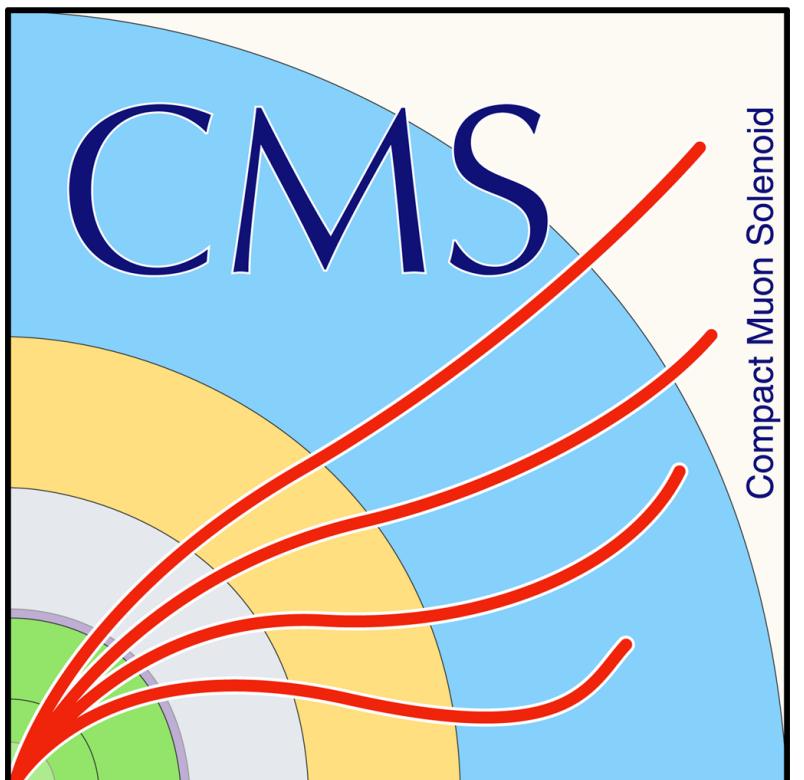


# Measurement of Higgs boson cross sections with $H \rightarrow \gamma\gamma$ decays at CMS

Johannes Erdmann  
on behalf of the CMS Collaboration

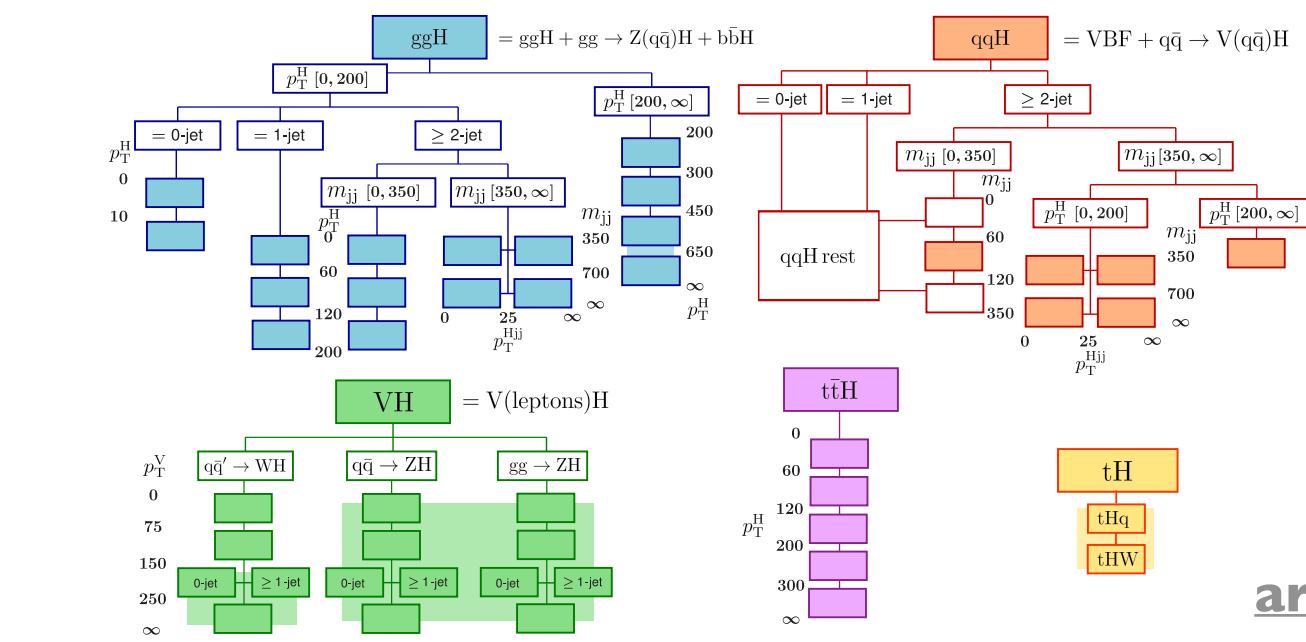
DIS2024  
Grenoble  
April 9, 2024



**RWTH**AACHEN  
UNIVERSITY

# Introduction

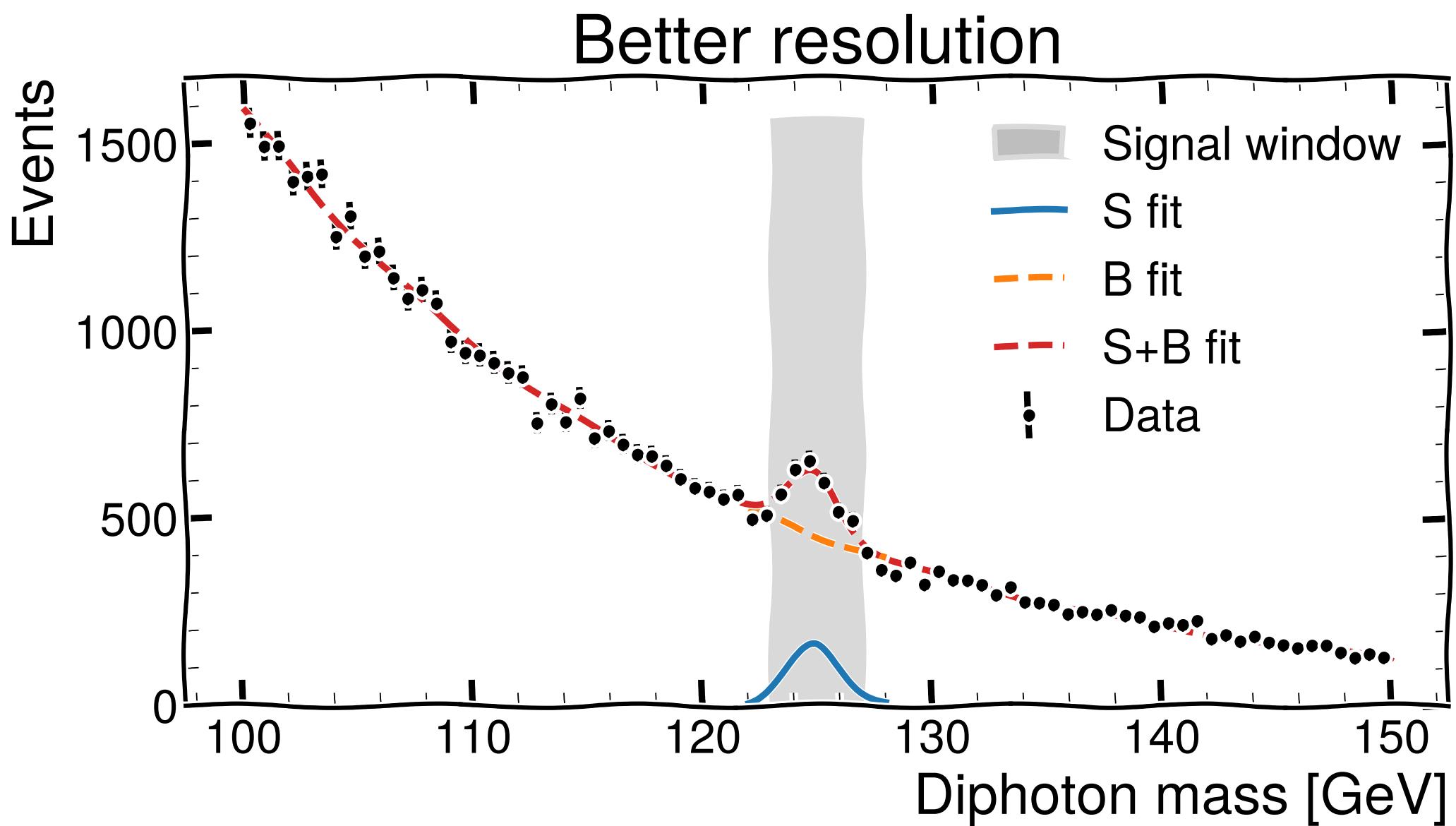
- Cross section measurements essential for pinning down Higgs properties
- Simplified template cross sections
  - optimized for reduced theory uncertainties
  - per production mode
- Fiducial and differential cross sections (this talk)
  - designed to maximize model independence
  - inclusive in production modes
- Full Run-2 measurement [arXiv:2208.12279](#) [JHEP 07 \(2023\) 091](#)



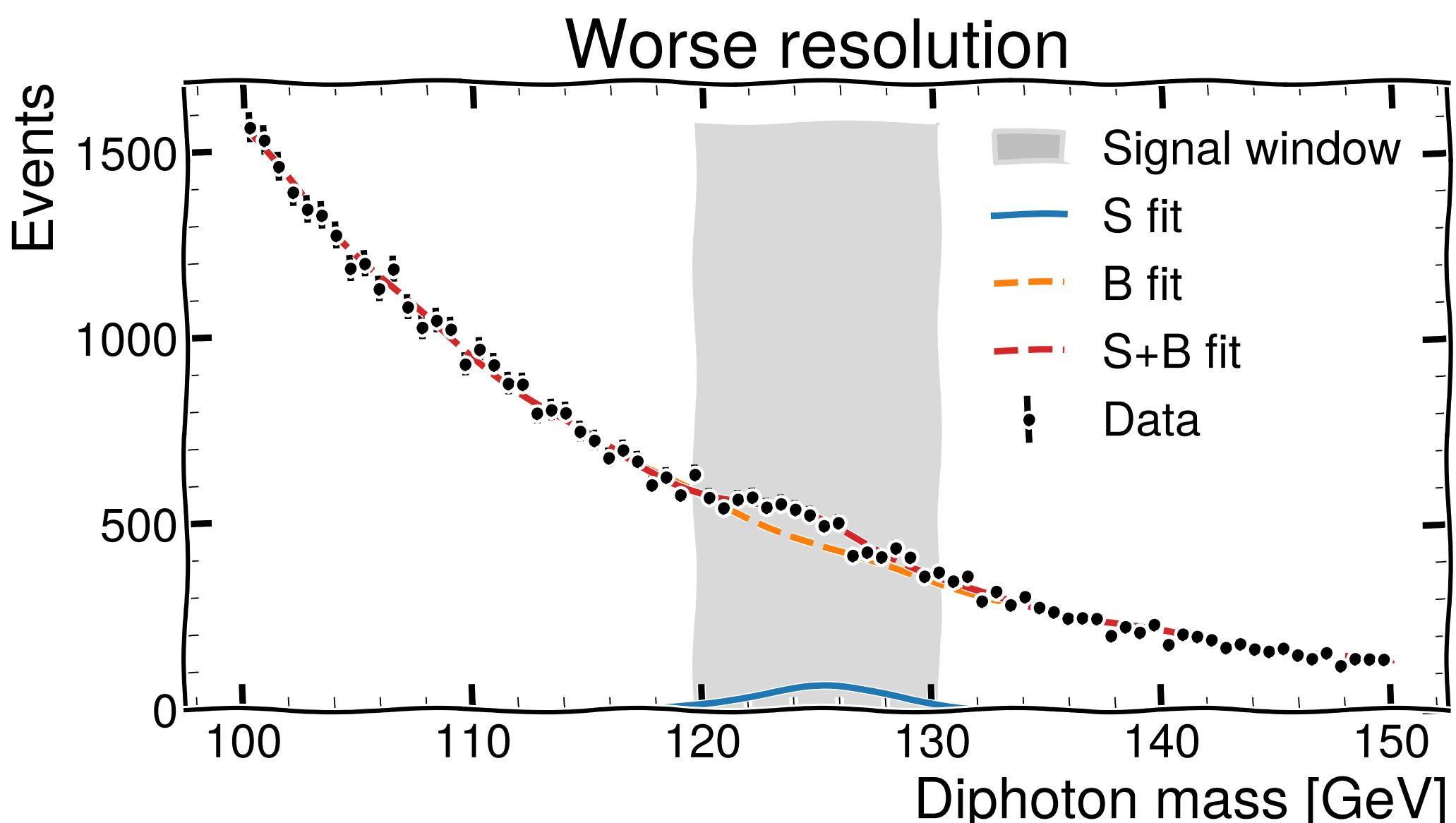
[arXiv:2103.06956](#)

# Analysis Strategy

- Fiducial phase spaces  $\sim$  reco. phase spaces
  - Inclusive
  - Motivated by production modes ( $\ell, E_T^{\text{miss}}, b \dots$ )
  - Split into bins ( $p_T^{\gamma\gamma}, \# \text{jets}, p_T^{\text{jet}}, \dots$ )



- Signal yields from fits to  $m_{\gamma\gamma}$
- Additional categorization  $\rightarrow$  model independent!
- Split only by:
  - $m_{\gamma\gamma}$  resolution
  - data-taking year

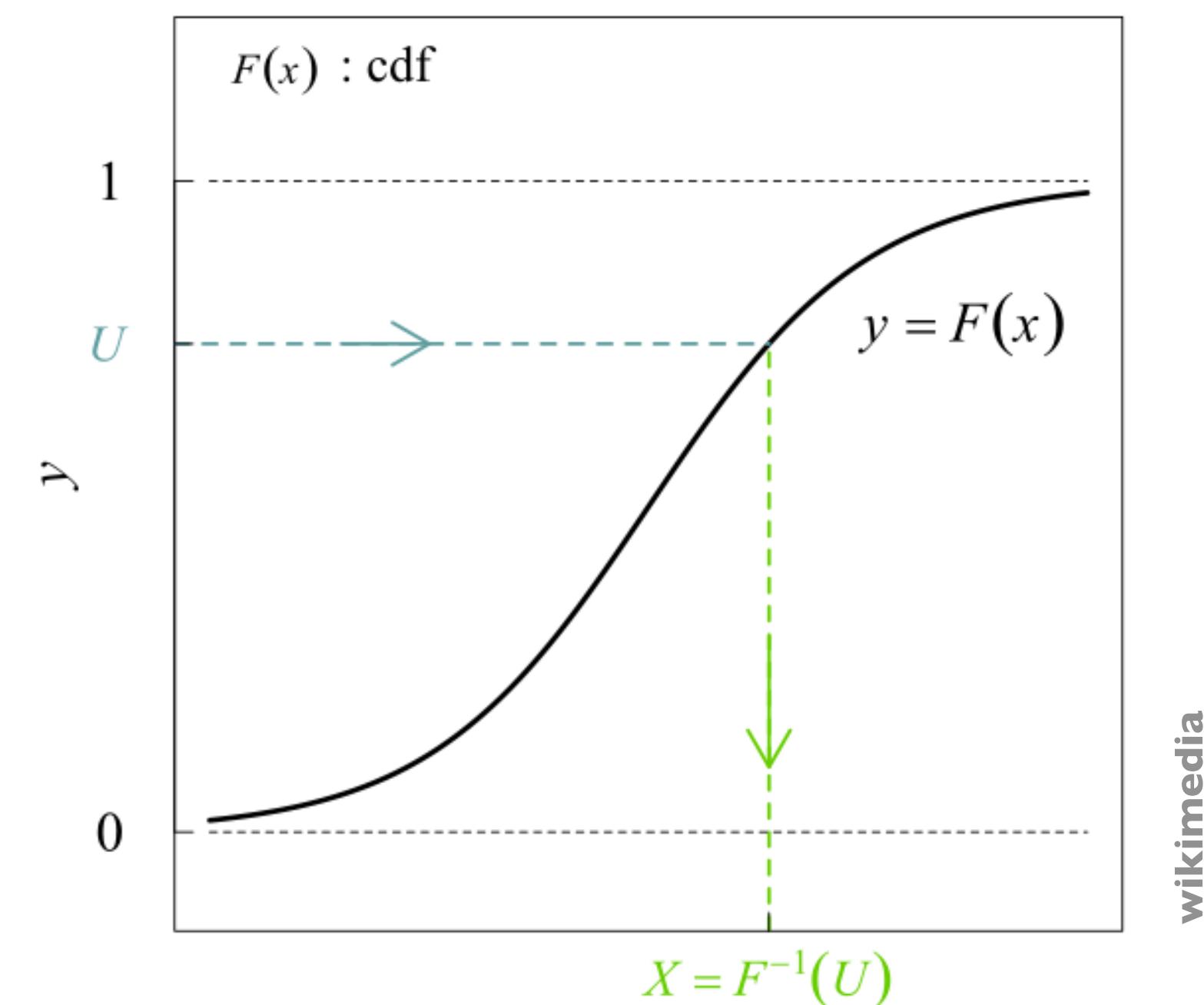


# Quantile Morphing

- Energy resolution estimated by “semi-parametric regression”
- As resolution of calorimeters improves with energy  
→  $\frac{\sigma_{m_{\gamma\gamma}}}{m_{\gamma\gamma}}$  improves with energy
- Produces turn-ons in  $m_{\gamma\gamma}$  spectrum...

- Solution: decorrelate vs.  $m_{\gamma\gamma}$  by quantile morphing to  $m_{\gamma\gamma} = 125 \text{ GeV}$ 
  - aka Smirnov transform
  - aka inverse transform sampling

$$y_{\text{corr}} = F_Y'^{-1}(F_Y(y))$$



wikimedia

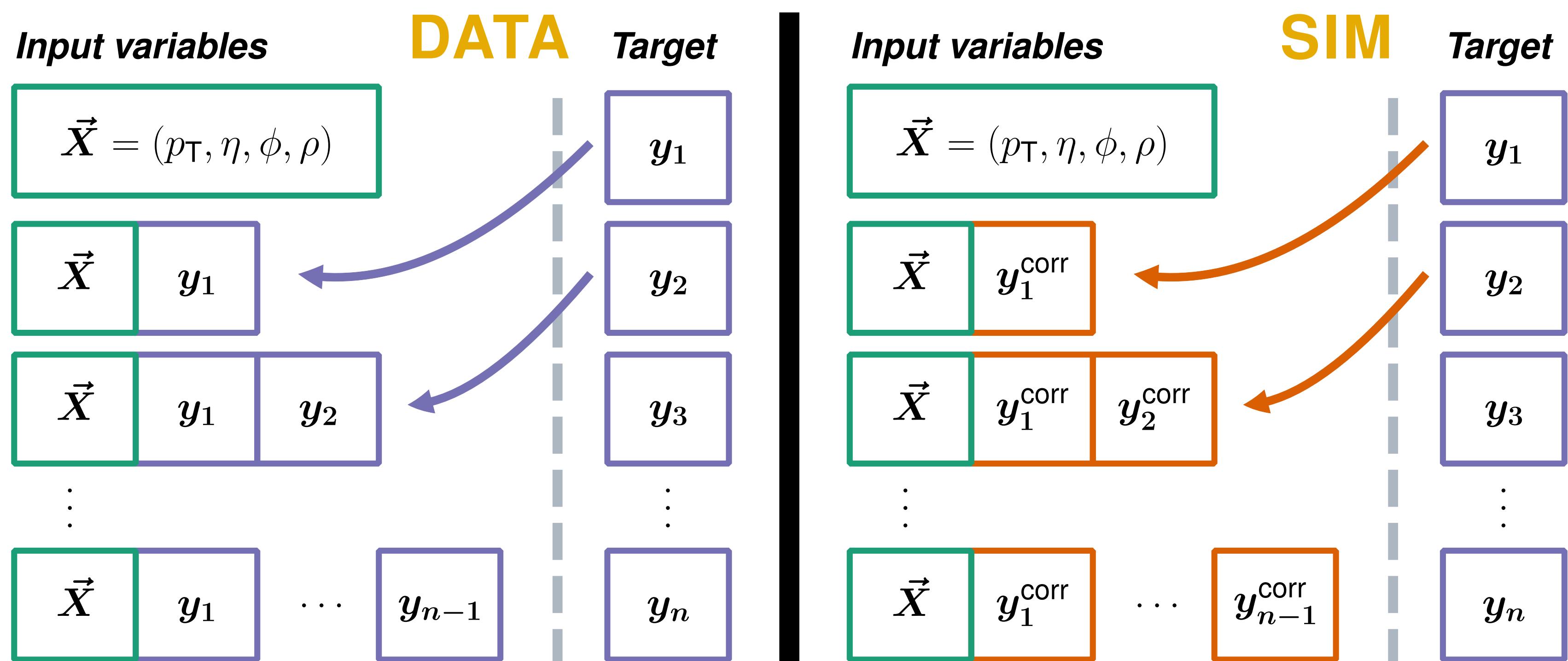
# Chained Quantile Regression

- Conditional quantile morphing → quantile regression with loss

$$q_Y(\tau) = \underset{u}{\operatorname{argmin}} \left\{ (\tau - 1) \int_{-\infty}^u (y - u) dF_Y(y) + \tau \int_u^\infty (y - u) dF_Y(y) \right\}$$

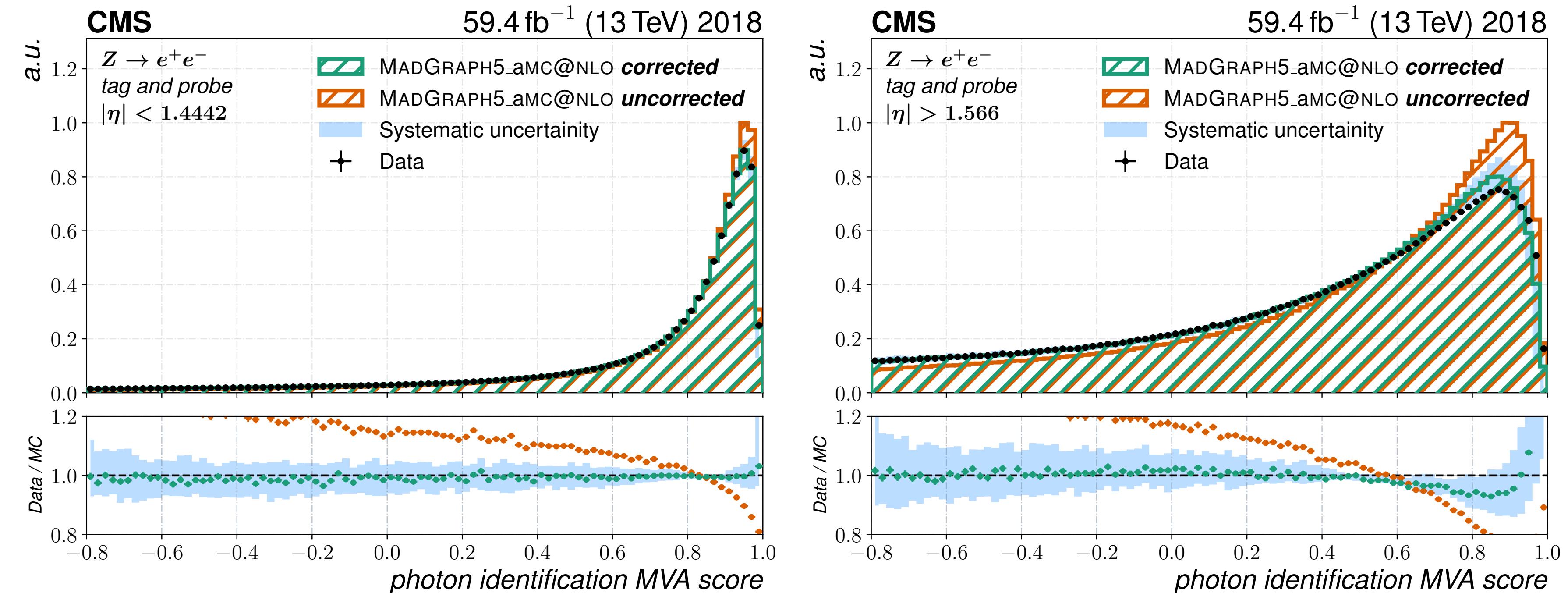
- Extension to several correlated variables

→ chained quantile regression:



# Chained Quantile Regression

- Used to correct modeling of input features to photon identification BDT



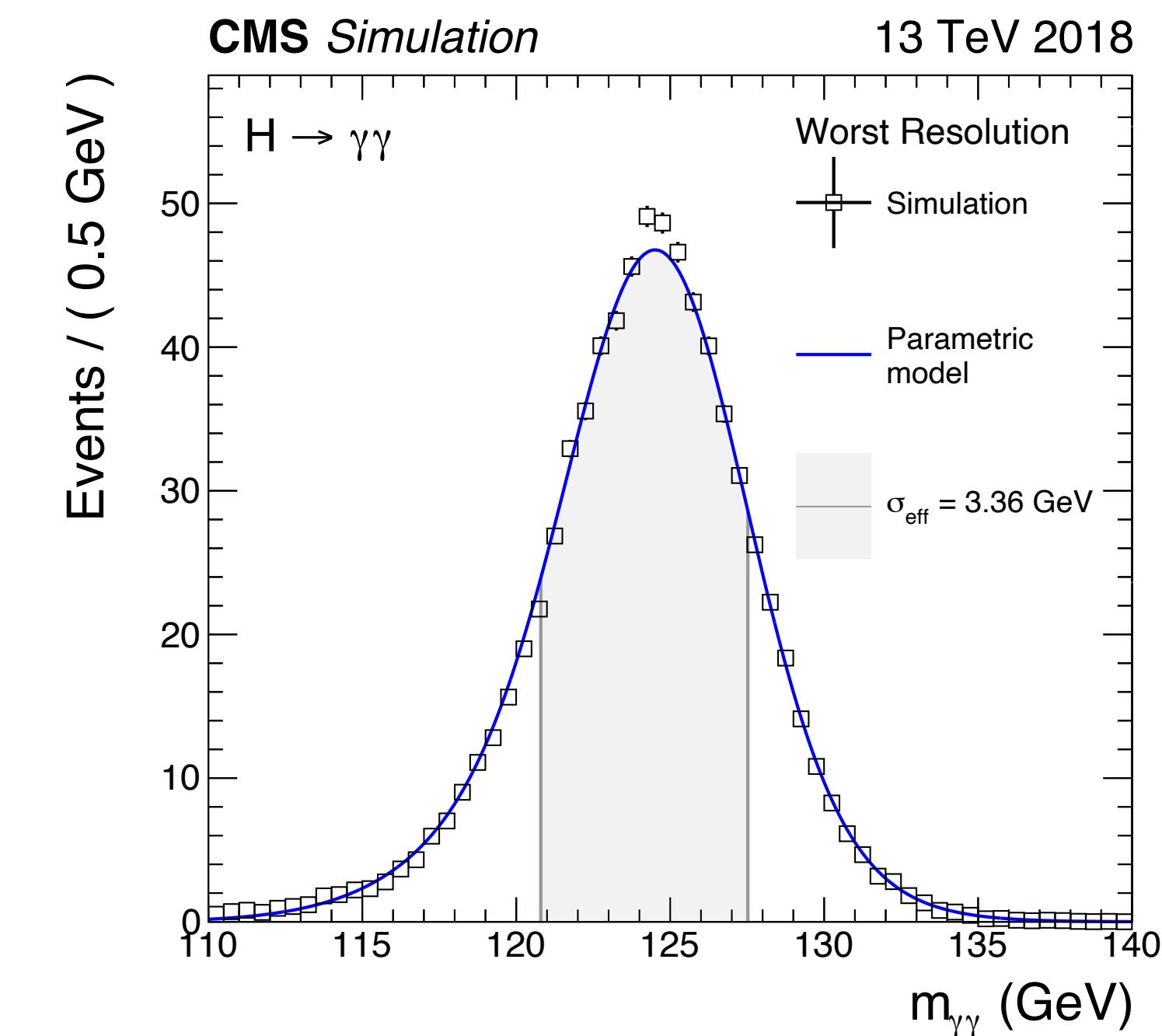
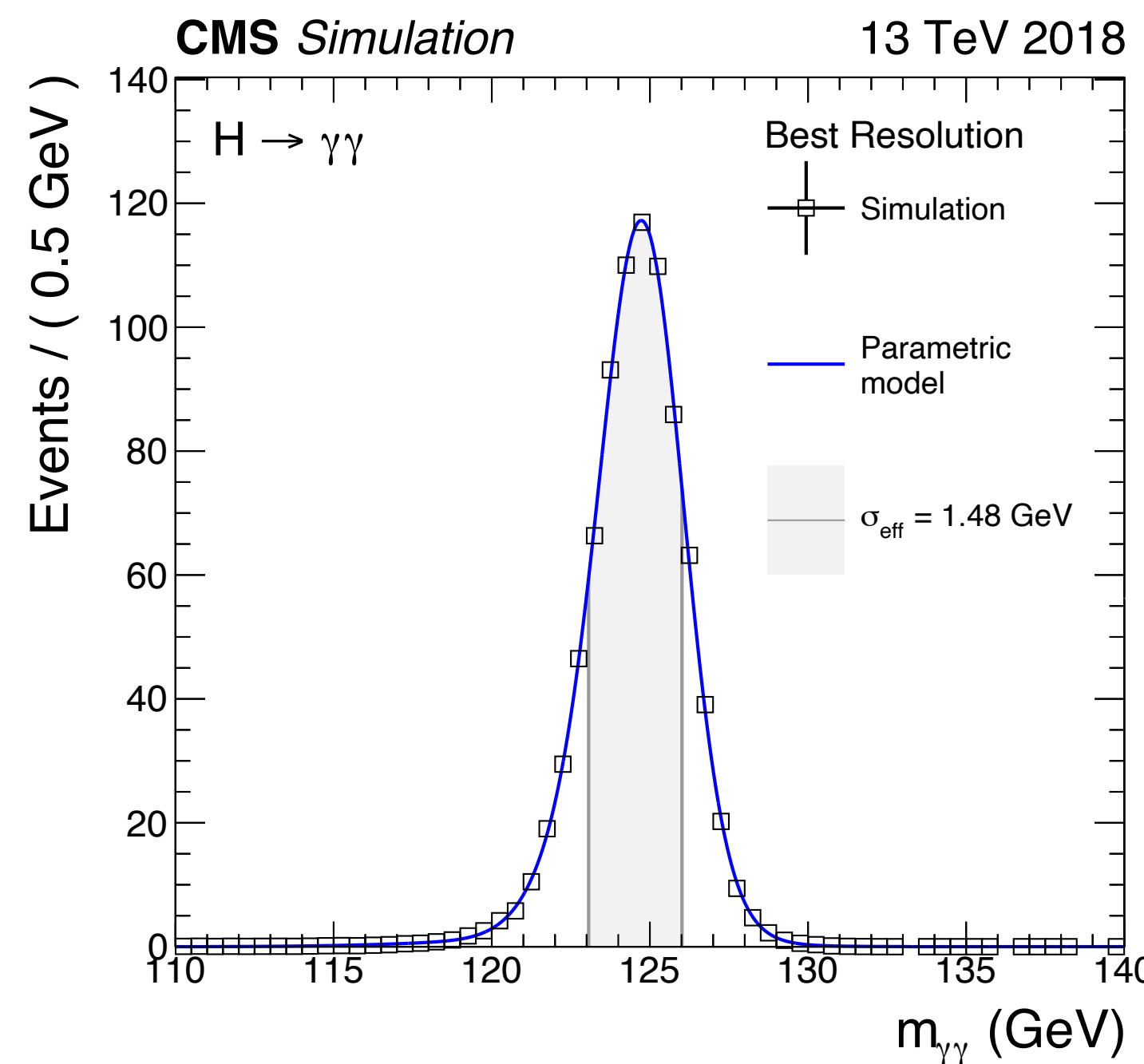
- CQR tedious: one BDT per quantile and per variable...
- Morphing proposals using deep learning: [2107.08648 \(OT\)](#) [2309.15912 \(CQR w/ NF\)](#)

[2304.14963 \(NF for data\)](#) [2309.06472 \(2-3 NFs\)](#) [2403.18582 \(1 NF + switch\)](#)

# Selections and Categorisation

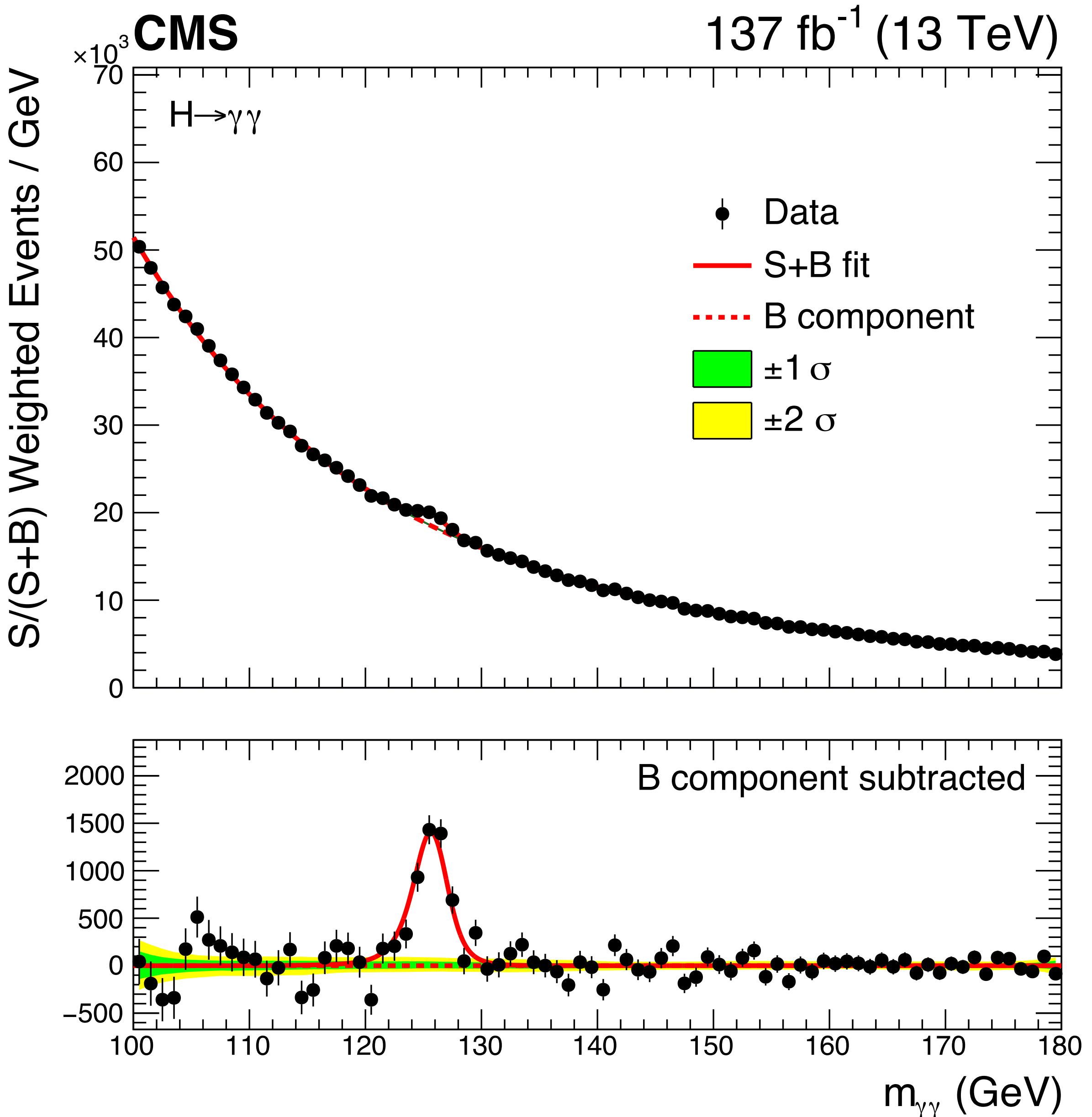
	Reco Level	Particle Level
$p_T^{\gamma 1} / m_{\gamma\gamma}$	$> 1/3$	$> 1/3$
$p_T^{\gamma 2} / m_{\gamma\gamma}$	$> 1/4$	$> 1/4$
ID	minimum cut on ID BDT score	$\text{Iso}_{\text{gen}}^{\gamma} < 10 \text{ GeV}$
$ \eta^{\gamma} $	$< 2.5$ and not in $[1.4442, 1.566]$	$< 2.5$

- 3 resolution categories  
+ optimized cut on the photon ID BDT score  
(with efficiency from 63.5% to 90.4%)

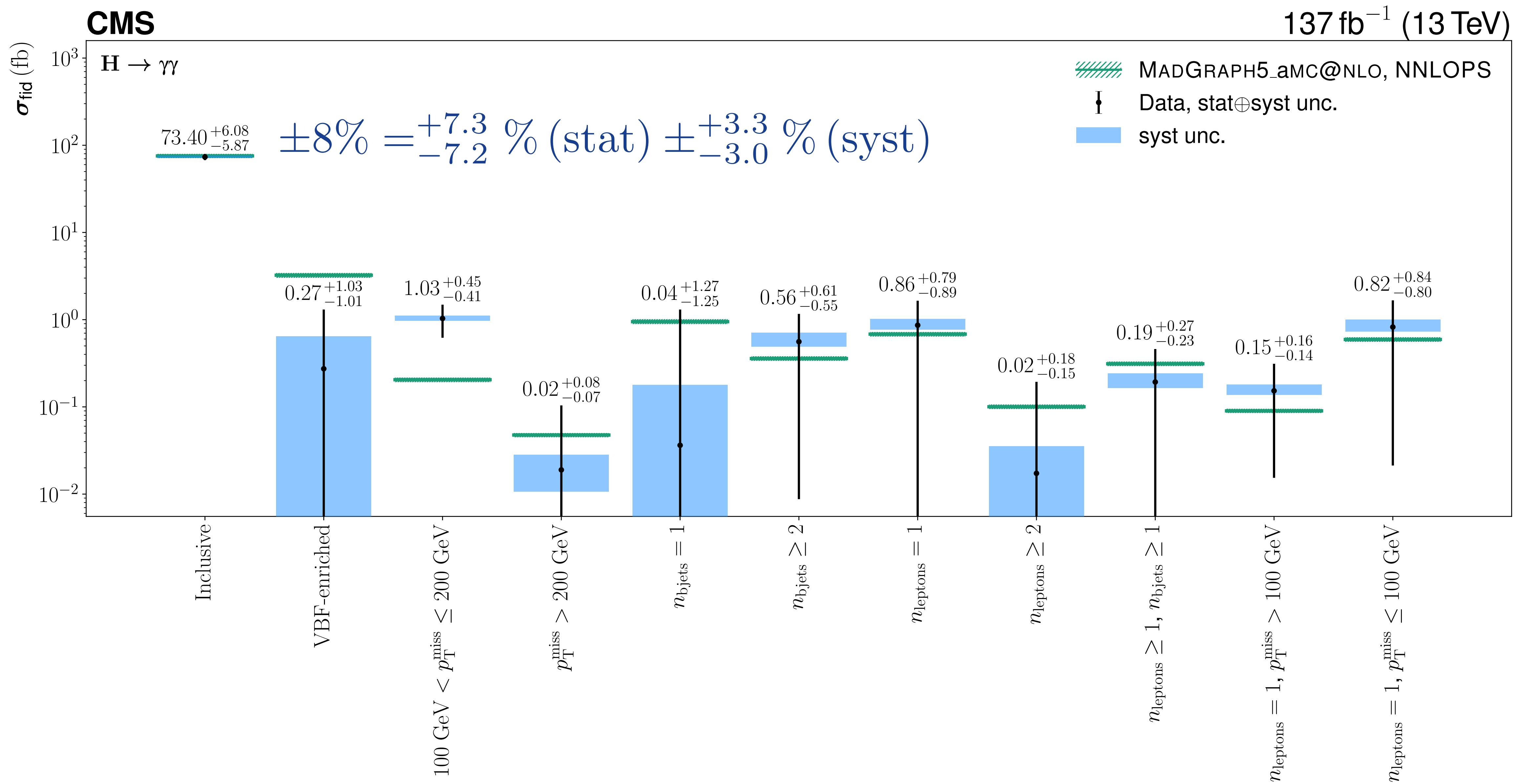


# Statistical Treatment

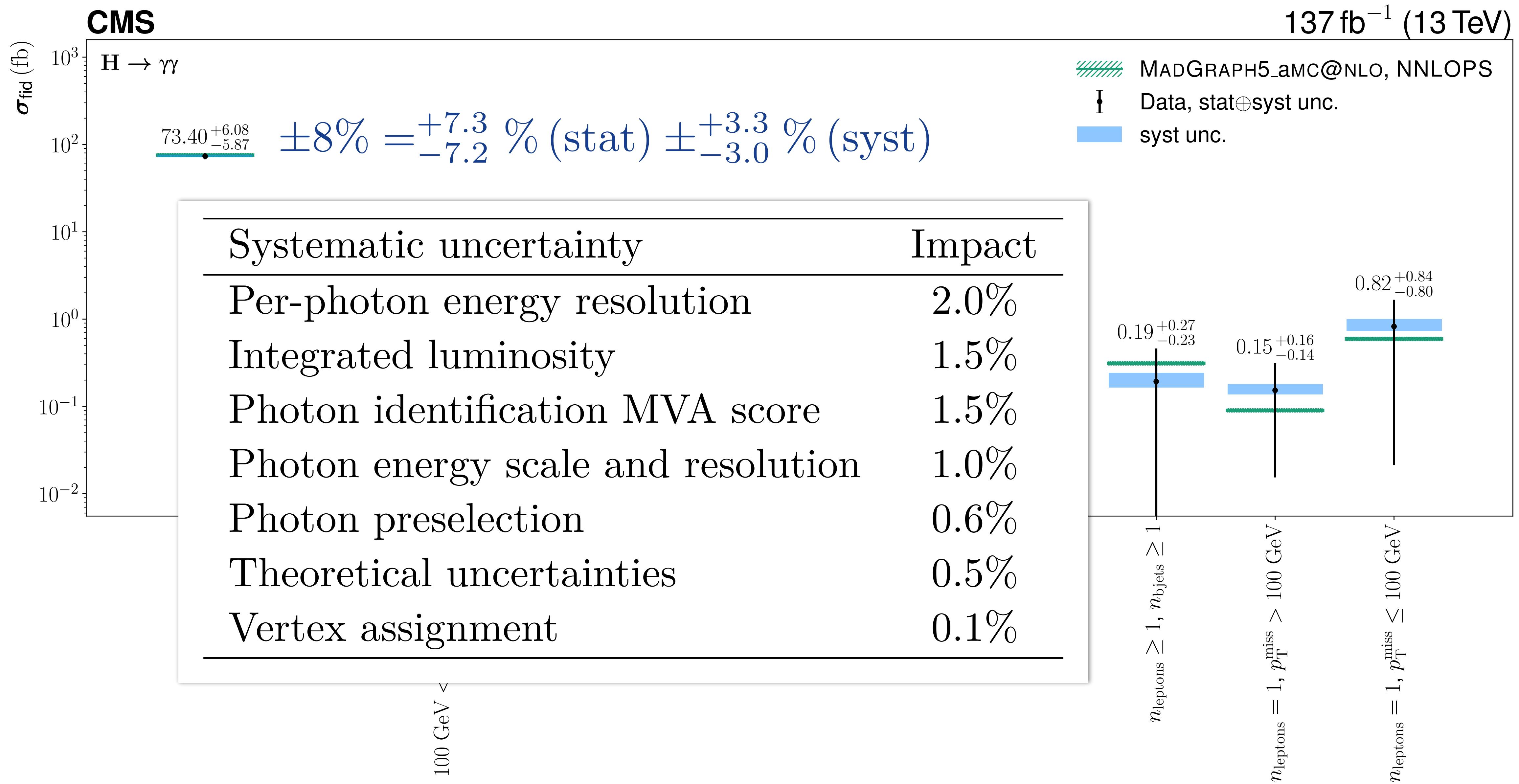
- Binned profile likelihood fit to  $m_{\gamma\gamma}$
- Signal modeled by up to four Gaussians
- Backgrounds from exp., power-law, Laurent and Bernstein polynomials included as nuisance parameter via discrete profiling



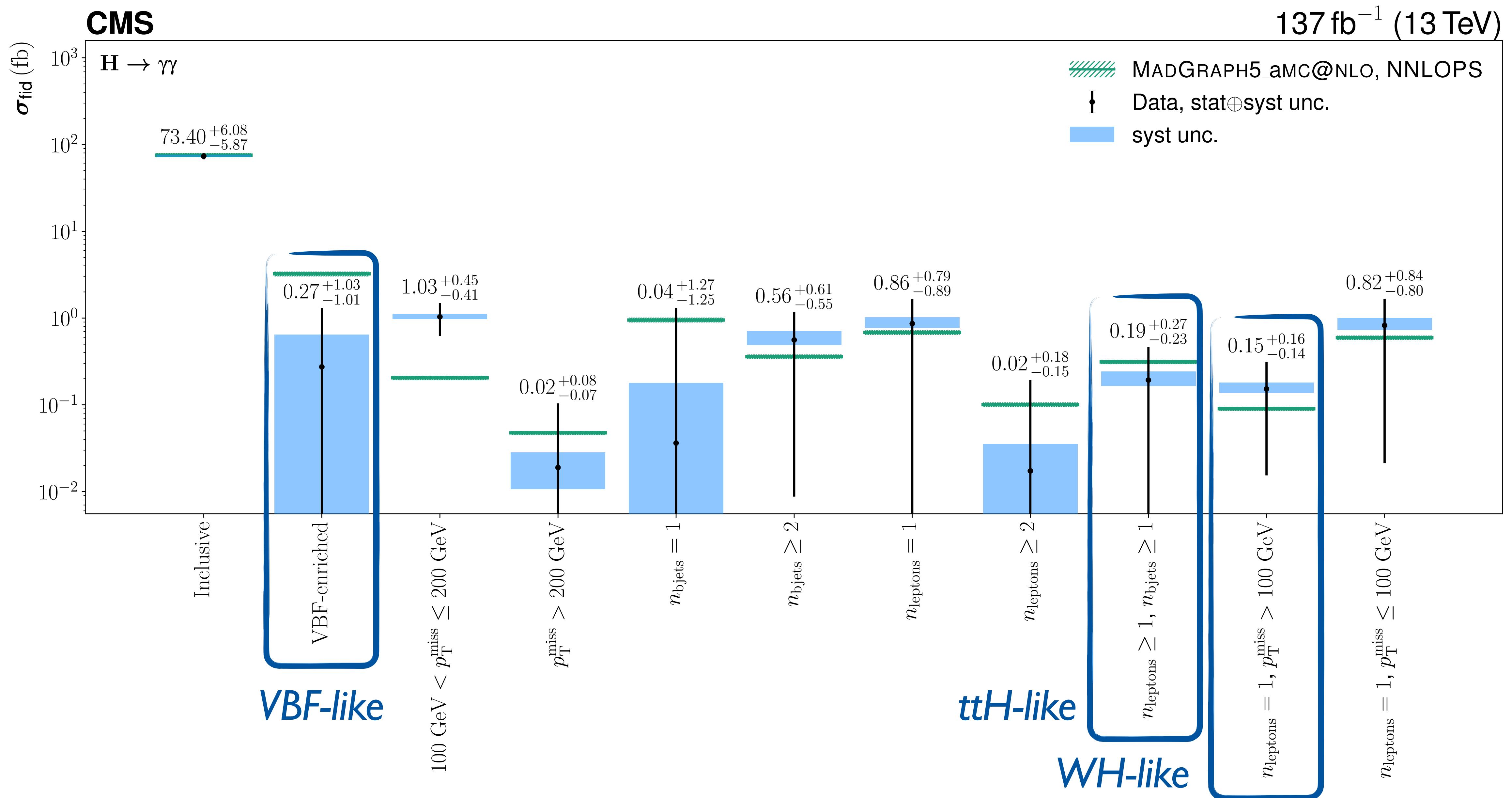
# Inclusive Cross Sections



# Inclusive Cross Sections

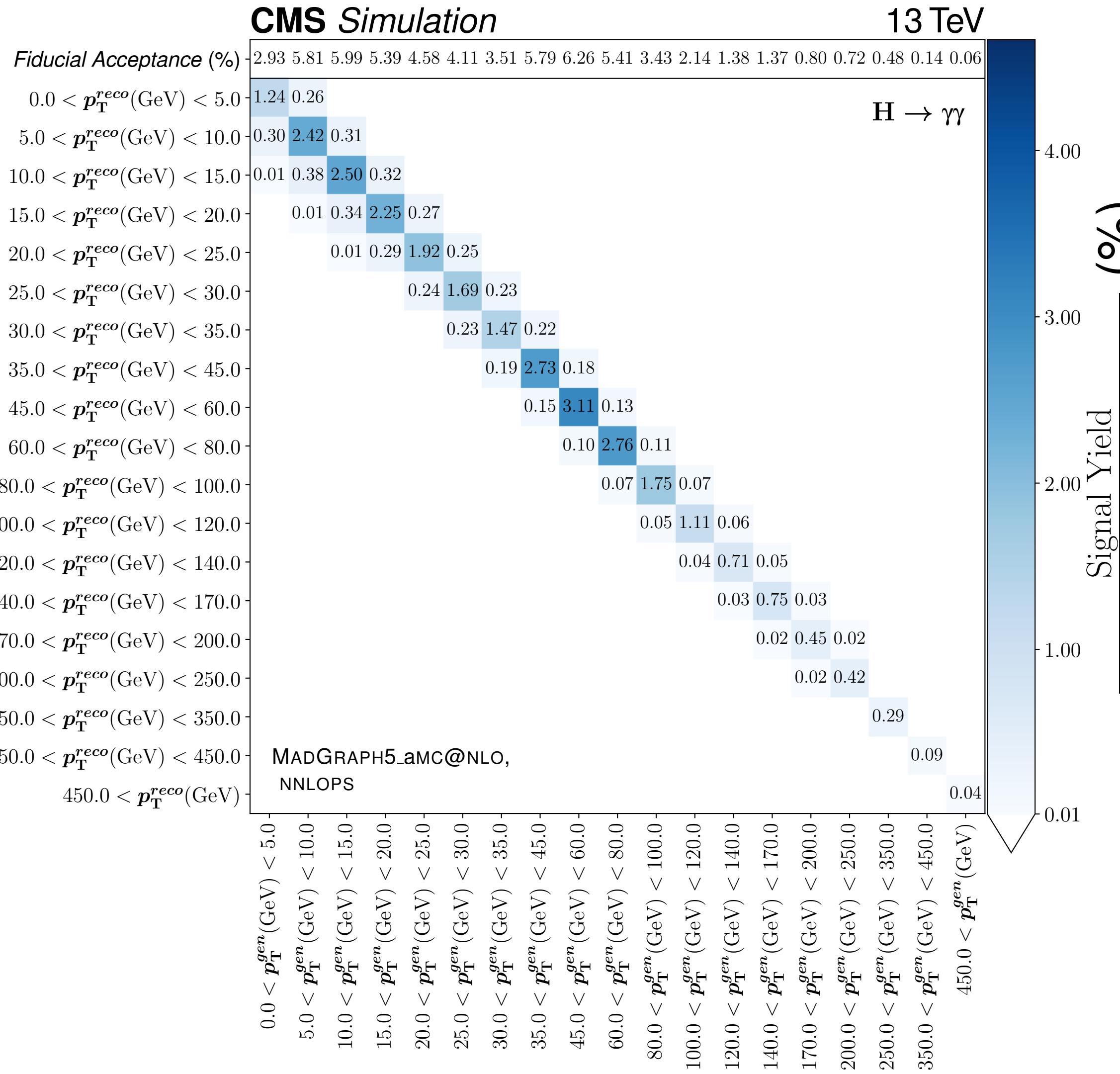


# Inclusive Cross Sections

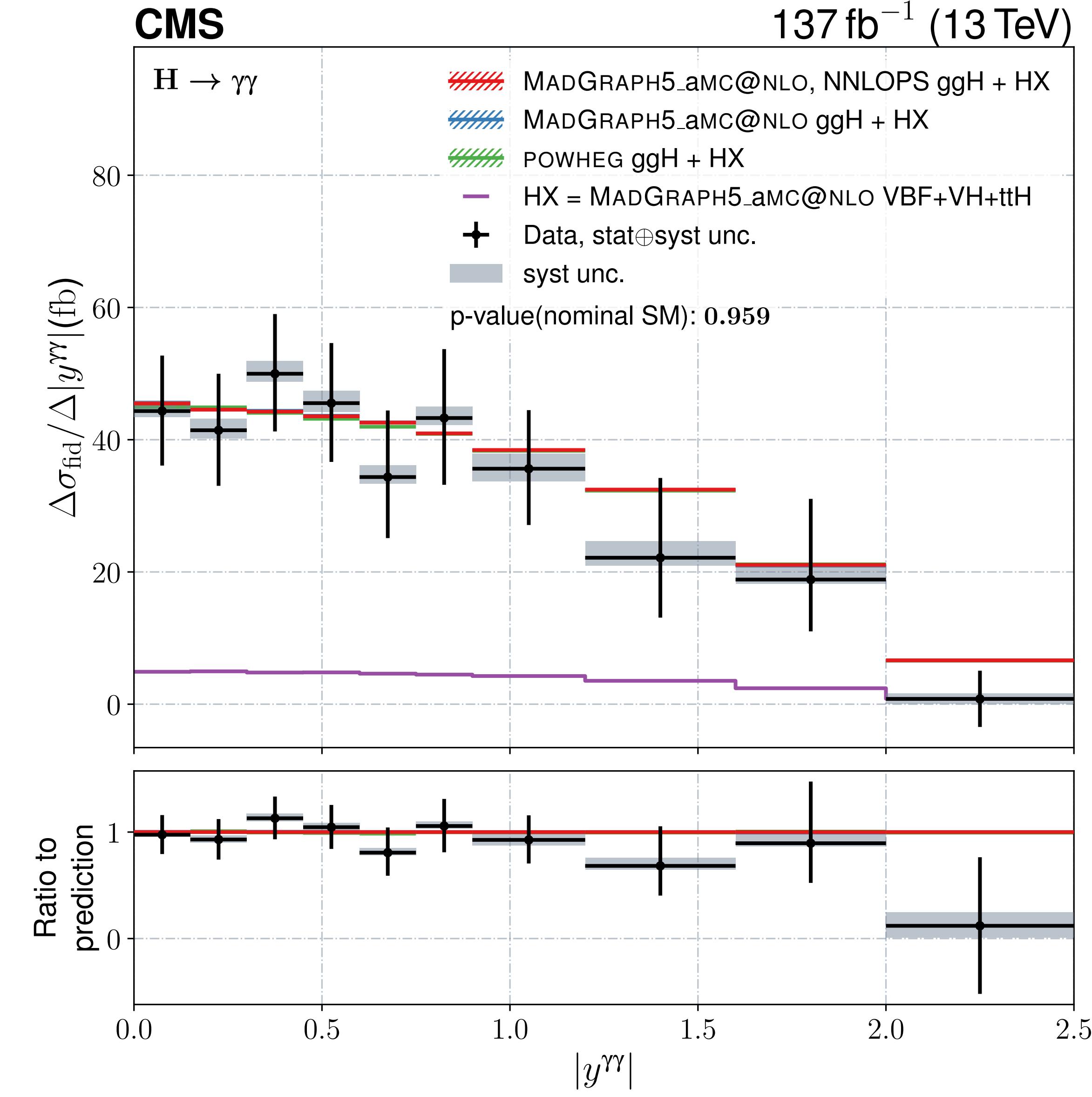
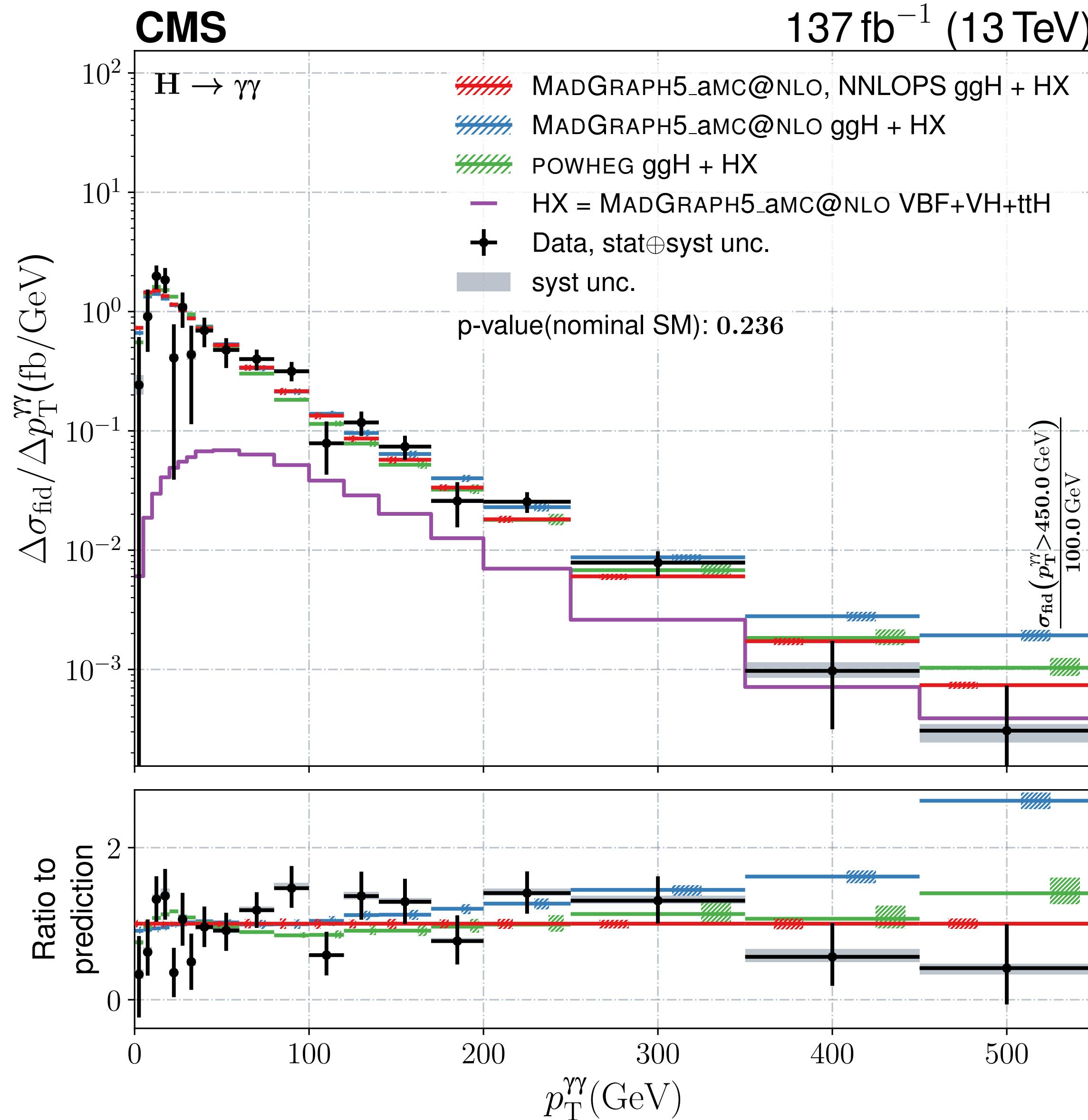


# Differential Cross Section: Migrations

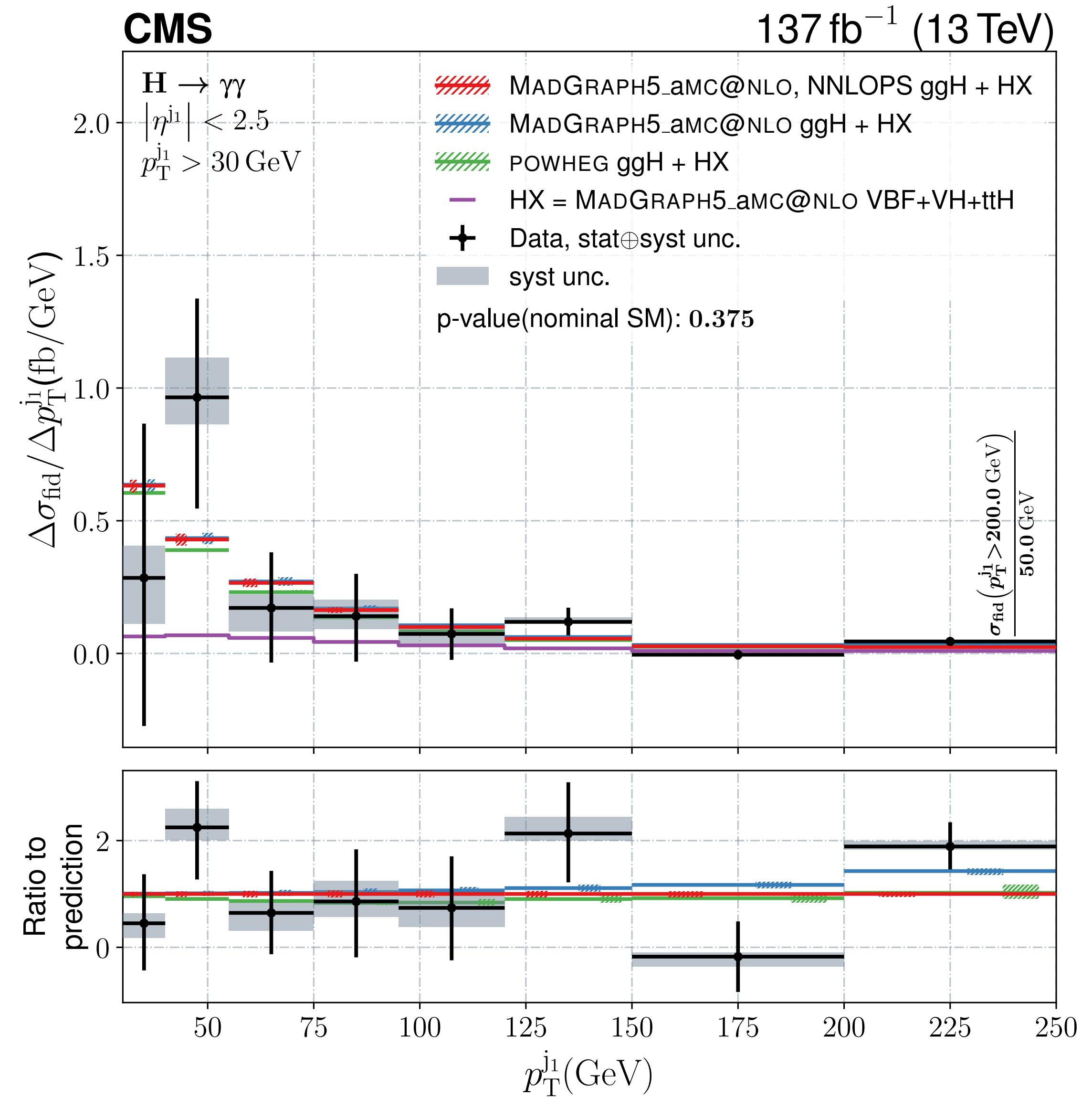
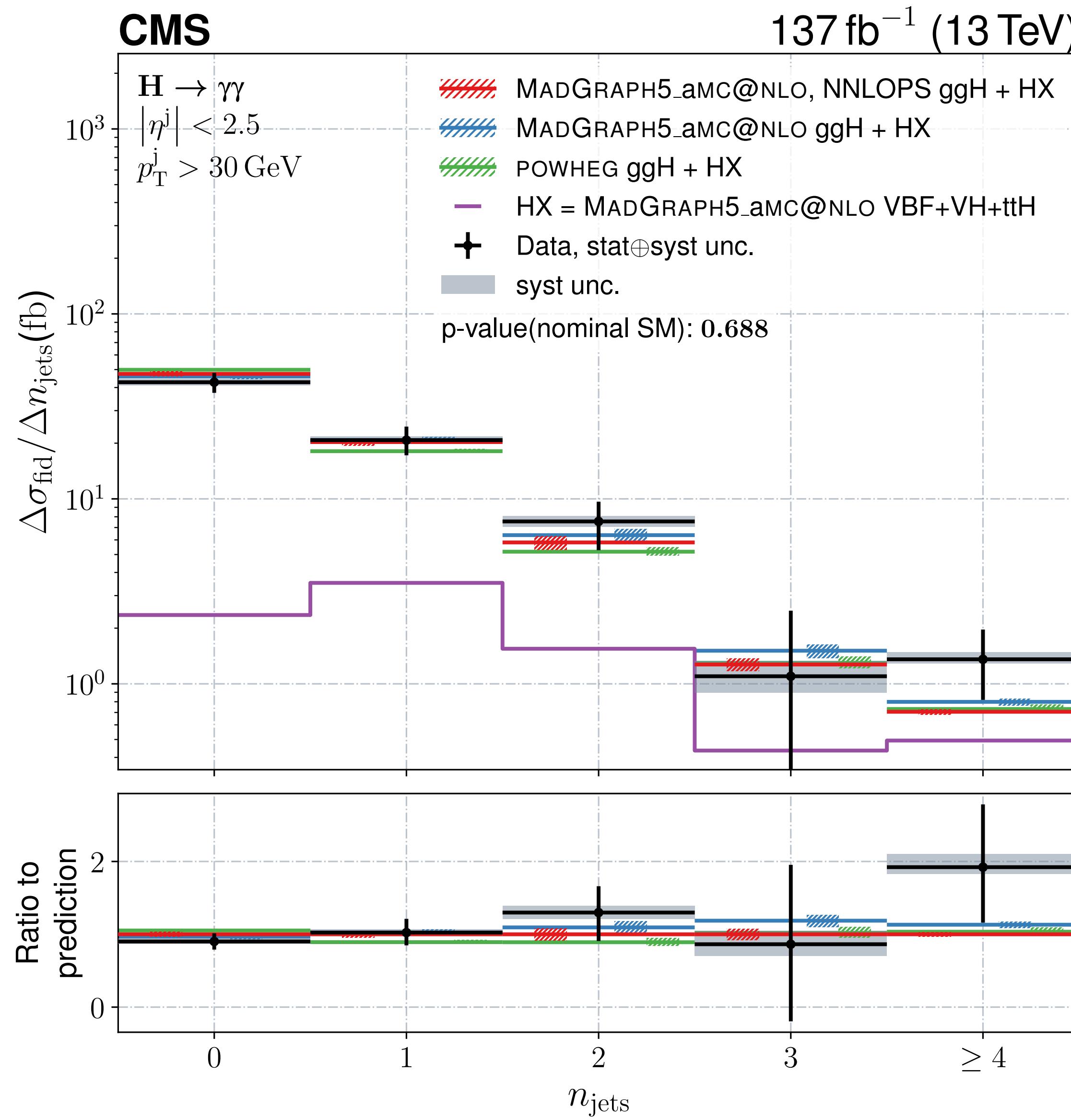
- As expected: small migrations for  $p_T^{\gamma\gamma}$ , stronger migrations for #jets



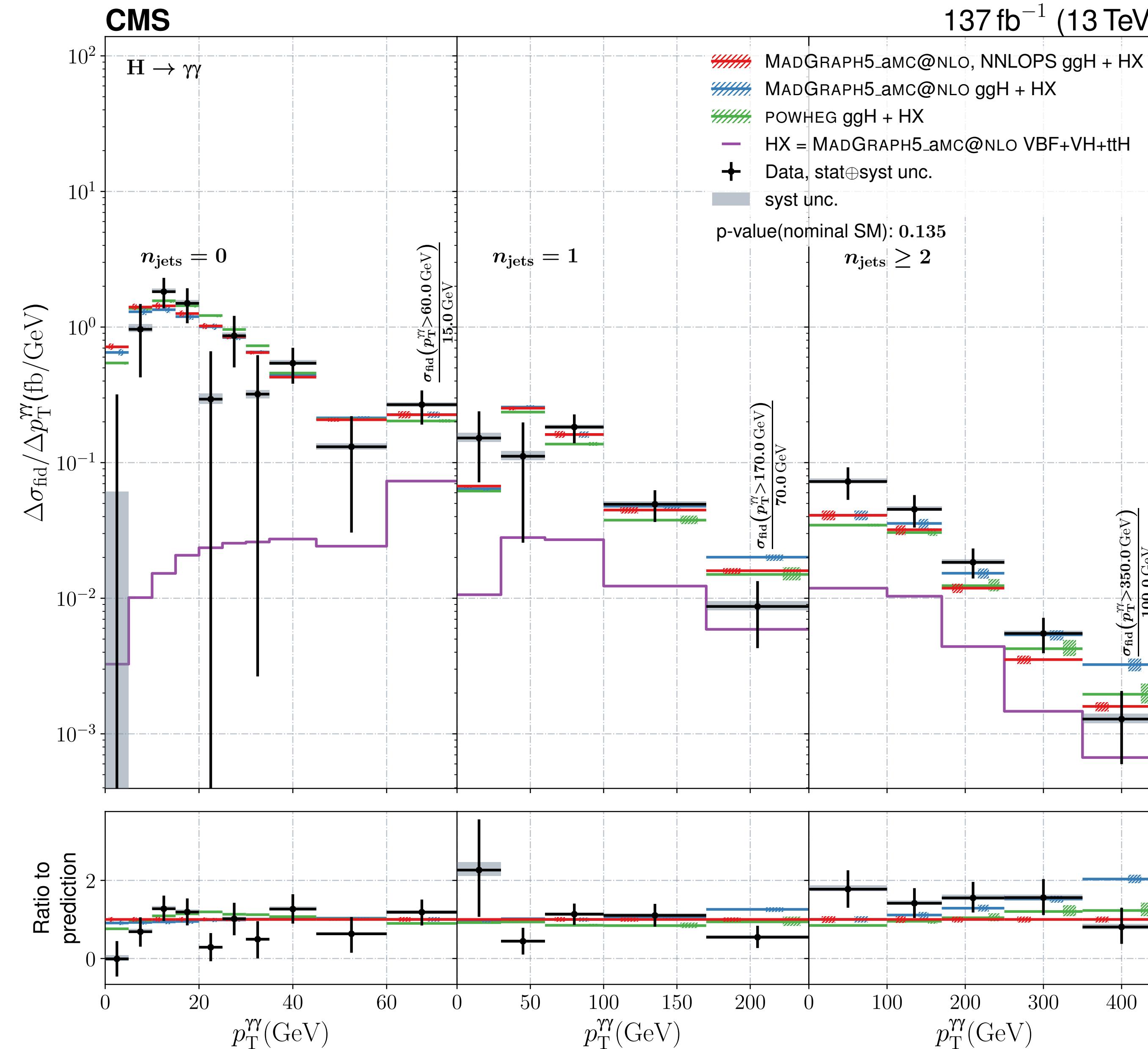
# Differentials: Diphoton Kinematics



# Differentials: Jets



# Double Differential: #jets vs. $p_T^{\gamma\gamma}$



## Summary

- Inclusive and fiducial cross sections with full Run-2 data
- Model-independent strategy based on expected  $m_{\gamma\gamma}$  resolution & photon ID
- Quantile morphing to improve modeling of these variables
- Uncertainty on fiducial inclusive cross section: 8%
- 1D and 2D differential cross sections
- *For more CMS Higgs differential measurements:*
  - *Alessandra Cappati's talk later this morning !*