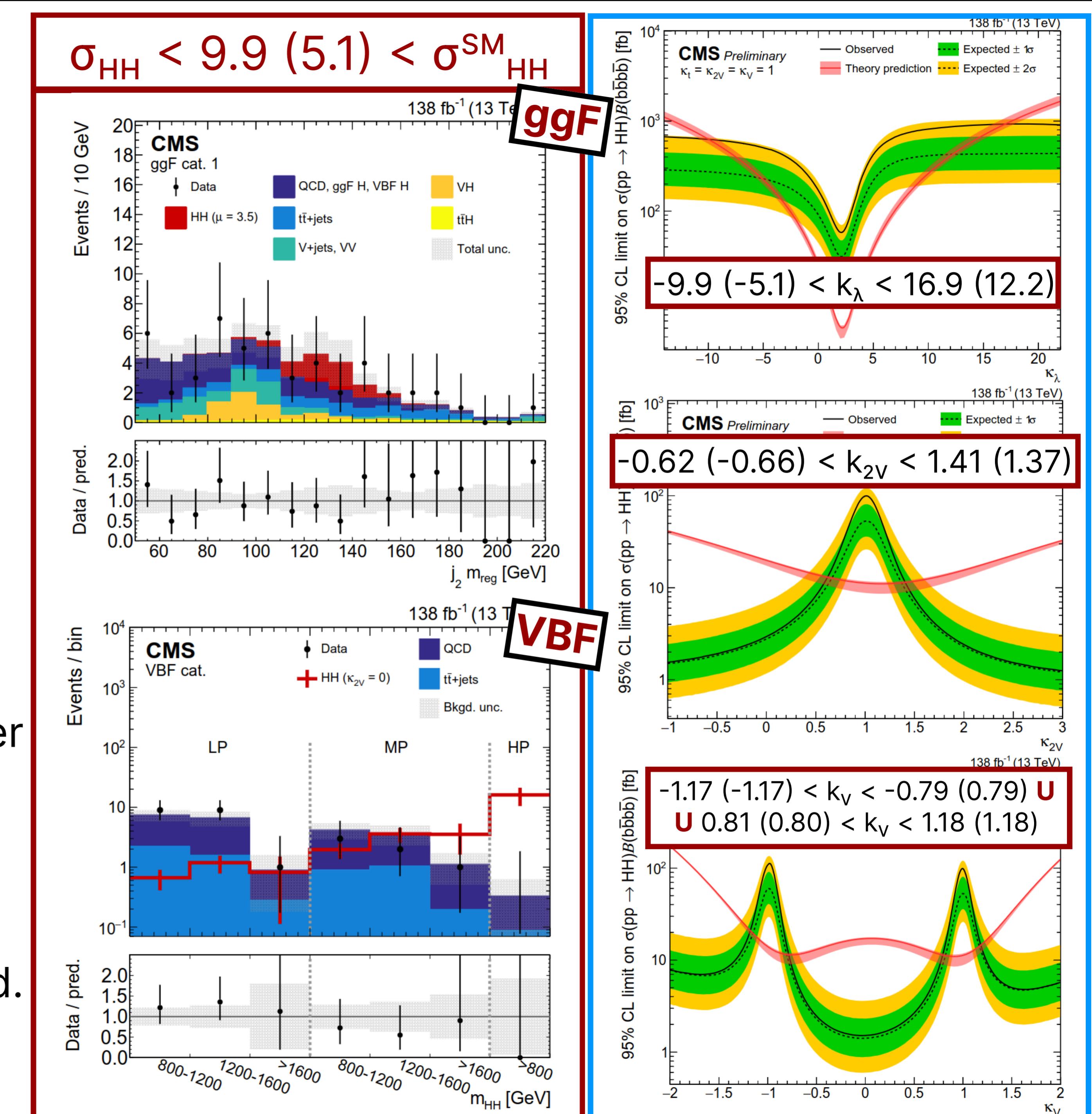
$$\sigma_{\text{VBF}} < 124 \text{ (154)} < \sigma_{\text{VBF}}^{\text{SM}}$$



# $\text{HH} \rightarrow 4\text{b}$ boosted

ggF VBF

- 2 AK8 jets w/  $p_T(\text{H}) > 300\text{GeV}$
- Background: 85% QCD, 15% ttbar
- **PNet tagger for AK8 jets**
  - discriminate QCD vs. b-jets
  - provides 4x improvement in bkg. rejection over DeepAK8-MD
  - $p_T$ -dependent calibration performed w/ data and QCD-enriched MC
- **PNet regressed jet mass  $m_{\text{PNet}}$** 
  - improved bkg. rejection wrt  $m_{\text{SD}}$
- ggF and VBF categories use PNet tagger
  - ggF also uses BDT, which has 2x better bkg. rejection wrt. cuts
- Simultaneous ML fit in all ggF and VBF categories, plus CRs (QCD and tt)
  - ggF: **fit to PNet mass** of one bb cand.
  - VBF: **fit to  $m_{\text{HH}}$**

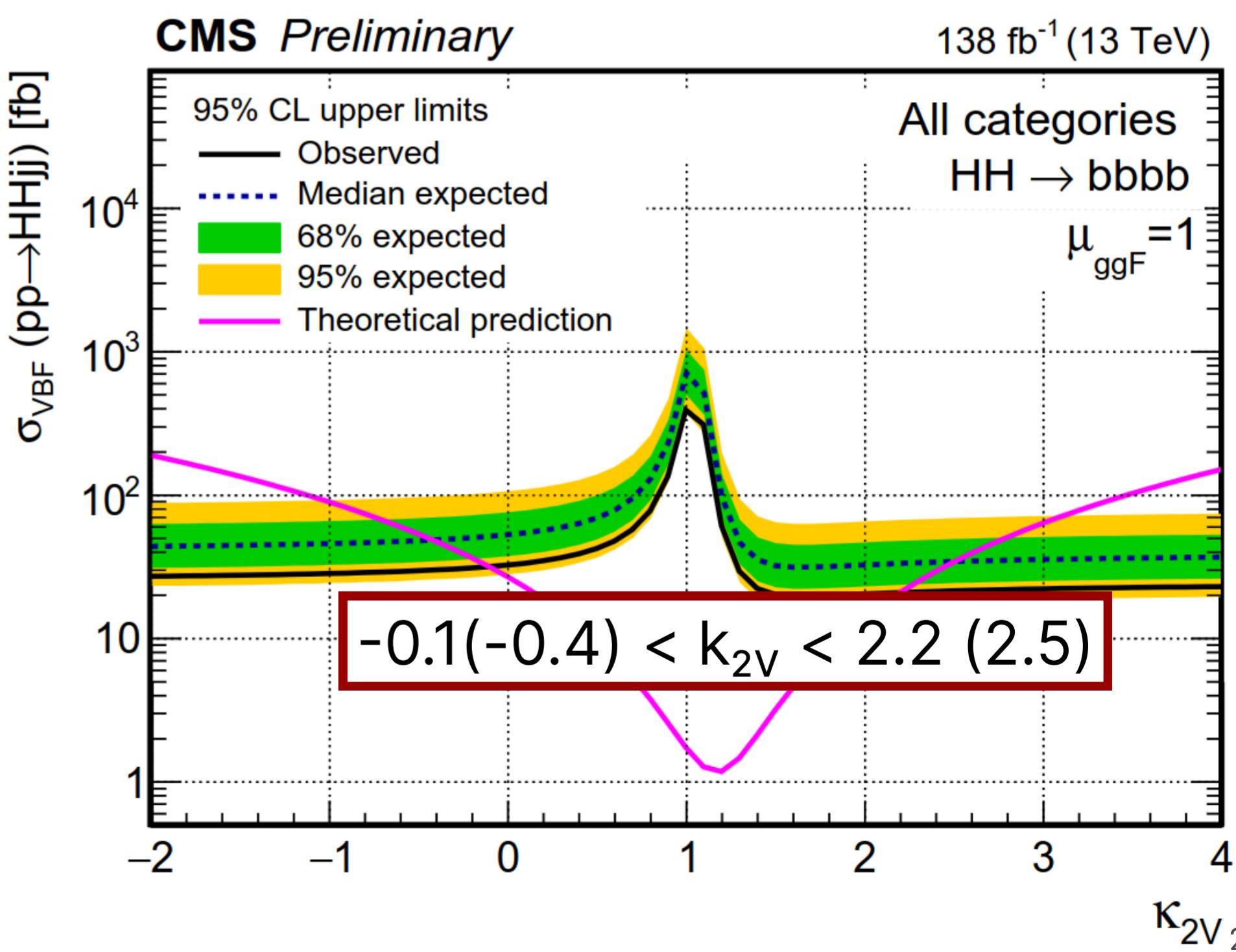
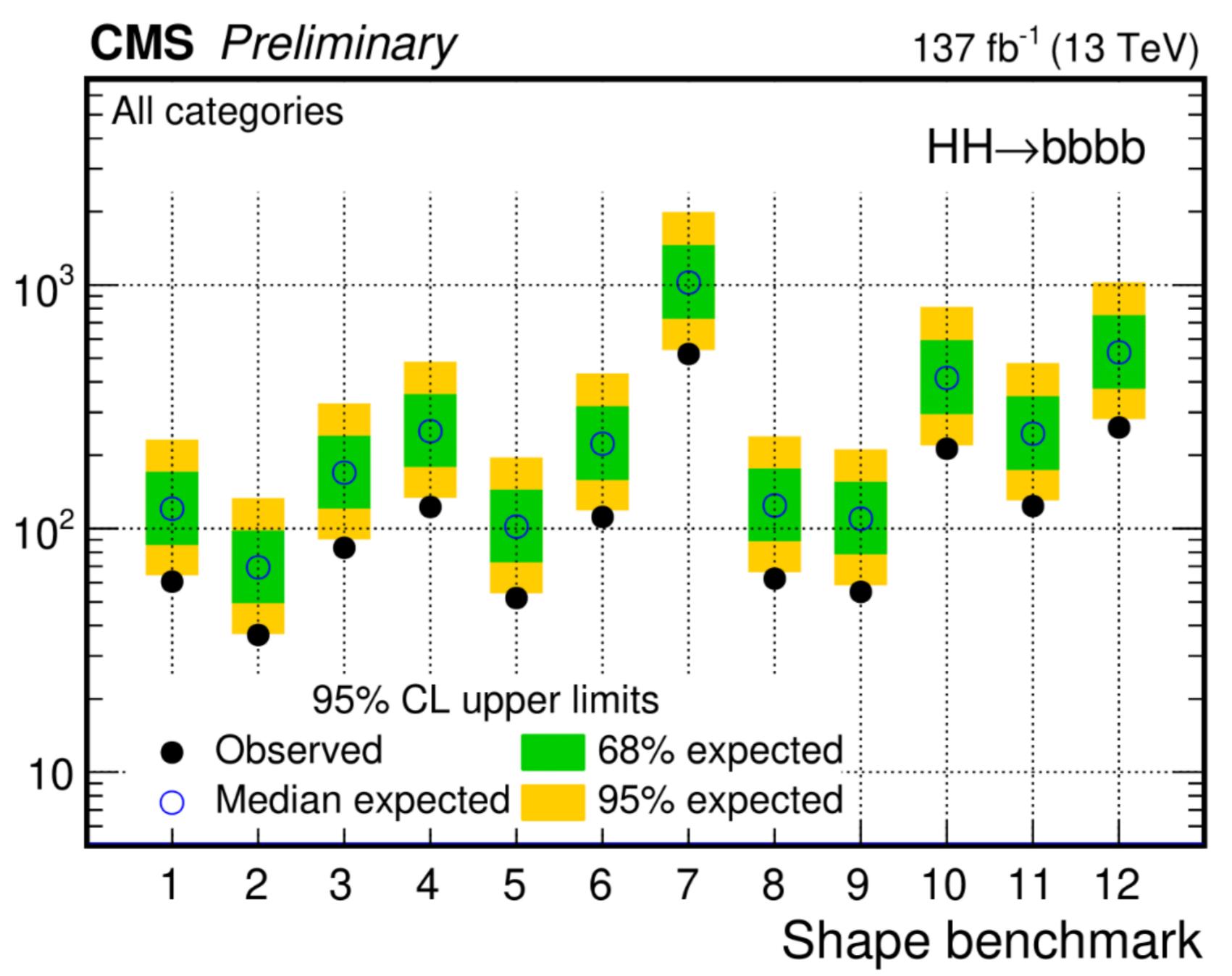
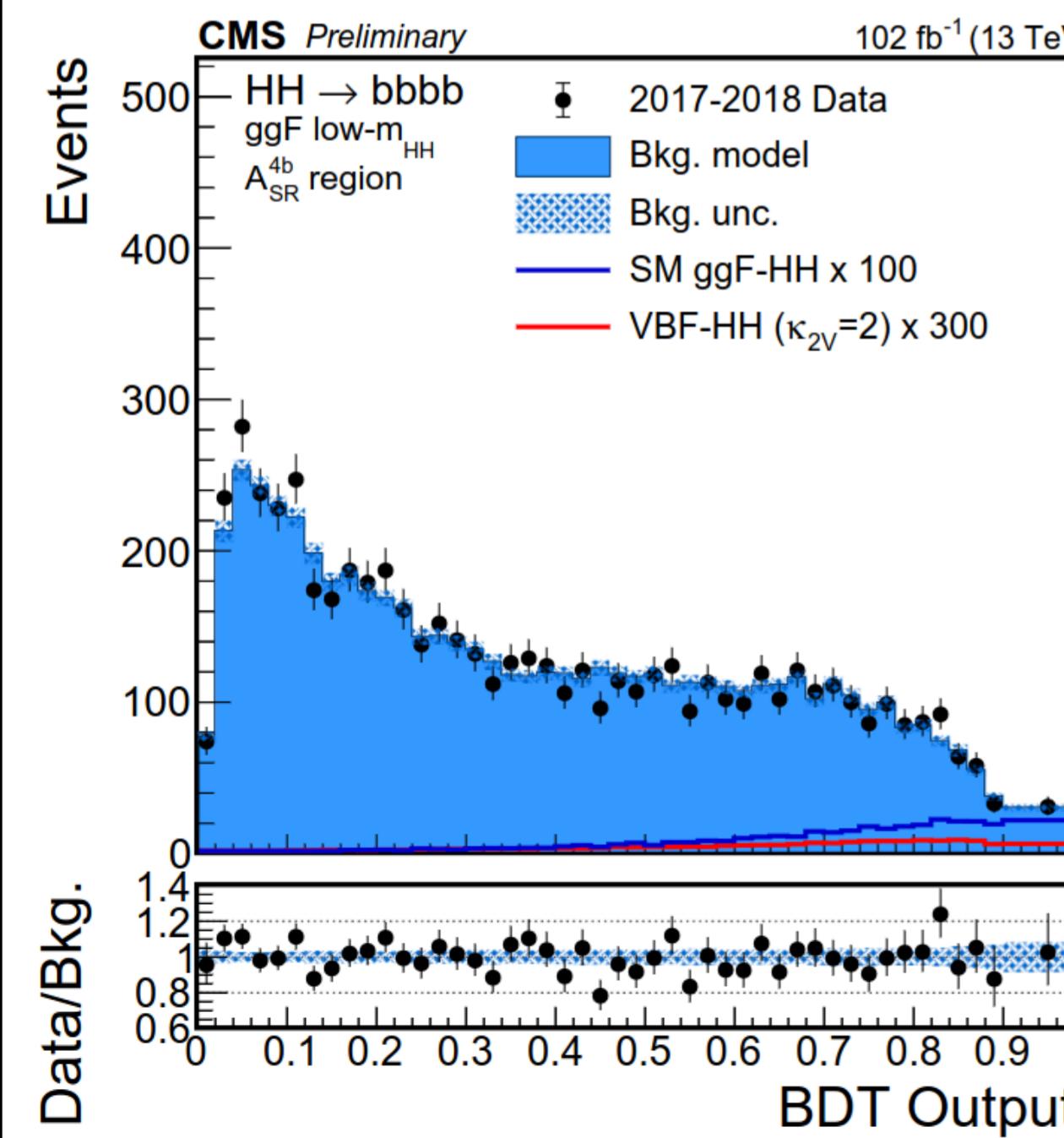
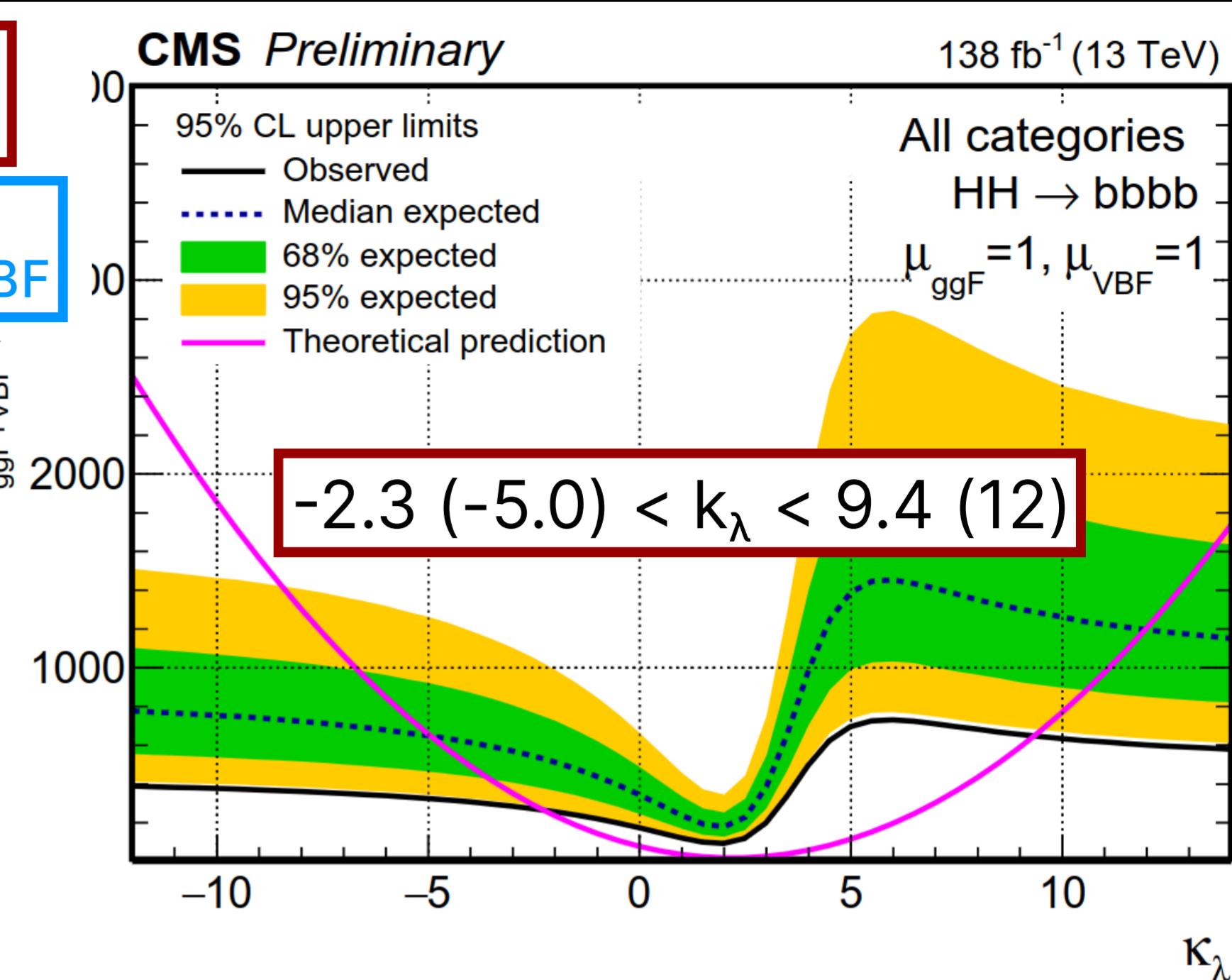


# HH $\rightarrow$ 4b resolved

ggF  
VBF

$\sigma_{\text{HH}} < 3.9 \text{ (7.8)} < \sigma^{\text{SM}}_{\text{HH}}$   
 $\sigma_{\text{VBF}} < 226 \text{ (412)} < \sigma^{\text{SM}}_{\text{VBF}}$

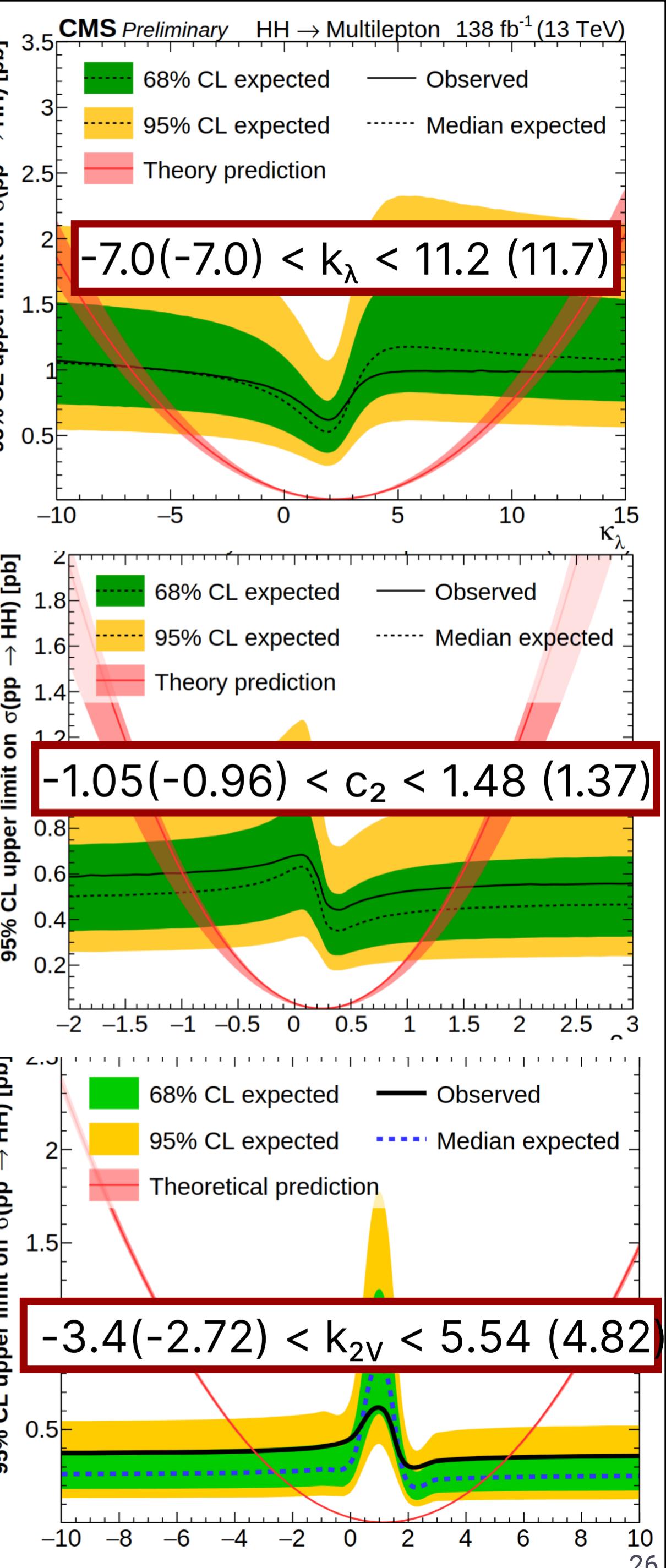
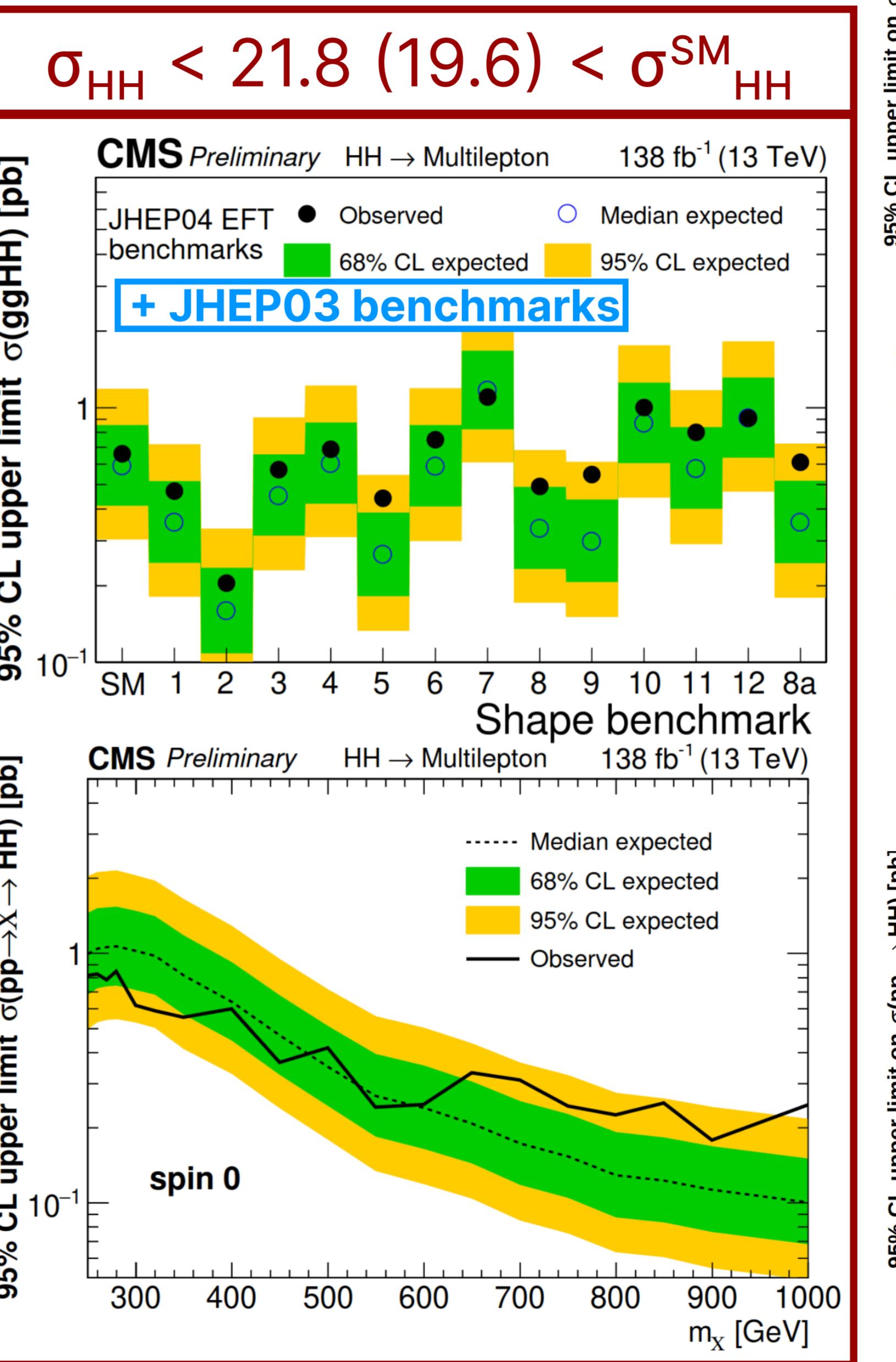
- Select jets with 4 highest DeepJet score
- 3 possible H pairs built w/  $m_{H_1}$  vs.  $m_{H_2}$  diagonal
- separate VBF and ggF categories
  - “ggF-killer” BDT increases purity in VBF category
  - additional low/high  $m_{\text{HH}}$  split for ggF
- QCD data-driven estimation using 3/4 b’s CRs
- Fit BDT ( $m_{\text{HH}}$ ) in ggF (VBF) categories



# HH $\rightarrow$ Multilepton (4V, 2V2 $\tau$ , 4 $\tau$ )

ggF VBF Res

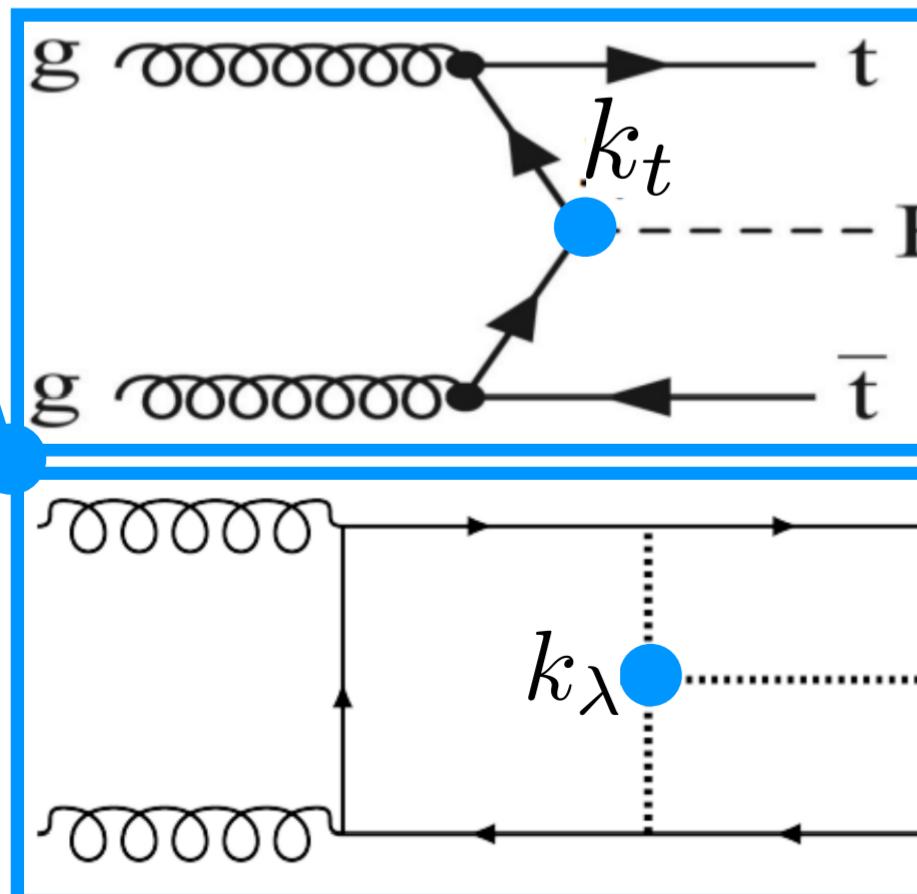
- 7.7% BR in total
- **7 channels**, depending on multipl. of hadronic  $\tau$ , electrons and muons
- Train 3 BDTs (spin0/2, nonres) per channel, parameterized on EFT benchmarks and resonance mass
- Background estimation
  - fakes: **fake factor method**
  - **lepton charge flip**: from data
  - irred. + photon conversion: from MC
- ML fit inputs: **1 BDT / channel + 2 CRs**
  - full stats used for BDT training
- 2 CRs to constrain WZ and ZZ bkgds.



# $\text{HH} \rightarrow \text{bb}\gamma\gamma$

ggF VBF

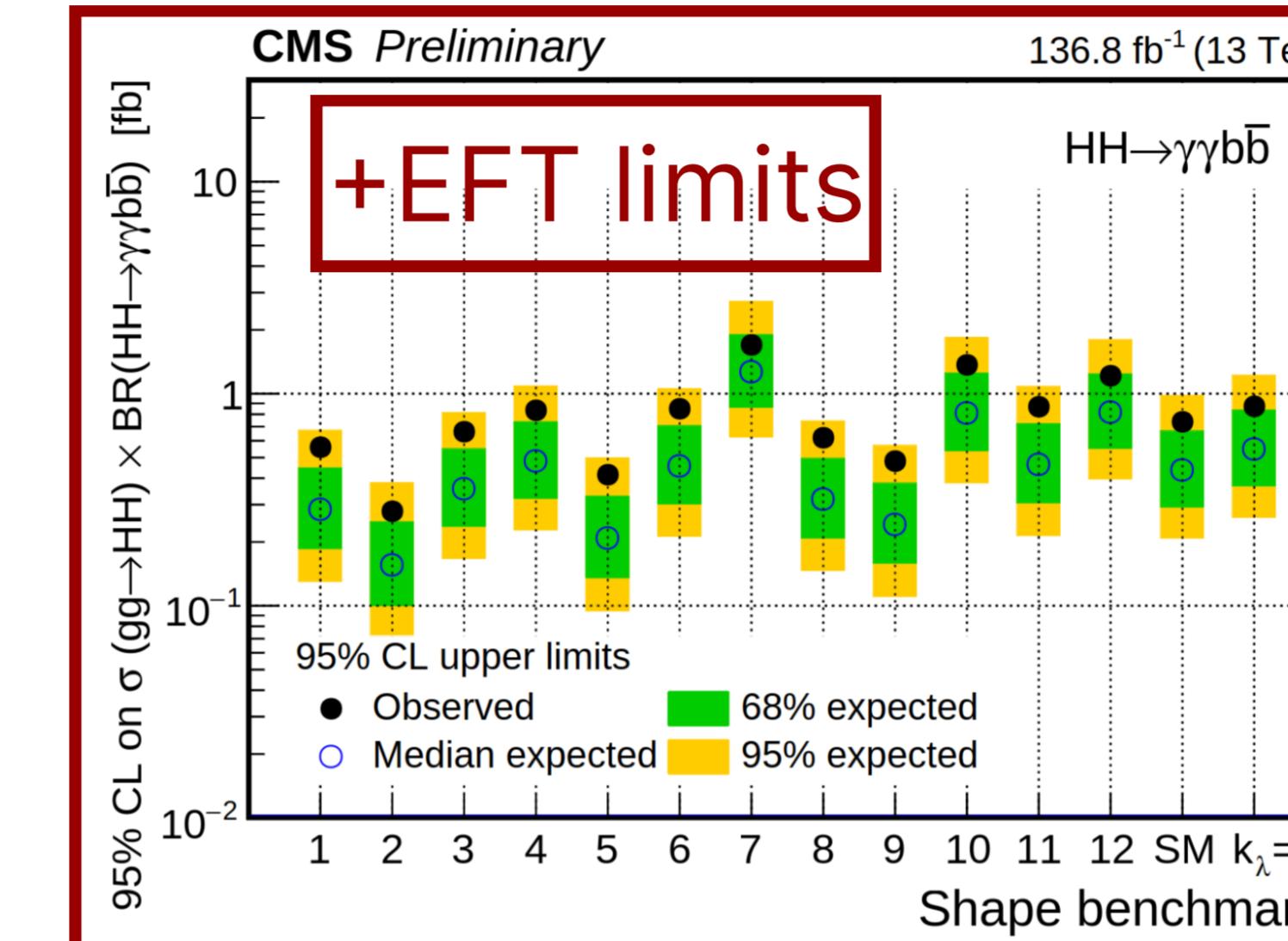
- optimized categories based on modified mass and ggF/VBF BDTs
- dedicated “ttH killer” DNN
- Signal extracted from unbinned 2D  $m_{\gamma\gamma}$  vs.  $m_{\text{bb}}$  parametric fit
  - $m_{\gamma\gamma}$ : sum of gaussians
  - $m_{jj}$ : double crystal-ball + gaus.
- HH+H combination: constrain  $\kappa_t$  w/ ttH phase-space



ttH analysis

$$\sigma_{\text{HH}} < 7.7 \text{ (5.2)} < \sigma^{\text{SM}}_{\text{HH}}$$

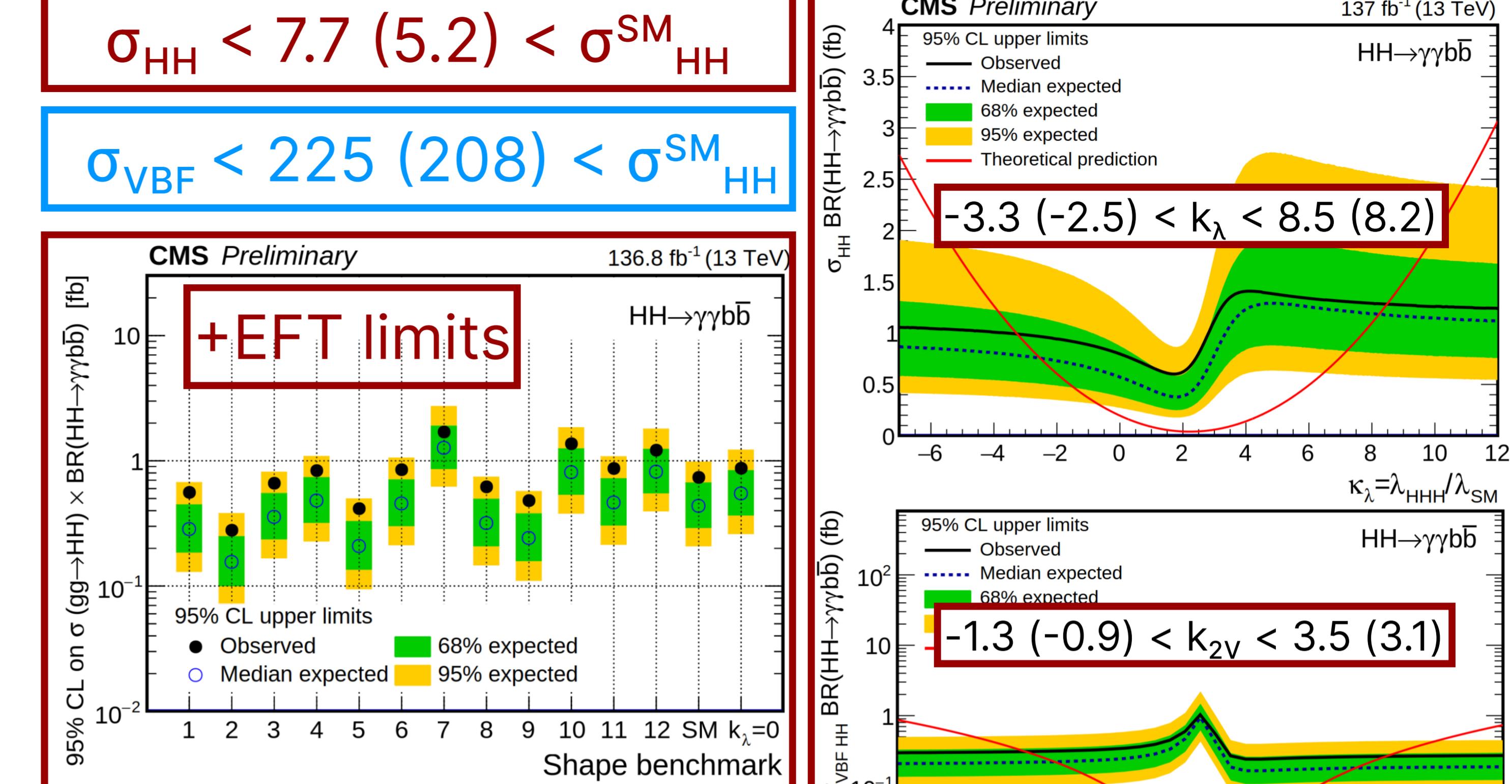
$$\sigma_{\text{VBF}} < 225 \text{ (208)} < \sigma^{\text{SM}}_{\text{HH}}$$



$\kappa_t$  parametrization is not reliable

Include ttH

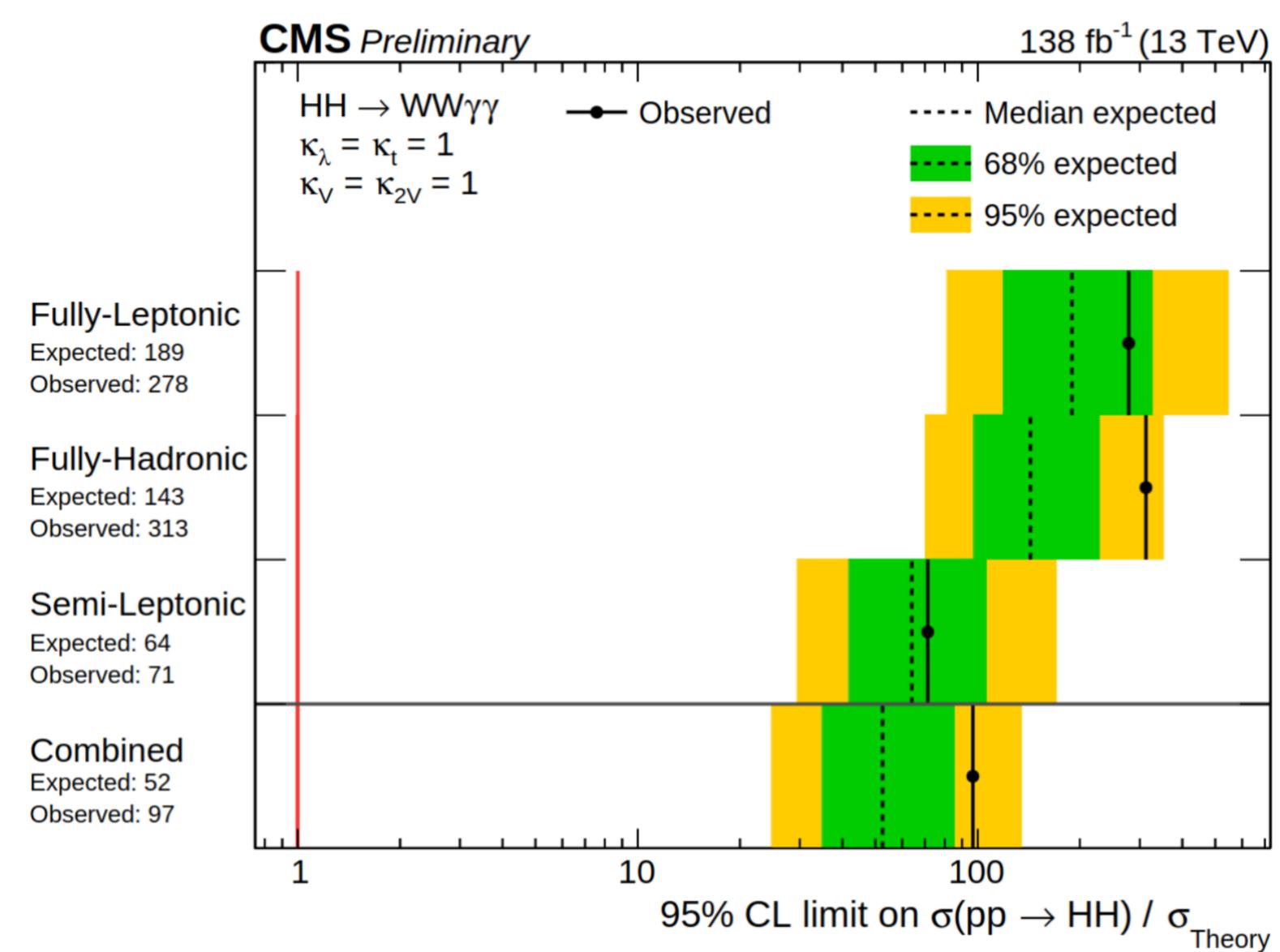
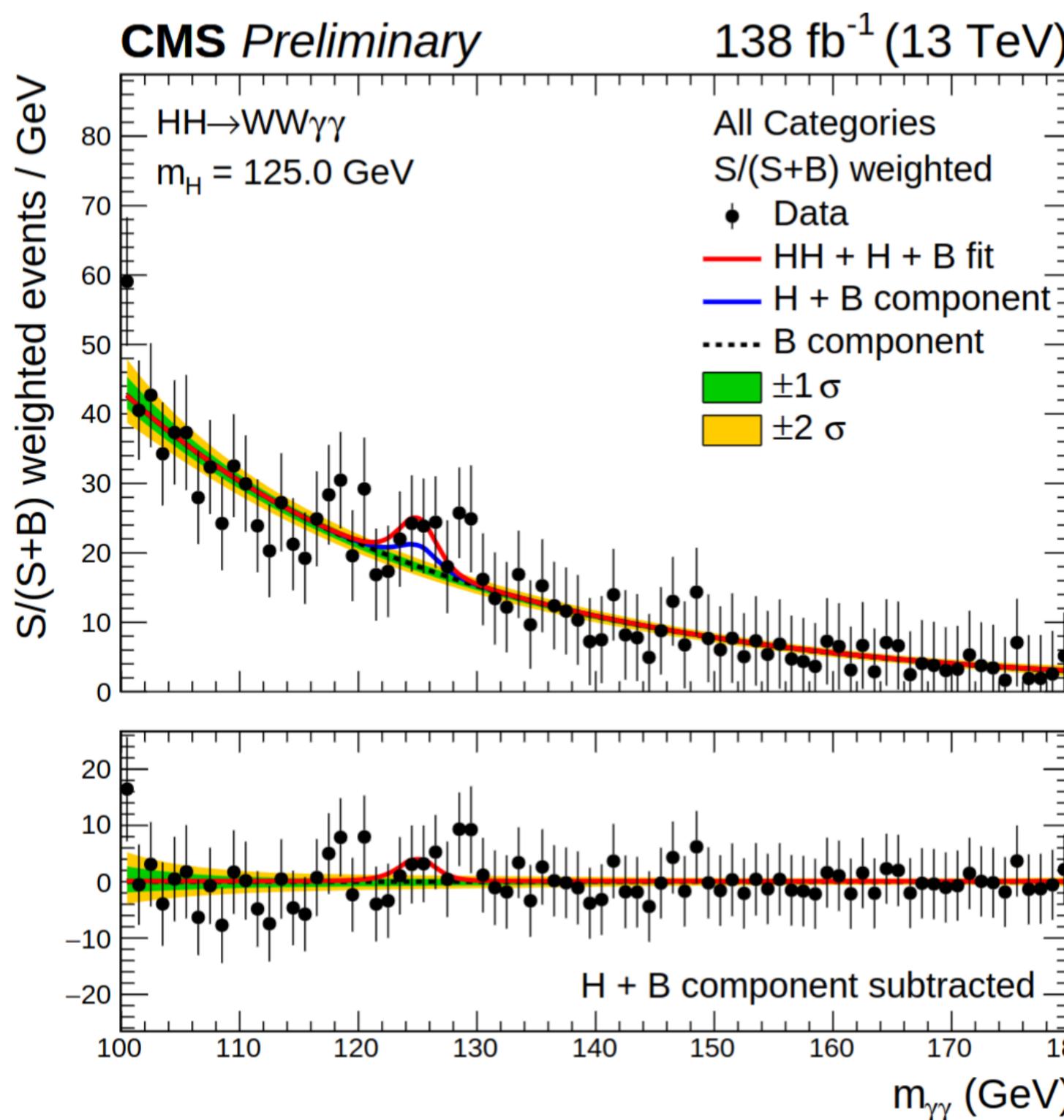
$$-2.7 \text{ (-3.3)} < \kappa_\lambda < 8.6 \text{ (8.6)}$$



# $\text{HH} \rightarrow \text{WW}\gamma\gamma$

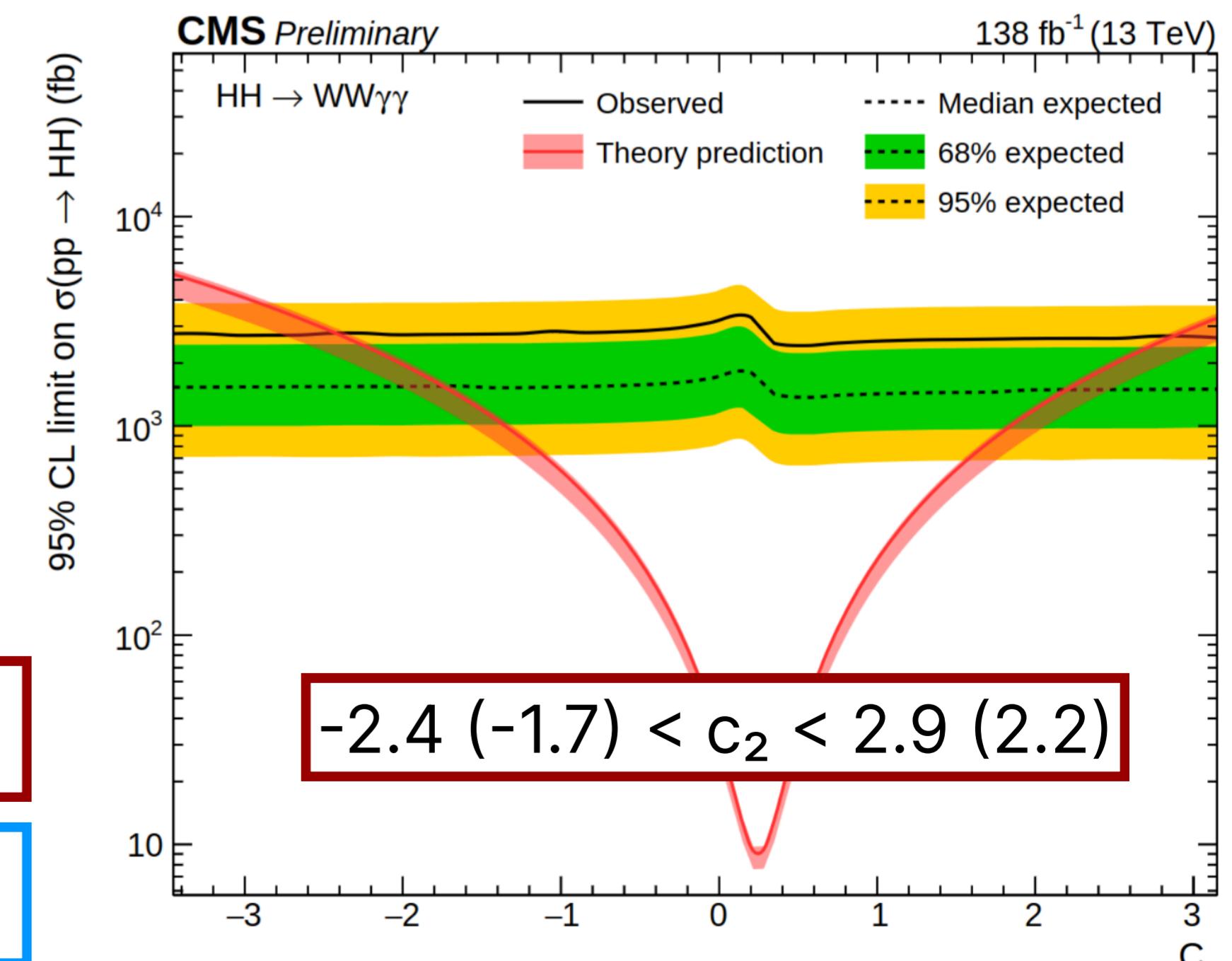
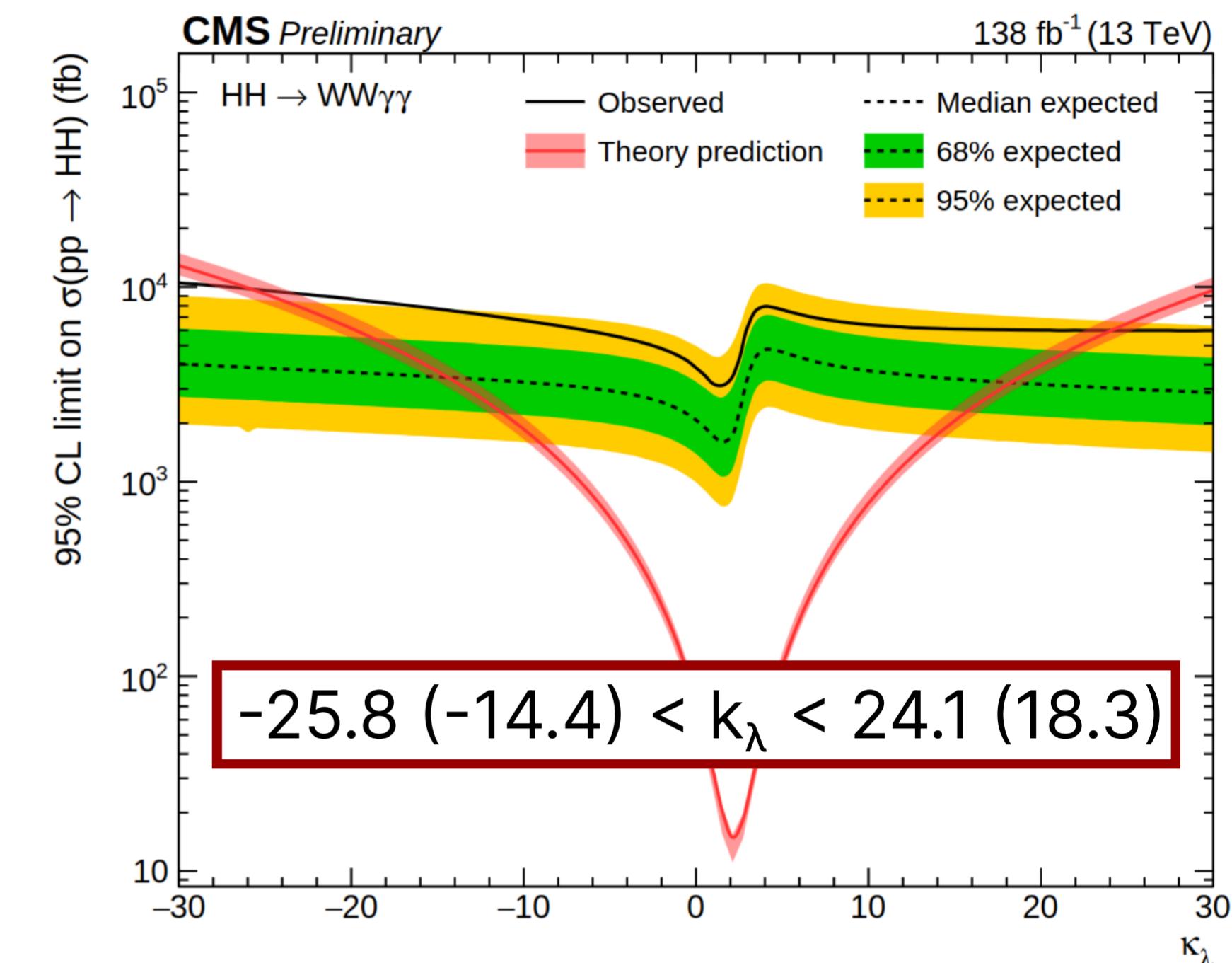
ggF

- Signal extracted from  $m_{\gamma\gamma}$  parametric fit
- 3 channels based on #leptons
  - 0: multiclass DNN to remove H and  $\gamma/\text{jets}$  bkg. + binary DNN for EFT benchmarks
  - 1:  $\text{WW}\gamma\gamma$  id and “ $\text{bb}\gamma\gamma$  killer” binary DNNs
  - 2: cut based

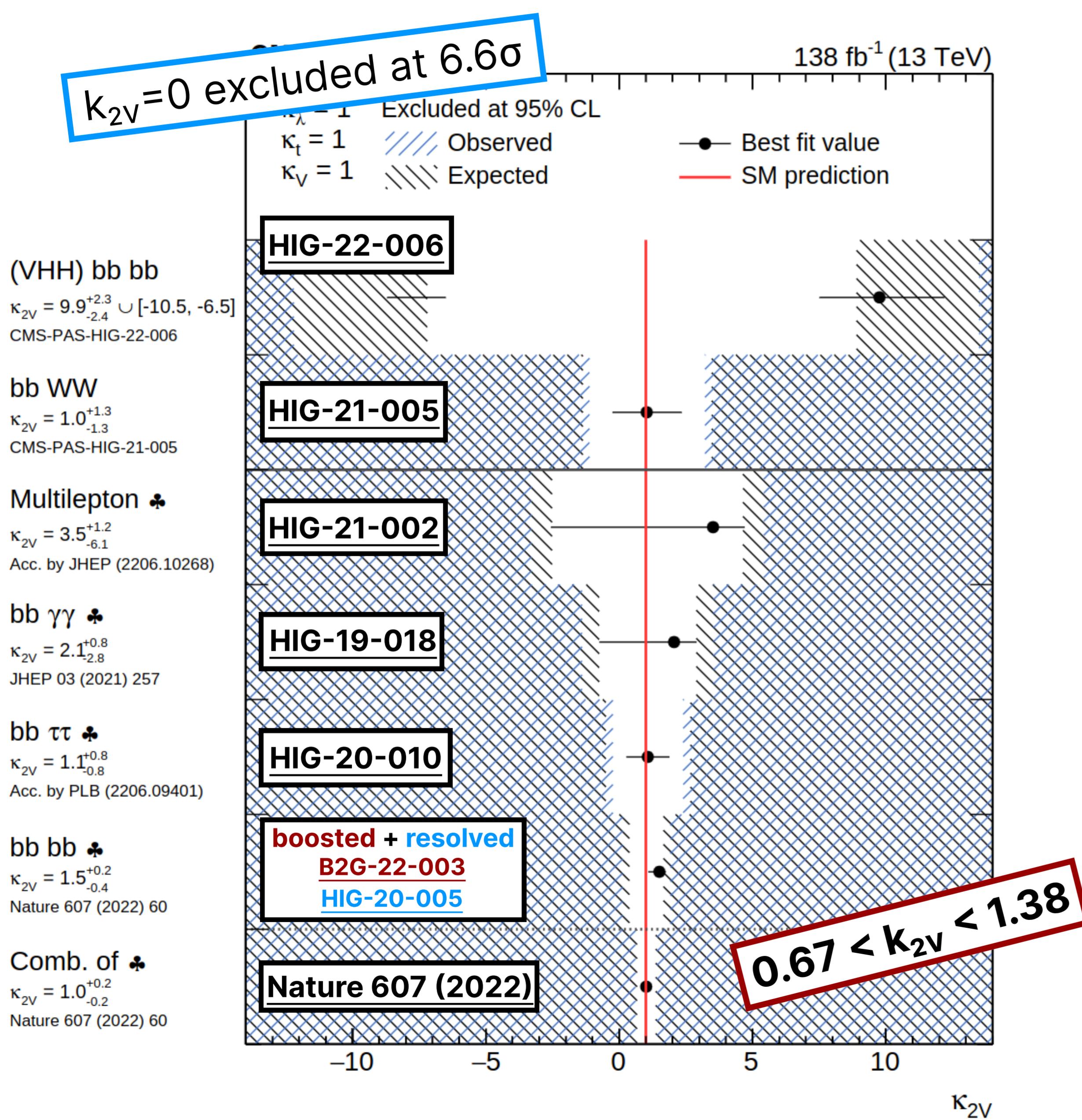


$$\sigma_{\text{HH}} < 96.8 (52.5) < \sigma^{\text{SM}}_{\text{HH}}$$

$$\sigma_{\text{EFT}} < 1.7 - 6.2 (1.0 - 3.9)$$



# CMS Internal links



$\kappa_t = 1$  Excluded at 95% CL  
 $\kappa_V = 1$  Observed  
 $\kappa_{2V} = 1$  Expected  
 ● Best fit value  
 — SM prediction

**HIG-22-006**

(VHH) bb bb  
 $\kappa_\lambda = -25.1^{+6.8}_{-5.6}$   
 CMS-PAS-HIG-22-006

WW  $\gamma\gamma$   
 $\kappa_\lambda = 14.8^{+5.5}_{-13.3}$   
 CMS-PAS-HIG-21-014

bb WW  
 $\kappa_\lambda = 4.2^{+5.3}_{-5.7}$   
 CMS-PAS-HIG-21-005

bb ZZ ♣  
 $\kappa_\lambda = 2.3^{+5.6}_{-5.4}$   
 Acc. by JHEP (2206.10657)

Multilepton ♣  
 $\kappa_\lambda = 2.3^{+5.2}_{-5.2}$   
 Acc. by JHEP (2206.10268)

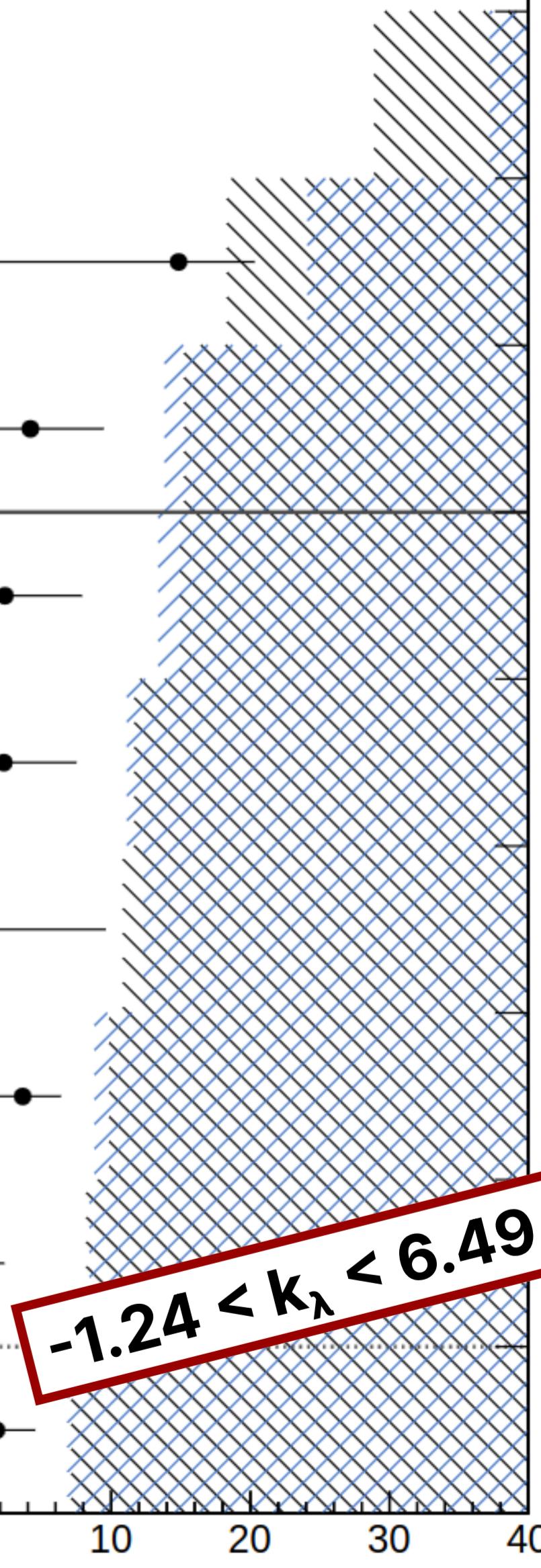
bb bb ♣  
 $\kappa_\lambda = -0.2^{+9.9}_{-2.8}$   
 Nature 607 (2022) 60

bb  $\gamma\gamma$  ♣  
 $\kappa_\lambda = 3.6^{+2.8}_{-2.9}$   
 JHEP 03 (2021) 257

bb  $\tau\tau$  ♣  
 $\kappa_\lambda = -0.2^{+2.5}_{-1.7}$   
 Acc. by PLB (2206.09401)

Comb. of ♣  
 $\kappa_\lambda = 1.7^{+2.8}_{-1.7}$   
 Nature 607 (2022) 60

● Best fit value  
 — SM prediction



# CMS Internal links

