

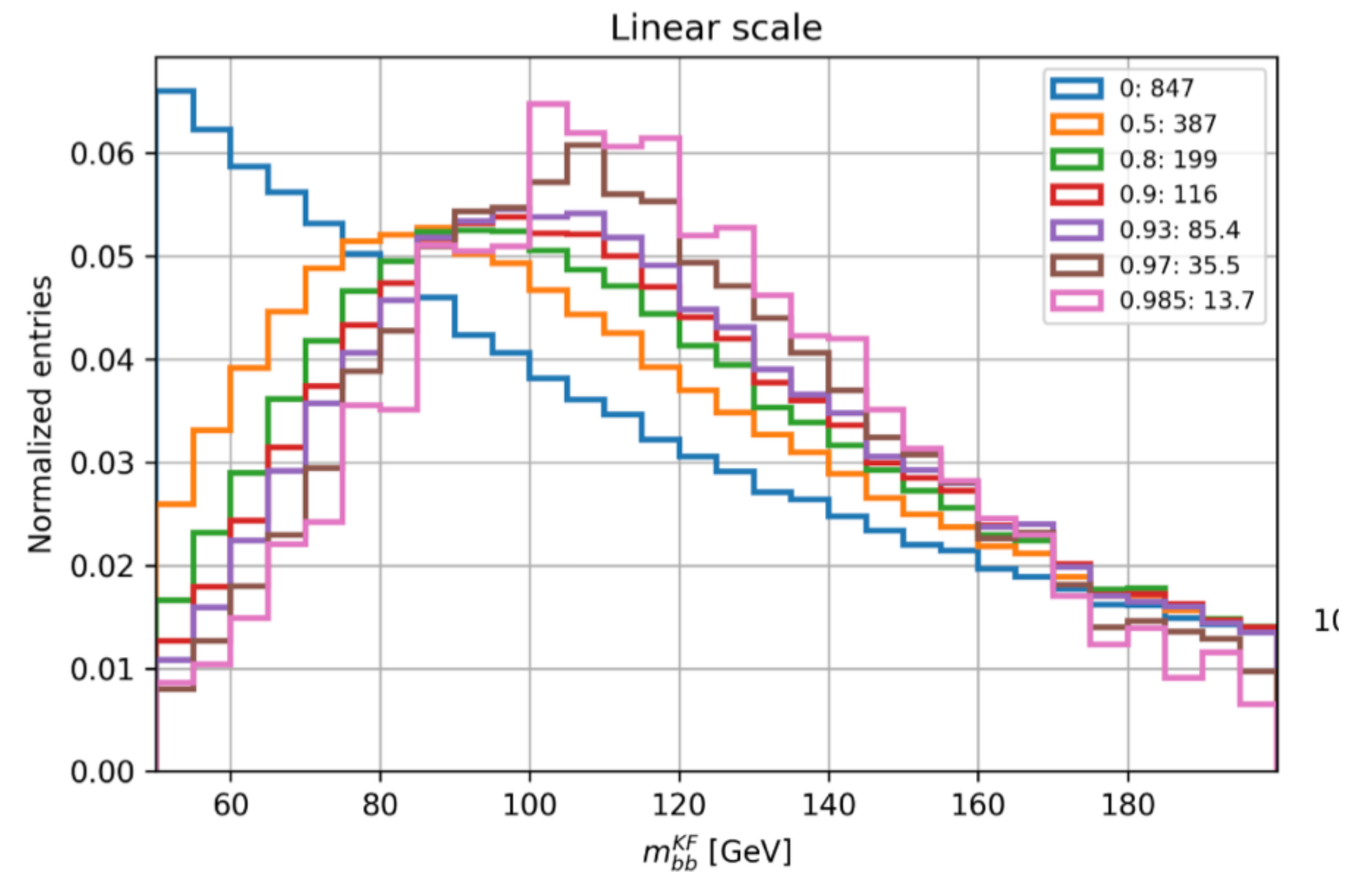


$HH \rightarrow b\bar{b}\gamma\gamma$ transformer and toys

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- I designed a model discriminating HH from backgrounds while
- Not sculpting myy of yyjets
- Not sculpting mbb of yyjets
- Here focus on High Mass region providing SM significance, LM currently is 1D
- On the plot: yybb sample mbb distribution for increasingly tight cut sof non-mbb-decorrrelated model

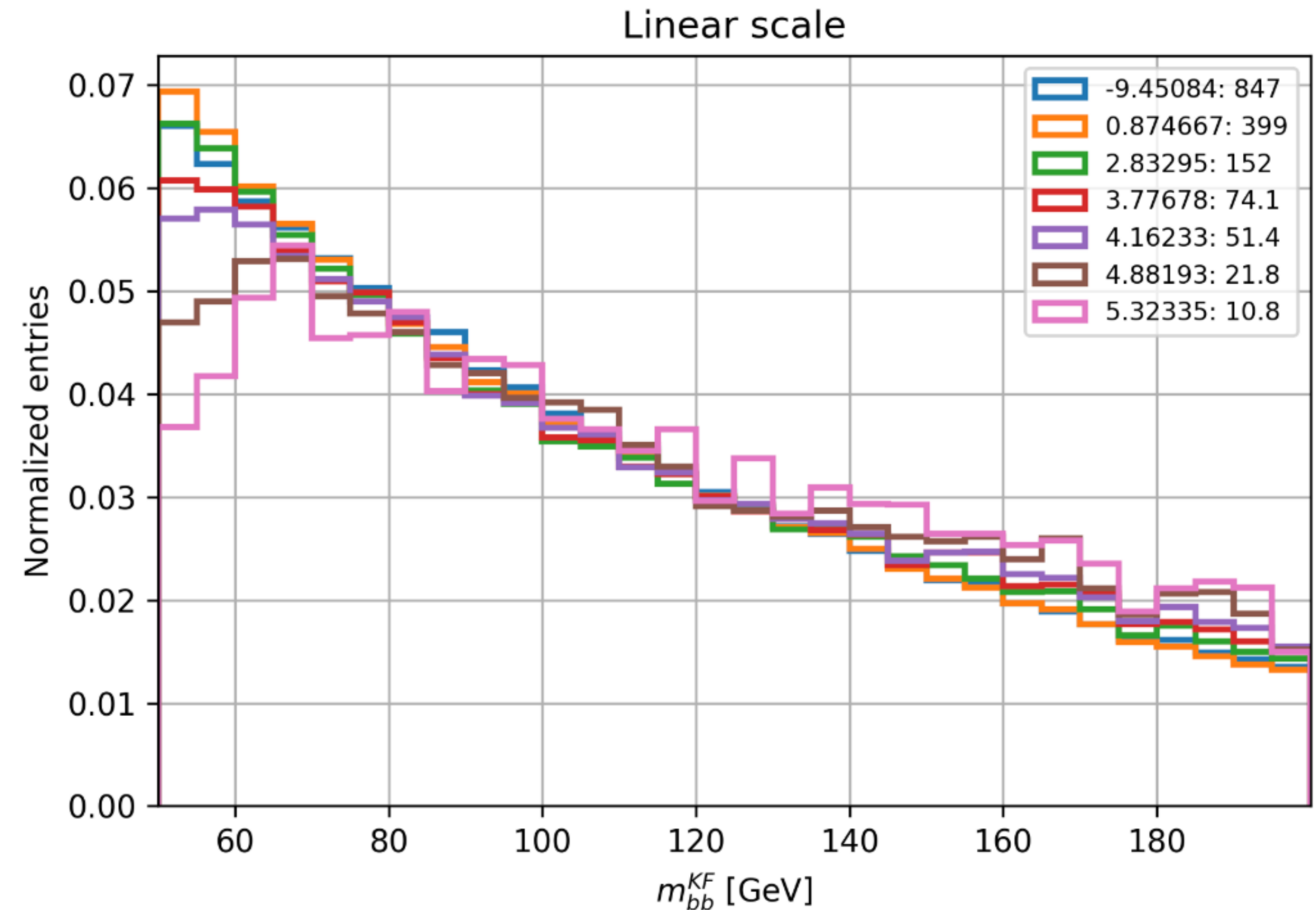


Potential ways to achieve decorrelation

- 1) Distance or other type of decorrelation added, adversarial network, ...
 - Additional component in the network
 - With additional parameters which are to be tuned
 - Overall intended to clean up accidental leakage of info, not to directly fight with the most important variable in 1D which is mbb
 - Tested and getting similar results to 2)
- 2) Usage of variables which are not sculpting
 - Either works or doesn't
 - Simpler and more robust

Decorrelation step 1,2

- Step 1 - remove all dRbb, pTbb, mbb and their proxies
- Step 2 - train only in region where variation of mbb is less
- Can't pick up correlation if it's not there in training
- Pick 100-150 GeV as region with HH mbb core



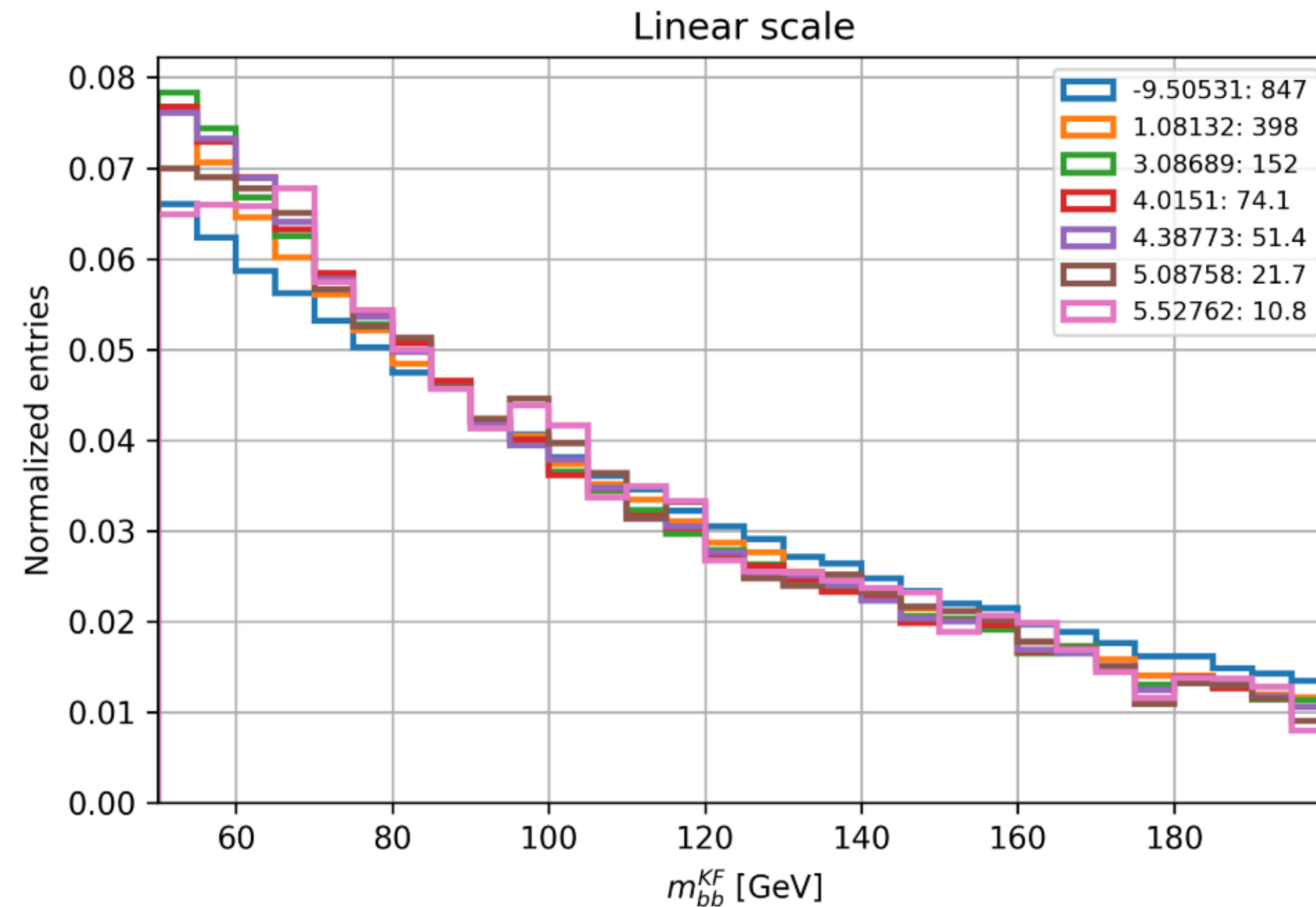
Decorrelation step3

🌐 Step 3 - as to make mbb need two jets dividing only of jets pt by mbb should enough

🌐 Divide second jet and keep information from 1st jet for discrimination

🌐 Safe to fit from 70, depending on where tightest category lands can fit from 80

🌐 Decorrelation achieved



Discriminants involved

🌐 Network provides seven “probabilities”

🌐 p_{VBFHH}, p_{ggFHH} for signal

🌐 $p_{ggF_bbH}, p_{ttH}, p_{ZH}$ for major resonant backgrounds

🌐 P_{yyjets} for $yycont$

🌐 HH-oriented ways of using it

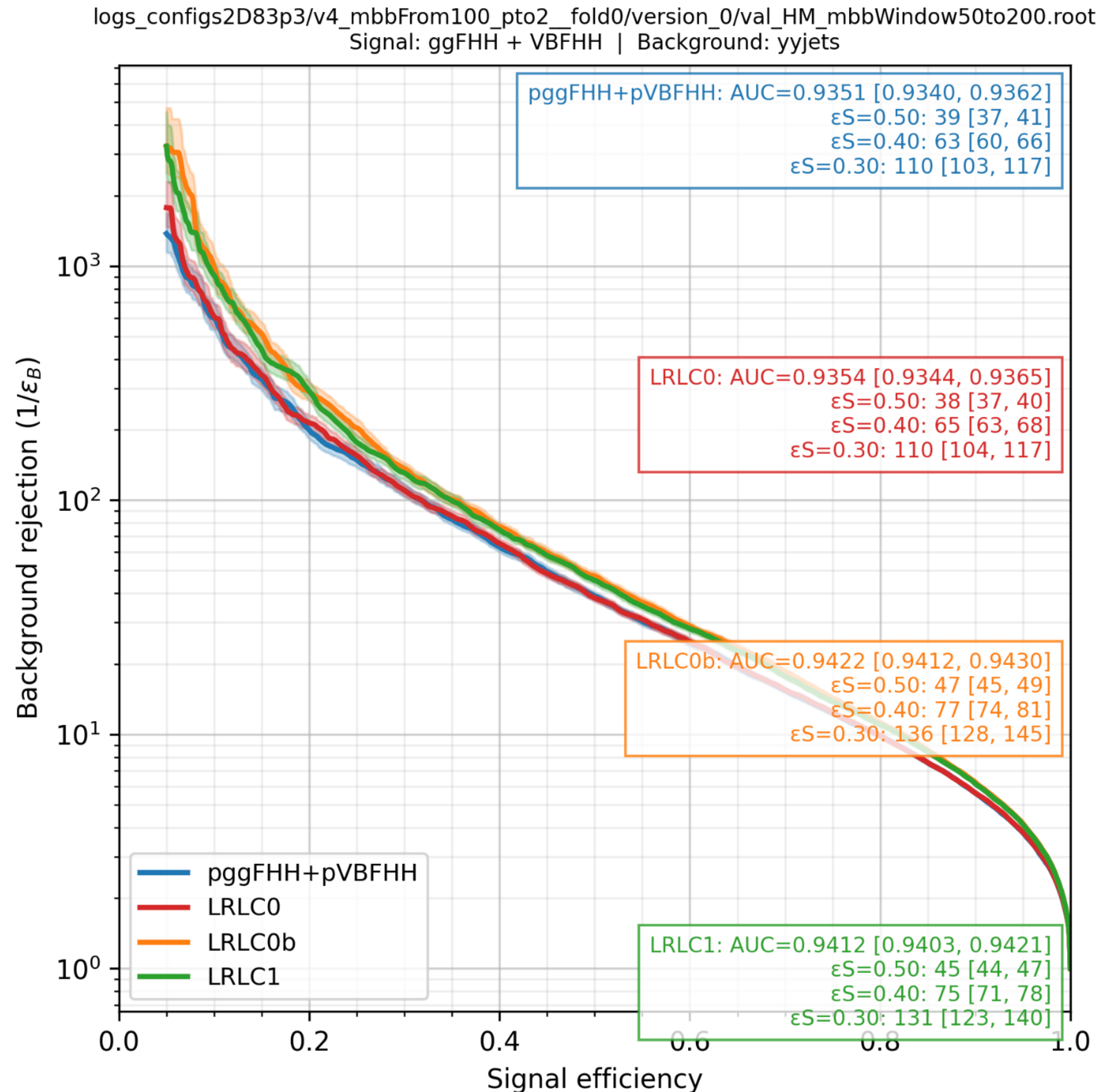
🌐 $p_{ggFHH} + p_{VBFHH}$ is equivalent of binary score

🌐
$$LRLC1 = \log \left(\frac{0.95 \cdot p_{ggFHH} + 0.05 \cdot p_{VBFHH}}{0.65 \cdot p_{yyjets} + 0.11 \cdot p_{ZH} + 0.11 \cdot p_{ttH} + 0.13 \cdot p_{ggFbbH} + 0 \cdot p_{otherSH}} \right) \text{ or}$$

other tunable fractions

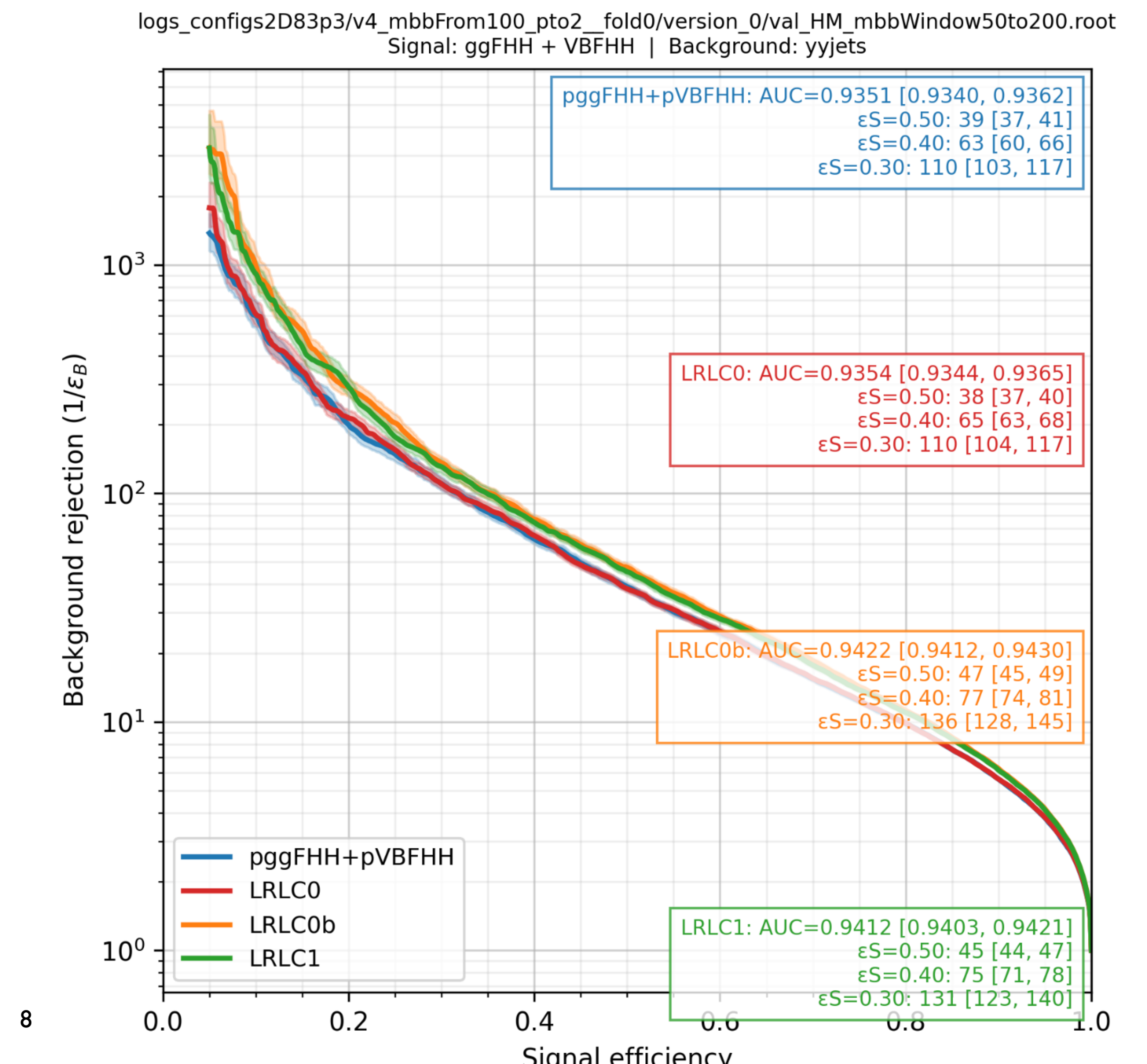
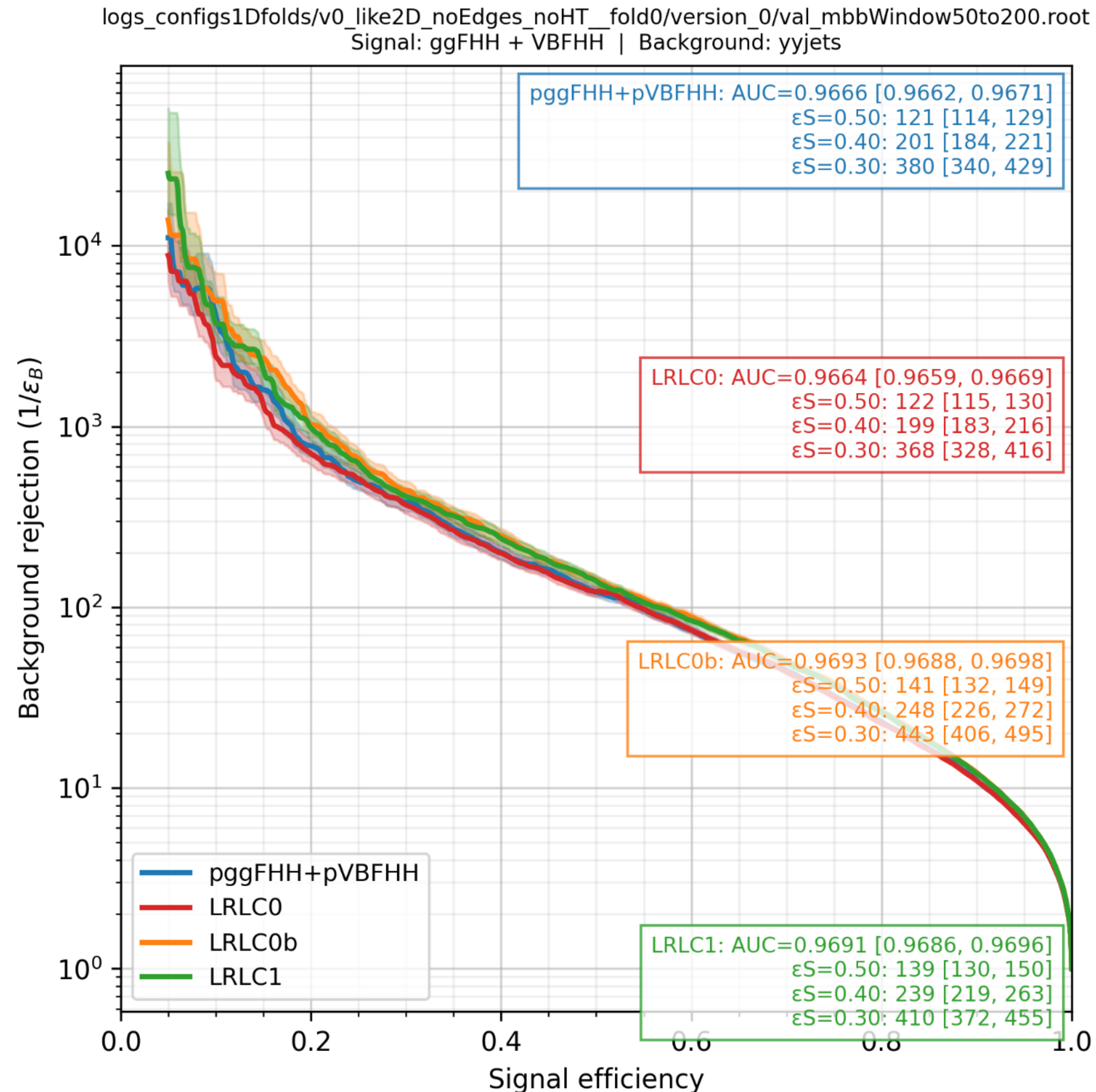
Discriminants involved and 2D ROC

- Evaluated in 50-200 mbb window
- For discriminant rejections, written for reference signal efficiencies (previous round tightest category 35%)
- Tuned discriminant provides better rejection



1D and 2D ROC

🌐 Better in 1D but not appropriate metric as it doesn't contain myy-mbb granularity



Determining boundaries of categories

 In 1D

 H/HH yield in 120-130 window

 Yycont yield in 105-160 window and multiplied by 10/55 - way to stabilize as this sample is jumpy in category that matters the most

 Take category edges that best give $S/\sqrt{B}$

 In 2D

 Necessarily done on myy-mbb grid

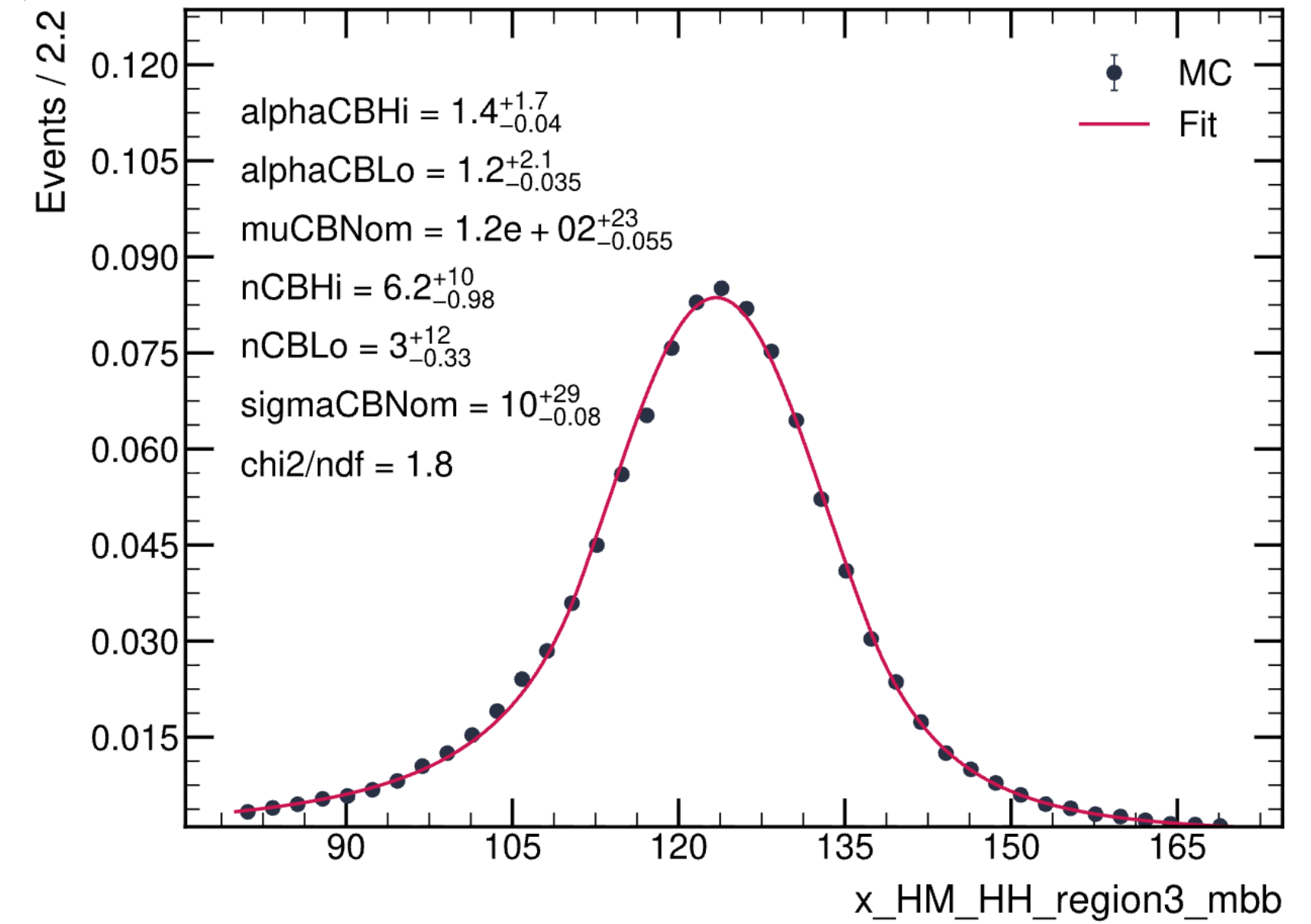
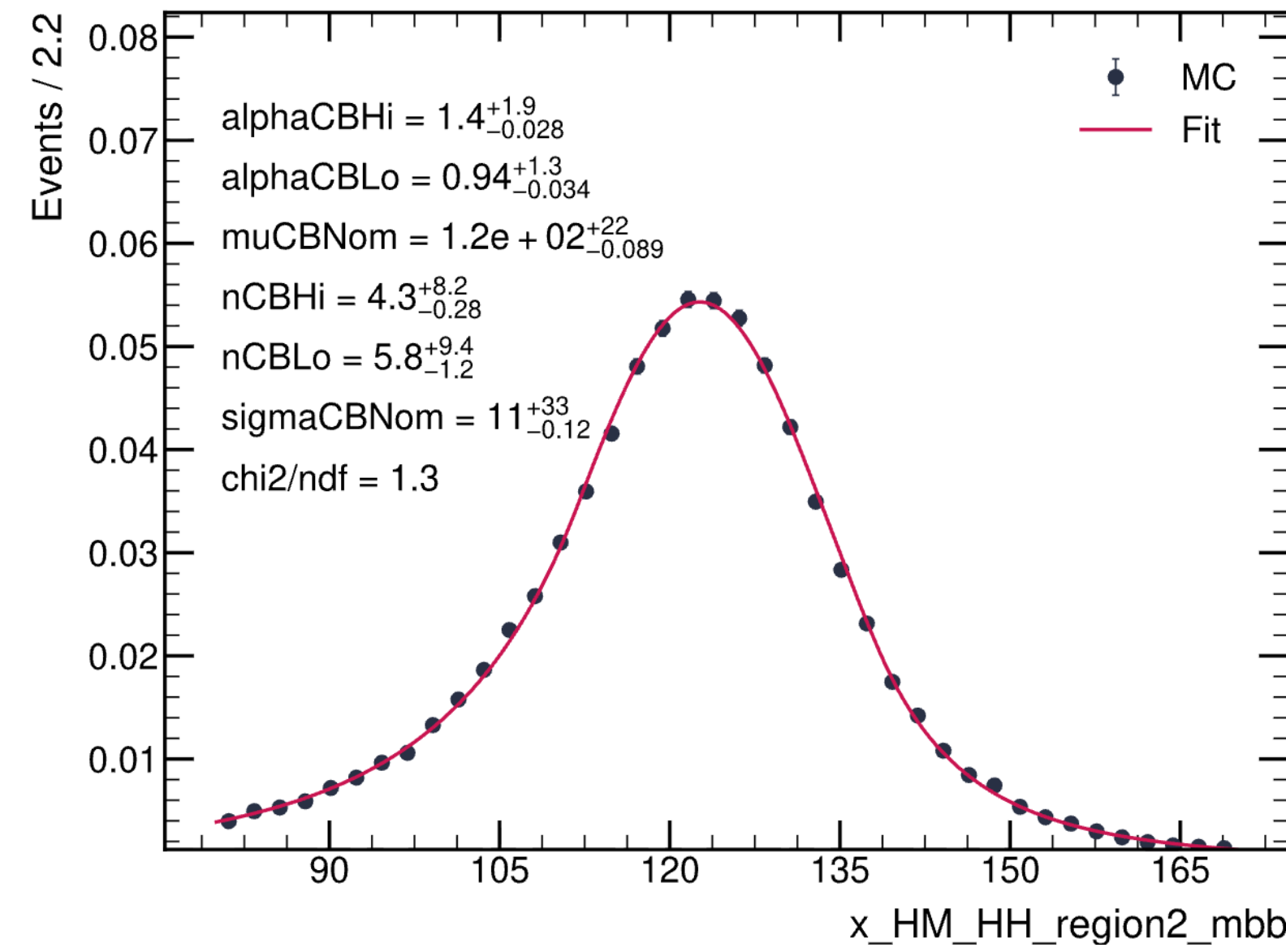
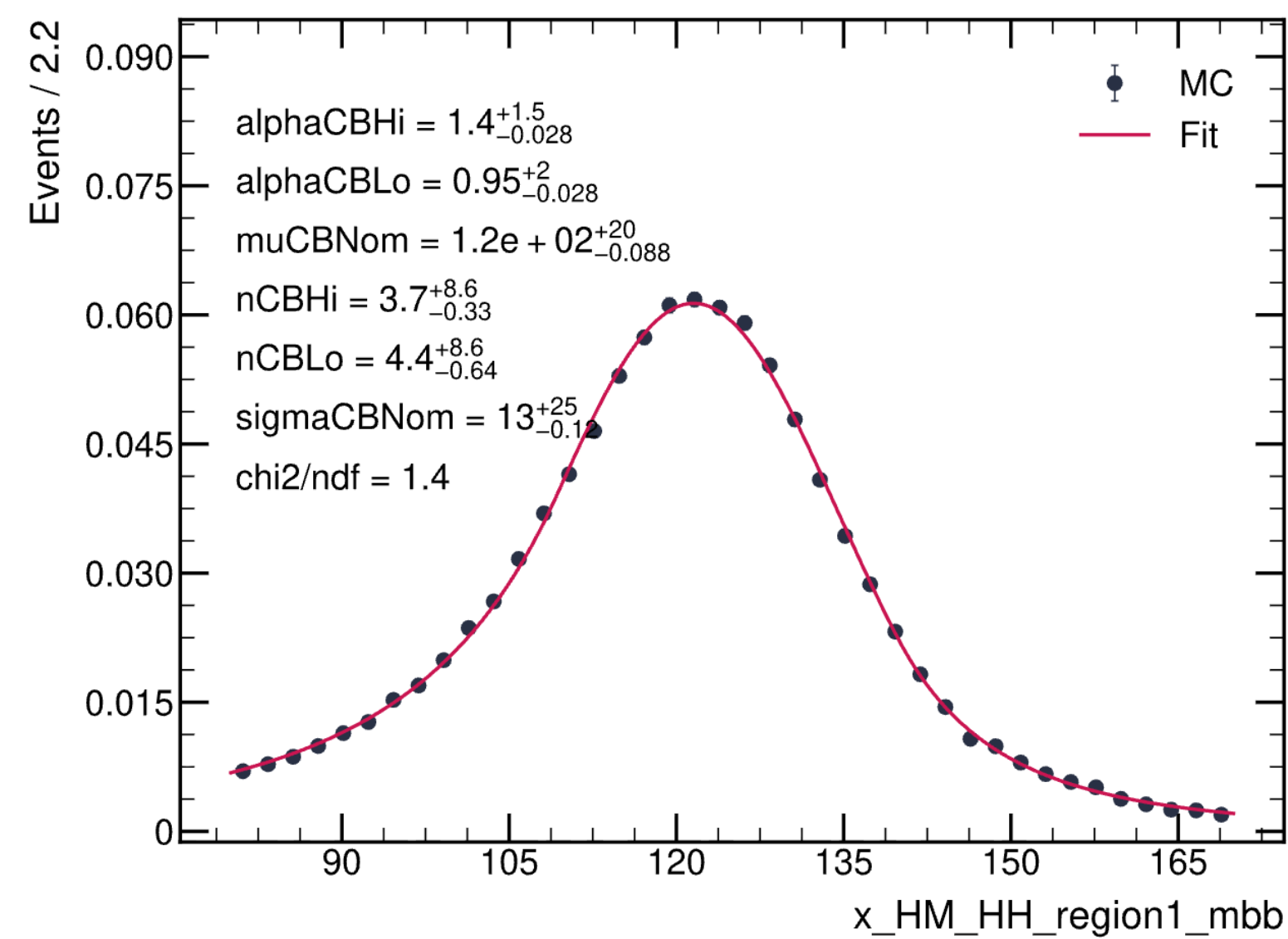
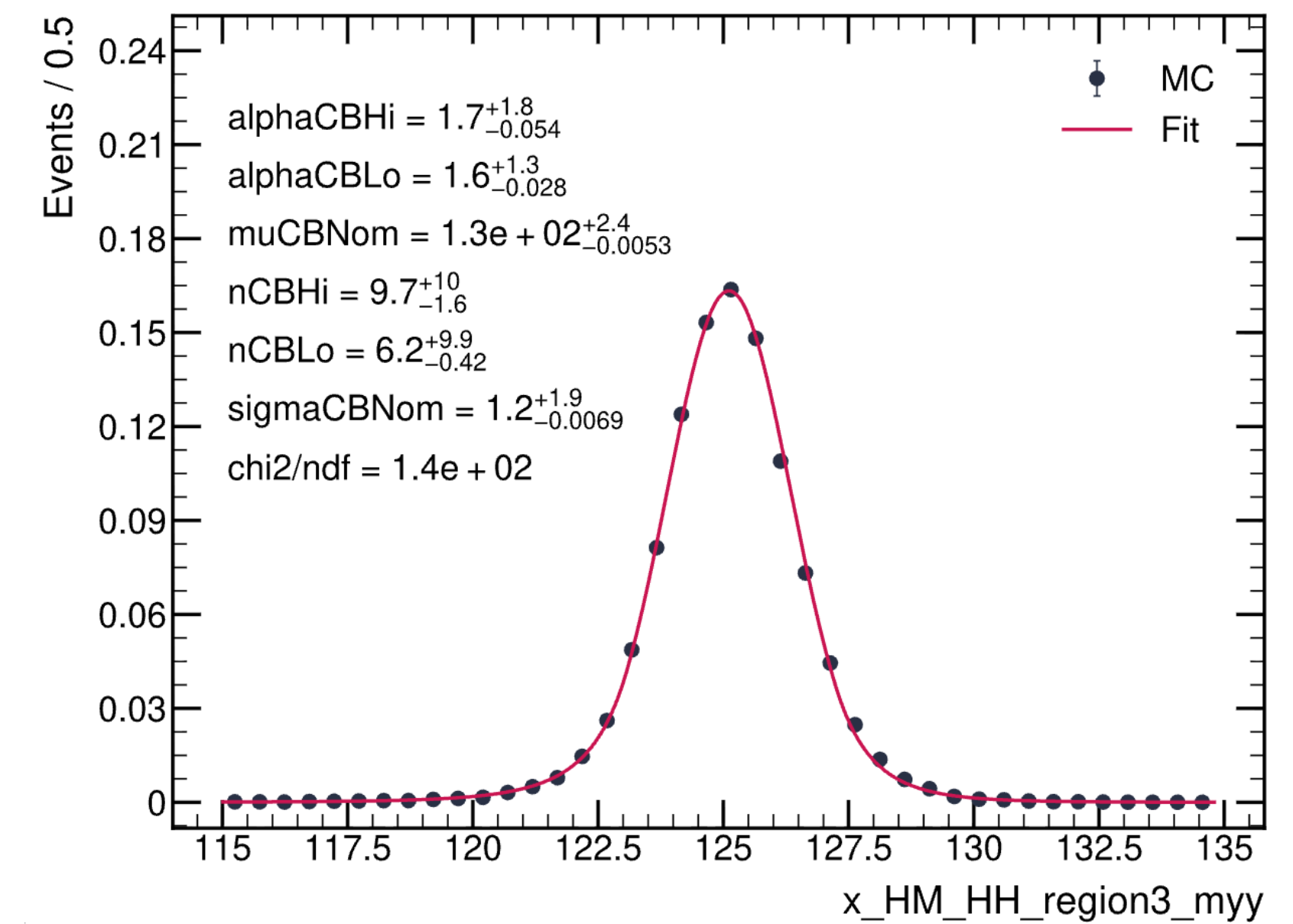
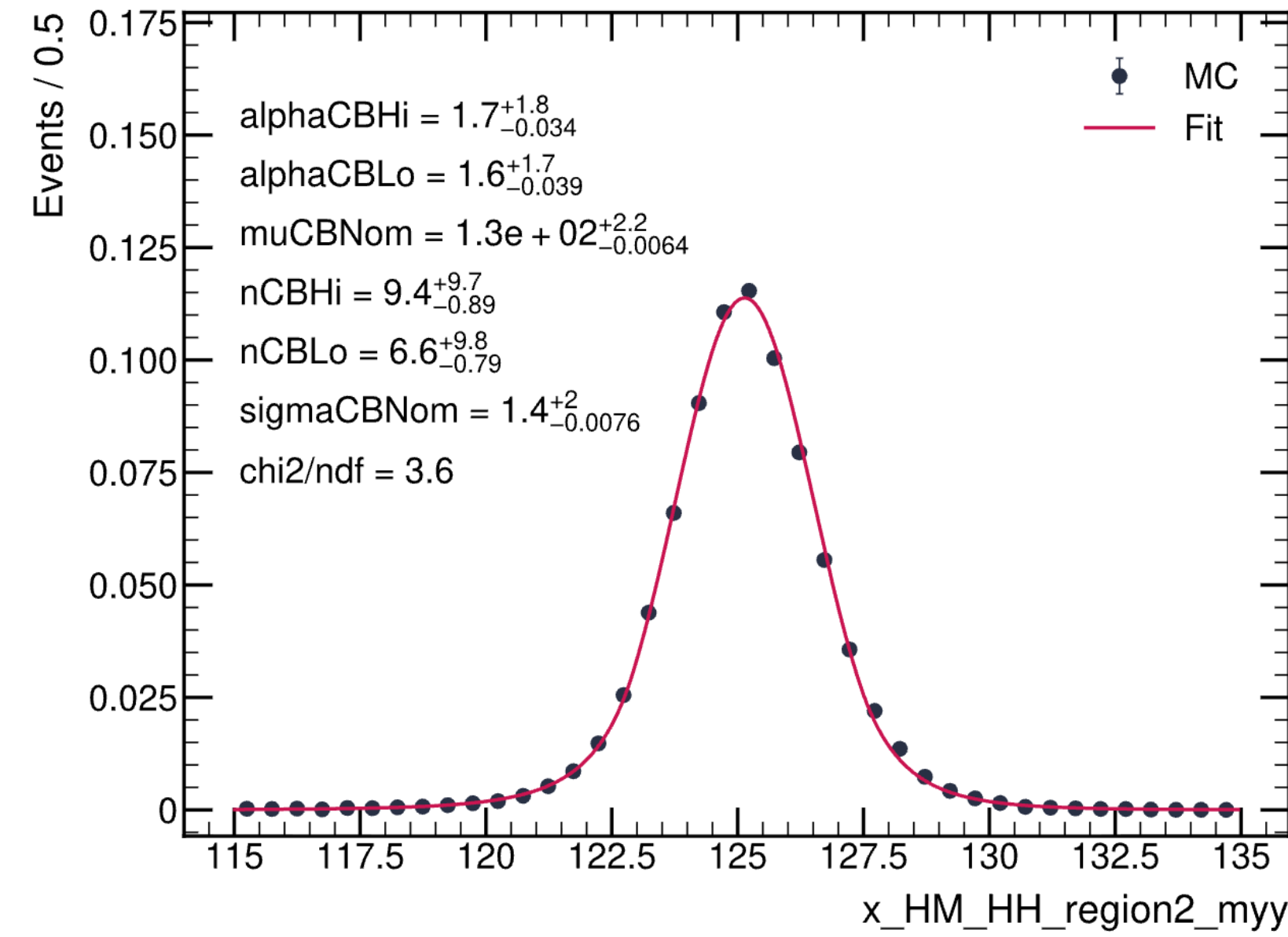
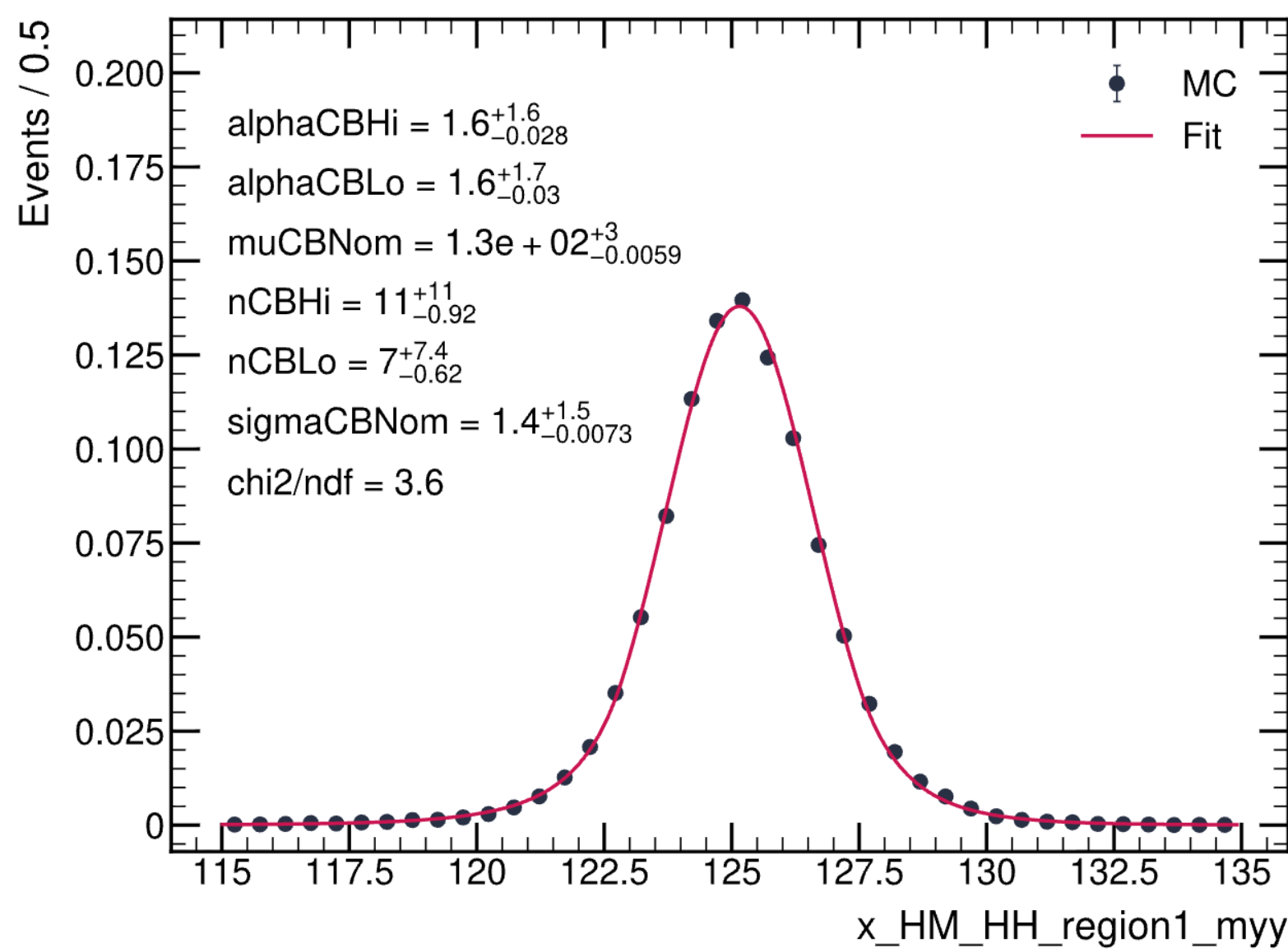
 Stabilization instead is running 300 toys and taking boundary from toys which is closest to median

 In both cases require min sumw 11 for yyjets

From category boundaries to stat results

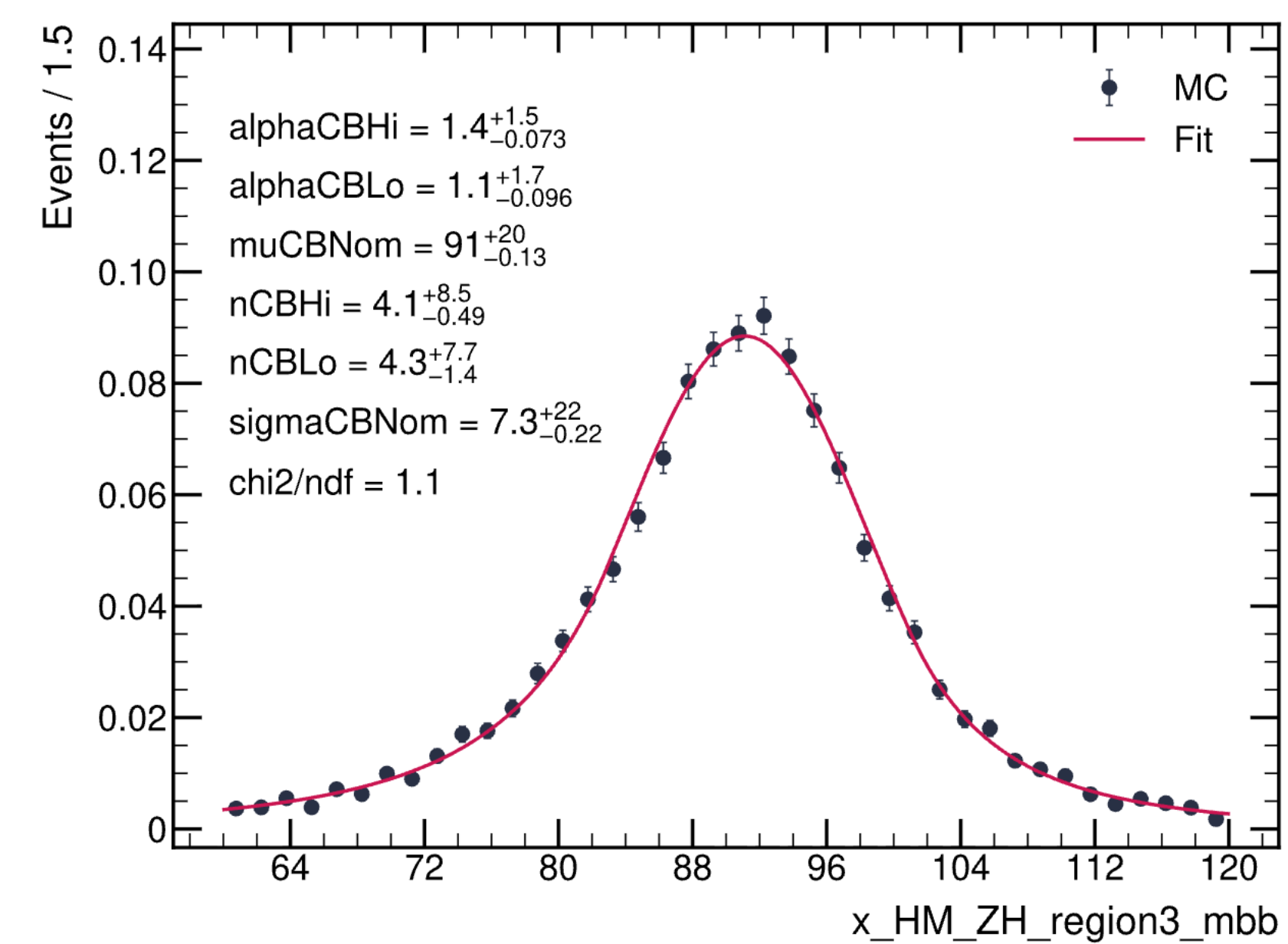
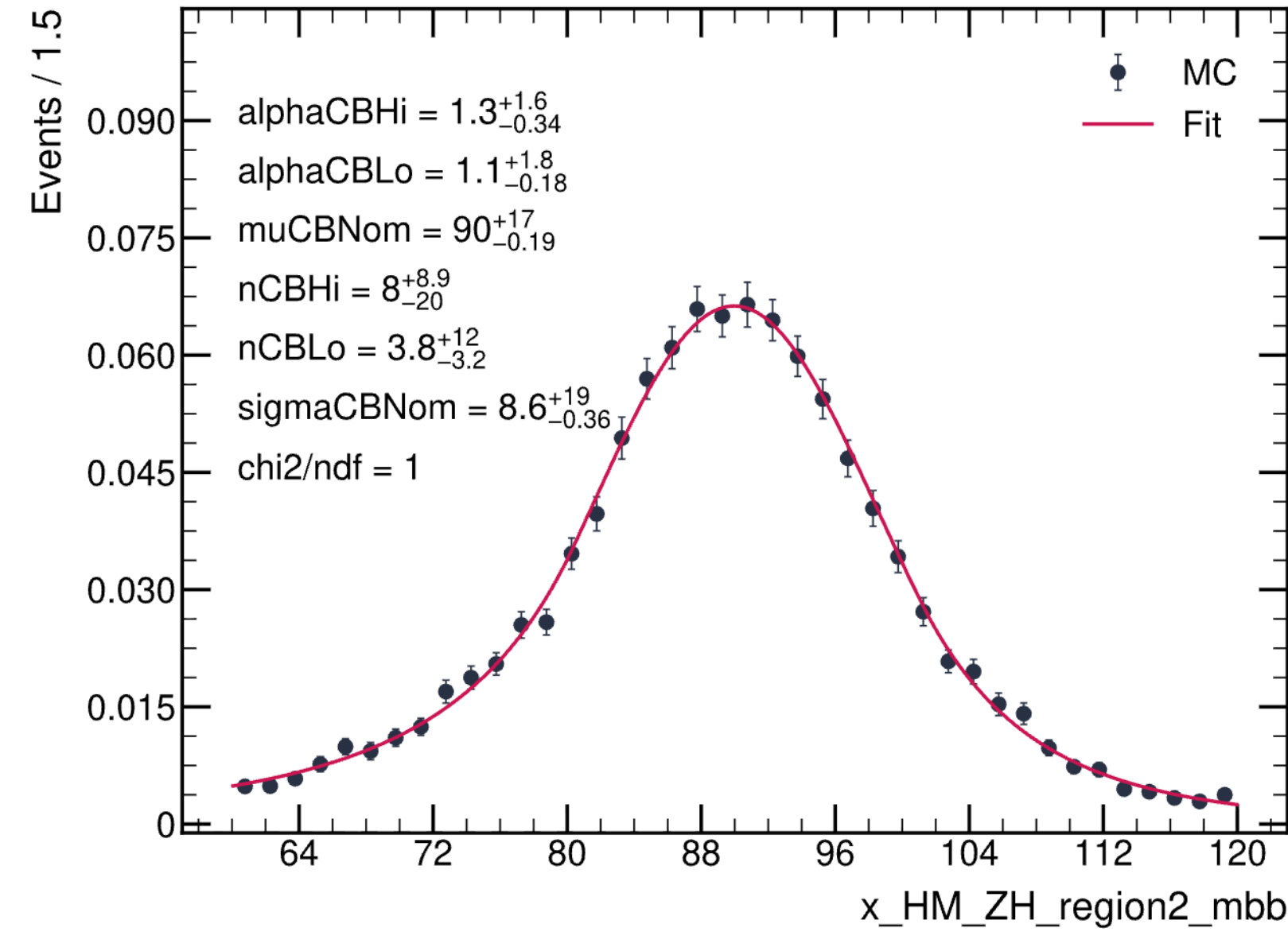
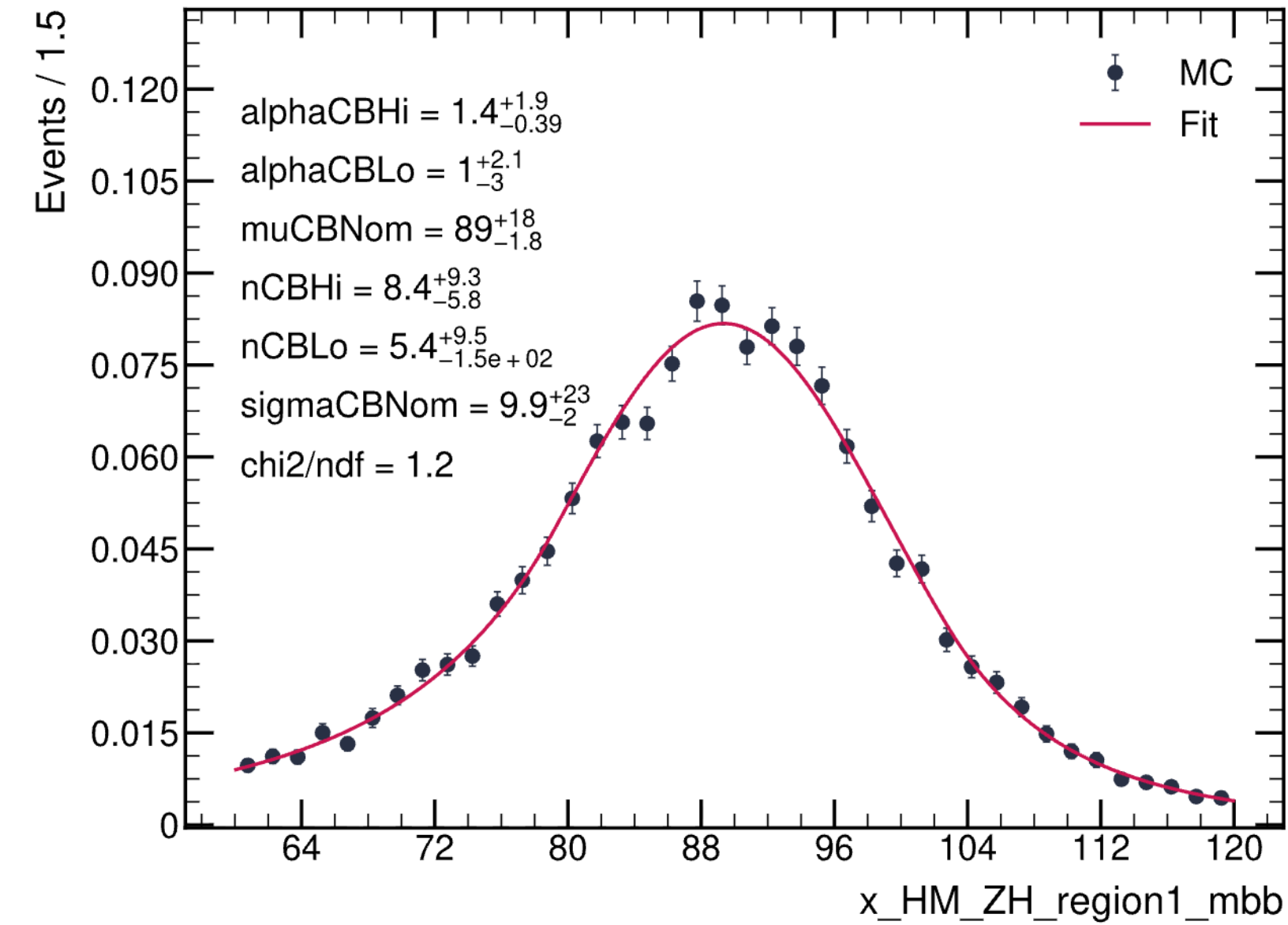


HH modelled with DCSB(myy)xDCSB(mbb)



From category boundaries to stat results

🌐 DCSB(mbb) for ZH



Results and stability with data ≤ 24

- 🌐 After training model, I prepared ntuples and workspace for others to look at
- 🌐 HH SM 1.3sigma (while 1D gives 1.102) from blinded stat-only fit
- 🌐 1.268 with toys (while 1D gives 1.109)
- 🌐 Unlike 1D, 2D fit allows simultaneous fit with ZH (validation of the analysis) without degrading HH results
- 🌐 Stability of fit is also investigated with software I wrote
 - 🌐 9934 converged for 1D model
 - 🌐 9935 converged for 2D model

Results with data $\leq$ 25

- 🌐 Data 25 is available, but corresponding MC is not
- 🌐 Run-3 MC is scaled by 1.69 to emulate

	μ_{HH}	μ_{ZH}	μ_H	HH expected significance	ZH expected significance
Fit with μ_{HH} floating, and $\mu_{ZH}=1$ and $\mu_H=1$	1 +/- (-0.728482, 0.97897)	-	-	1.5165091	-
Fit with (μ_{HH} , μ_{ZH}) floating and $\mu_H=1$	1 +/- (-0.739566, 0.9993)	1.00001 +/- (-0.67483, 0.857299)	-	1.4870104	1.6287418
Fit with (μ_{HH} , μ_H) floating and $\mu_{ZH}=1$	1 +/- (-0.780662, 1.05001)	-	0.999998 +/- (-0.506505, 0.589593)	1.3747727	-

The End