



Analysis of non-resonant $HH \rightarrow b\bar{b}\gamma\gamma$ using CMS Run3 data

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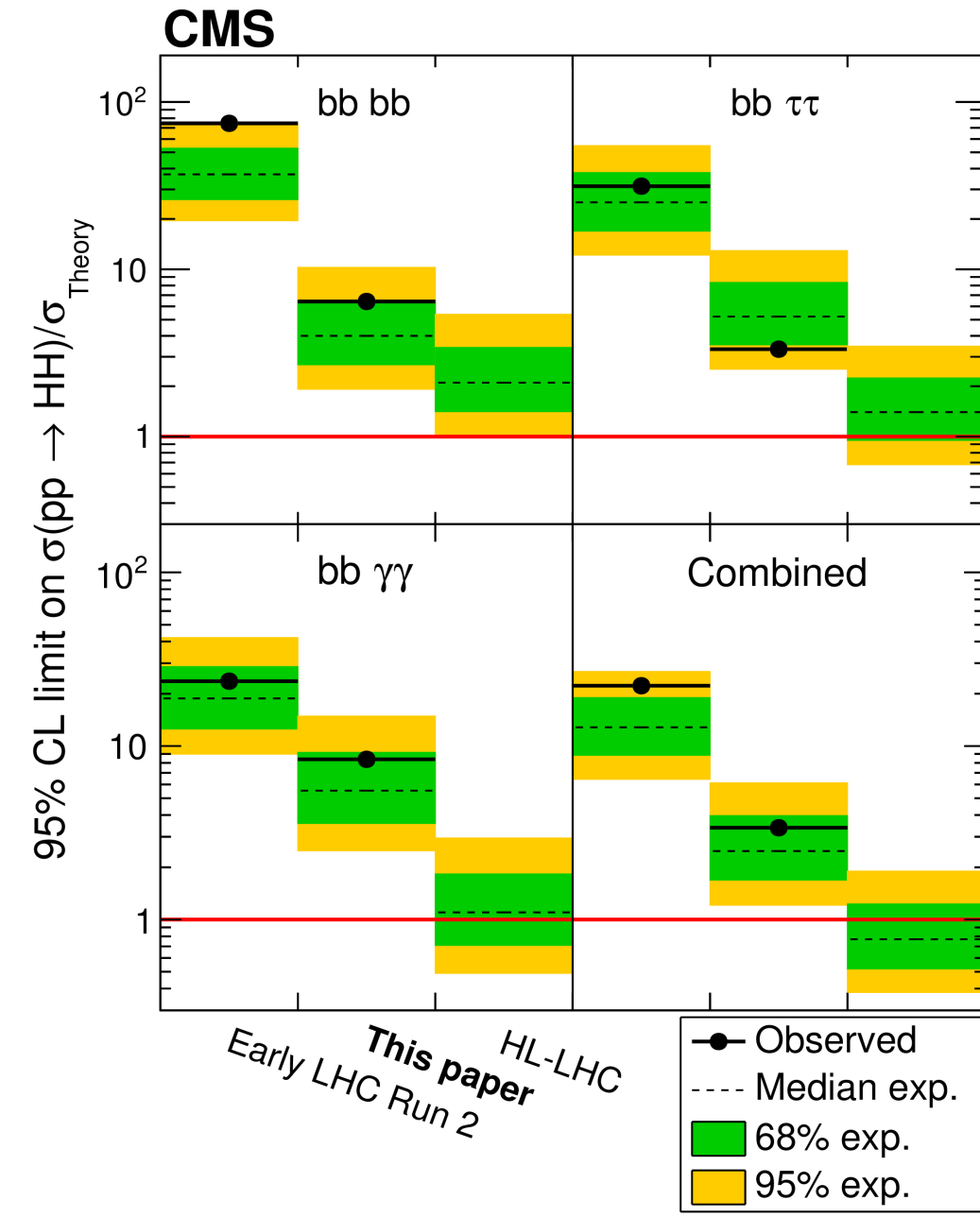
Introduction

Why Study HH?

- Tests the shape of the Higgs potential, still unprobed after Higgs discovery.
- Directly sensitive to the Higgs self-coupling, a key SM prediction.
- Probes BSM effects in the Higgs sector.

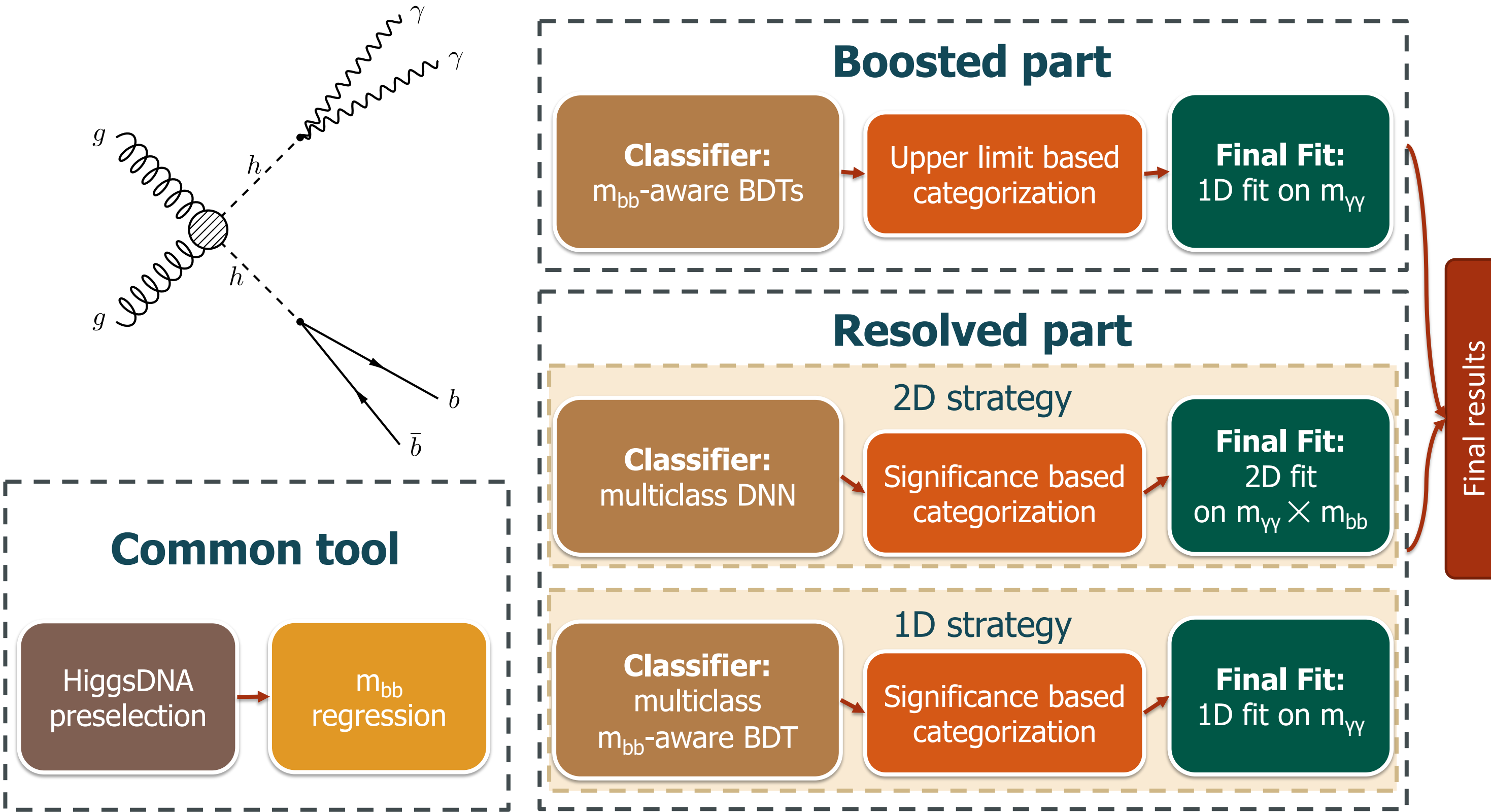
Why the $b\bar{b}\gamma\gamma$ final state?

- $H \rightarrow b\bar{b}$: largest branching ratio among Higgs decays.
- $H \rightarrow \gamma\gamma$: clean final state with excellent mass resolution and trigger efficiency.
- This channel provides a **powerful balance of rate and resolution**, ideal for early Run 3 studies.



Analysis Overview

- Dataset: **61.9 fb⁻¹** in **2022-2023** @13.6TeV



Event Categorization

Resolved Category – MVA-Based Event Selection

- The final selection relies on multiclass MVA discriminators to separate ggHH signal from backgrounds.
- Two complementary strategies are developed, differing in how m_{bb} is handled:
 - 1D Strategy**
 - m_{bb} is used as an input to the BDT, fit performed on **$m_{\gamma\gamma}$ only**;
 - Training optimized to suppress single-Higgs backgrounds resonant in $m_{\gamma\gamma}$;
 - Achieves $\sim 10^3$ (10^2) background rejection for QCD (top quark) at $\sim 60\%$ (50%) signal efficiency.
 - 2D Strategy**
 - DNN trained to be agnostic to the m_{bb} mass scale, m_{bb} is used explicitly in the 2D fit: **$m_{\gamma\gamma} \otimes m_{bb}$** ;
 - Designed to suppress all backgrounds, including QCD and top quark;
 - Achieves $\sim 10^3$ (10^2) background rejection for QCD (top quark) at $\sim 50\%$ (55%) signal efficiency.

- Both strategies provide **strong signal vs. background separation**.
- Use different assumptions and statistical treatments, yet give consistent results.

Boosted Category – Targeting High-pT $H \rightarrow b\bar{b}$ Events

- At high pT, AK4 jets fail to resolve close b quarks $\Rightarrow$ Loss in signal efficiency.
 - AK8 jets recover the efficiency, motivating a separate boosted category.

Preselection (from gen-level studies):

- AK8 jet with:
 - ≥ 2 subjets;
 - $\tau_2/\tau_1 < 0.75$;
 - $P_{\text{Net}}(X_{bb} \text{ vs. QCD}) > 0.4$;
 - $30 < m_{\text{softdrop}} < 210$ GeV.
- Candidate: AK8 jet with highest $P_{\text{Net}}(X_{bb} \text{ vs. QCD})$.

Final selection (by a two-step BDT approach):

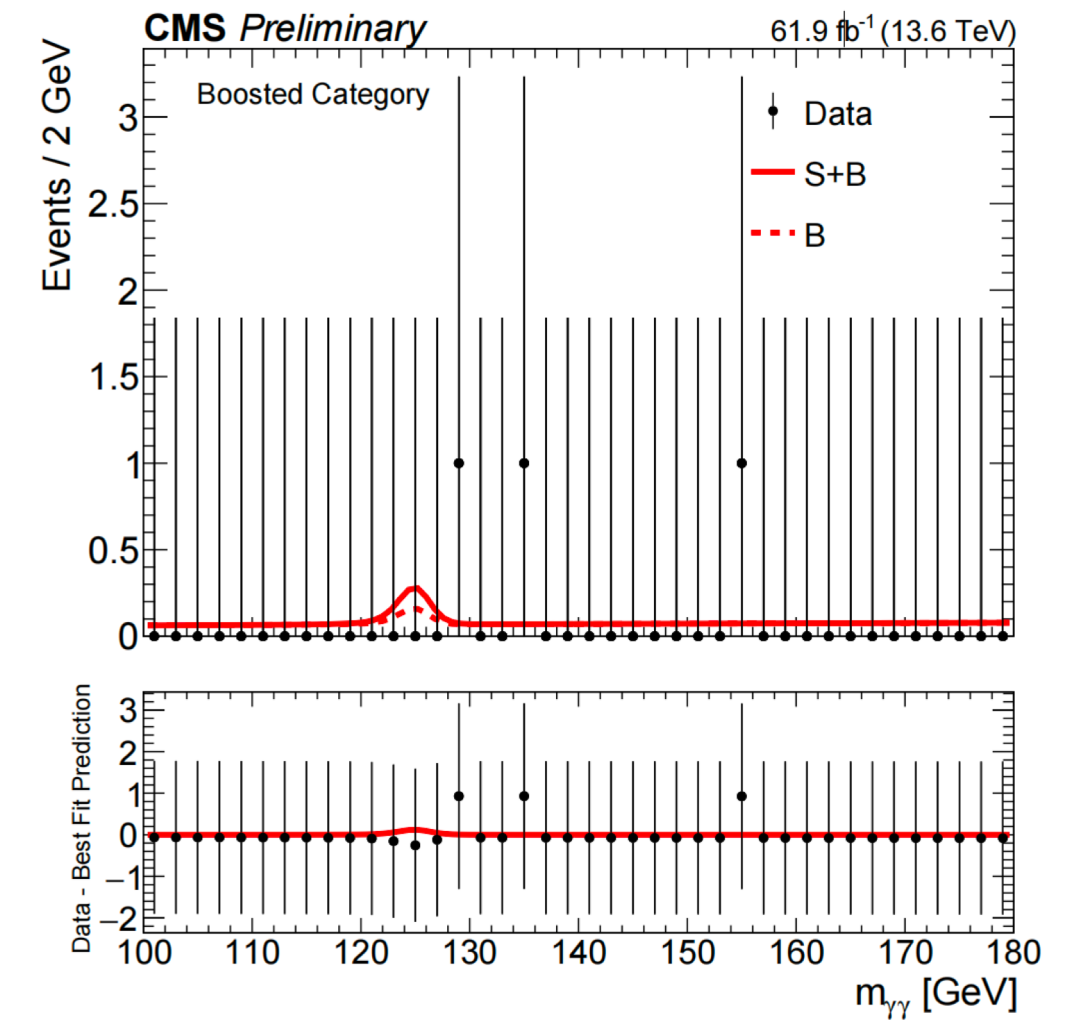
- BDT1: SM ggHH vs. VH;
- BDT2: SM ggHH vs. all background:
 - using BDT1 score as input.

- $\rightarrow$ Achieves strong background rejection.

1 Signal region defined by optimizing expected limit in cut-and-count.

Combination of Resolved & Boosted Categories

- Events are first assigned to the boosted category, which has high purity and low background;
- Remaining events are used to build the resolved category, maximizing sensitivity.



Event Selection

Preselection

- Events are triggered using the dedicated **$H \rightarrow \gamma\gamma$ HLT path**.
- The **photon selection** follows the $H \rightarrow \gamma\gamma$ group preselection.
- A **loose dijet preselection** is applied on the $H \rightarrow b\bar{b}$ dijet pair:
 - Specific enough to focus on the interesting phase space;
 - Generic enough to retain sufficient statistics for downstream MVA training.
- A tight lepton selection is used to implement **loose lepton vetoes**.
- An additional technique applied already at the preselection stage:

m_{bb} regression

- Improving the m_{bb} resolution enhances both event selection and signal extraction.
- Two regression methods are used:
 - PNet-based regression** (centrally provided);
 - Custom DNN regression** (trained on ggHH signal):
 - Trained to predict (gen – reco) / reco using jet- and event-level inputs.
- 10–20% m_{bb} resolution improvement**, varying with data-taking period.

Fitting Strategy

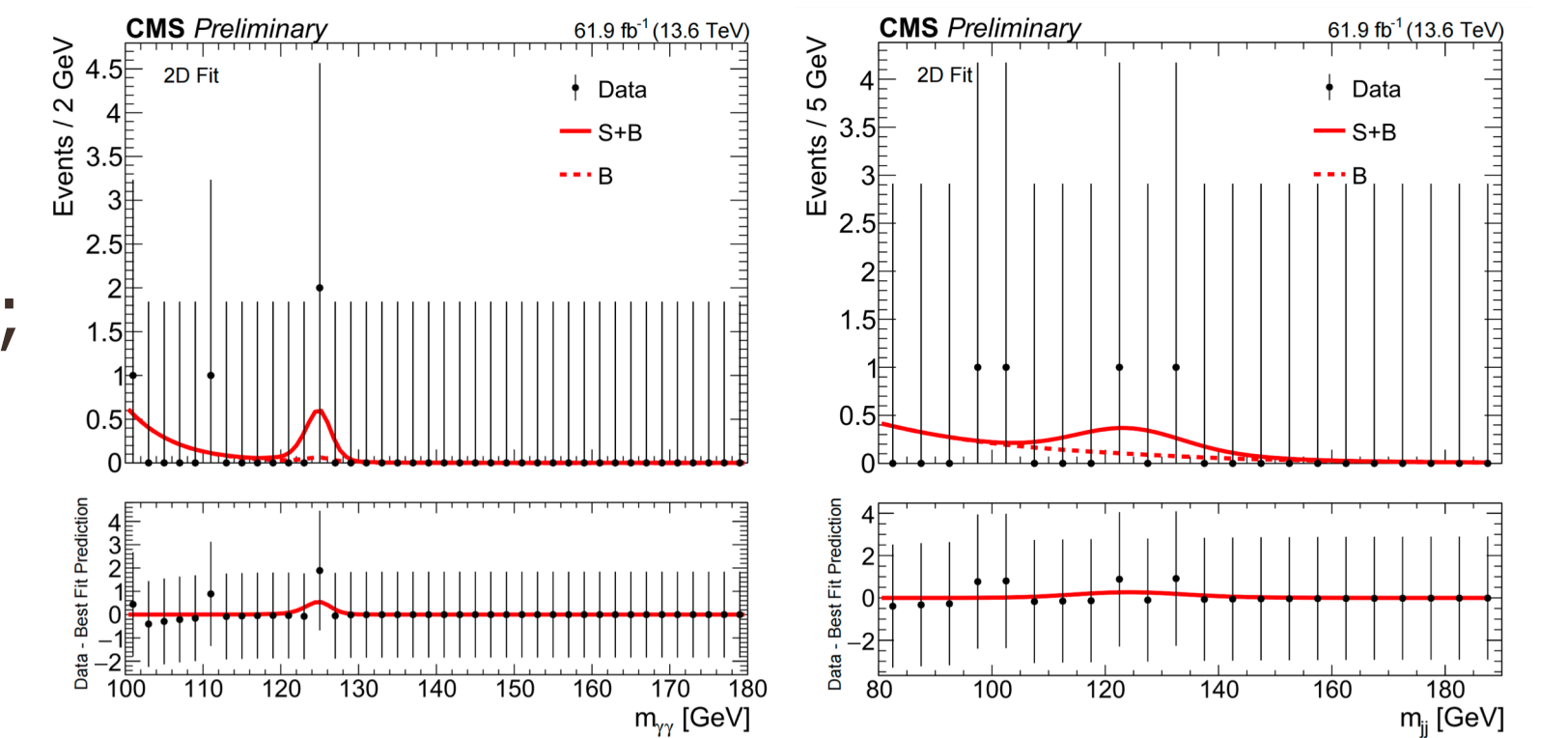
We use two signal extraction strategies:

1D Fit (Resolved & Boosted Categories)

- Fit on **$m_{\gamma\gamma}$ only**;
- Signal & resonant backgrounds**: sum of Gaussians or DCB;
- Nonresonant background**: discrete profiling over smooth analytic functions.

2D Fit (Resolved only)

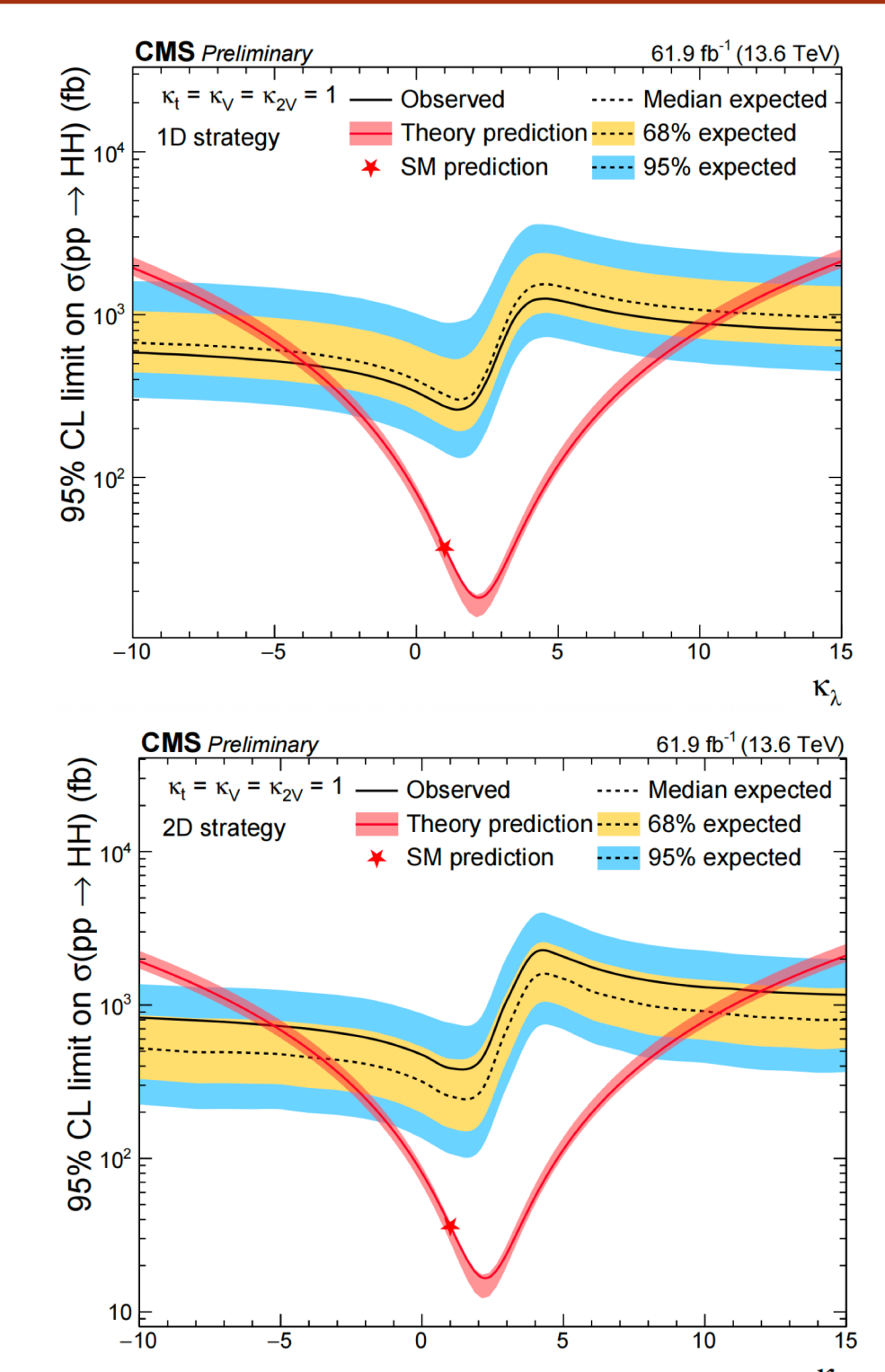
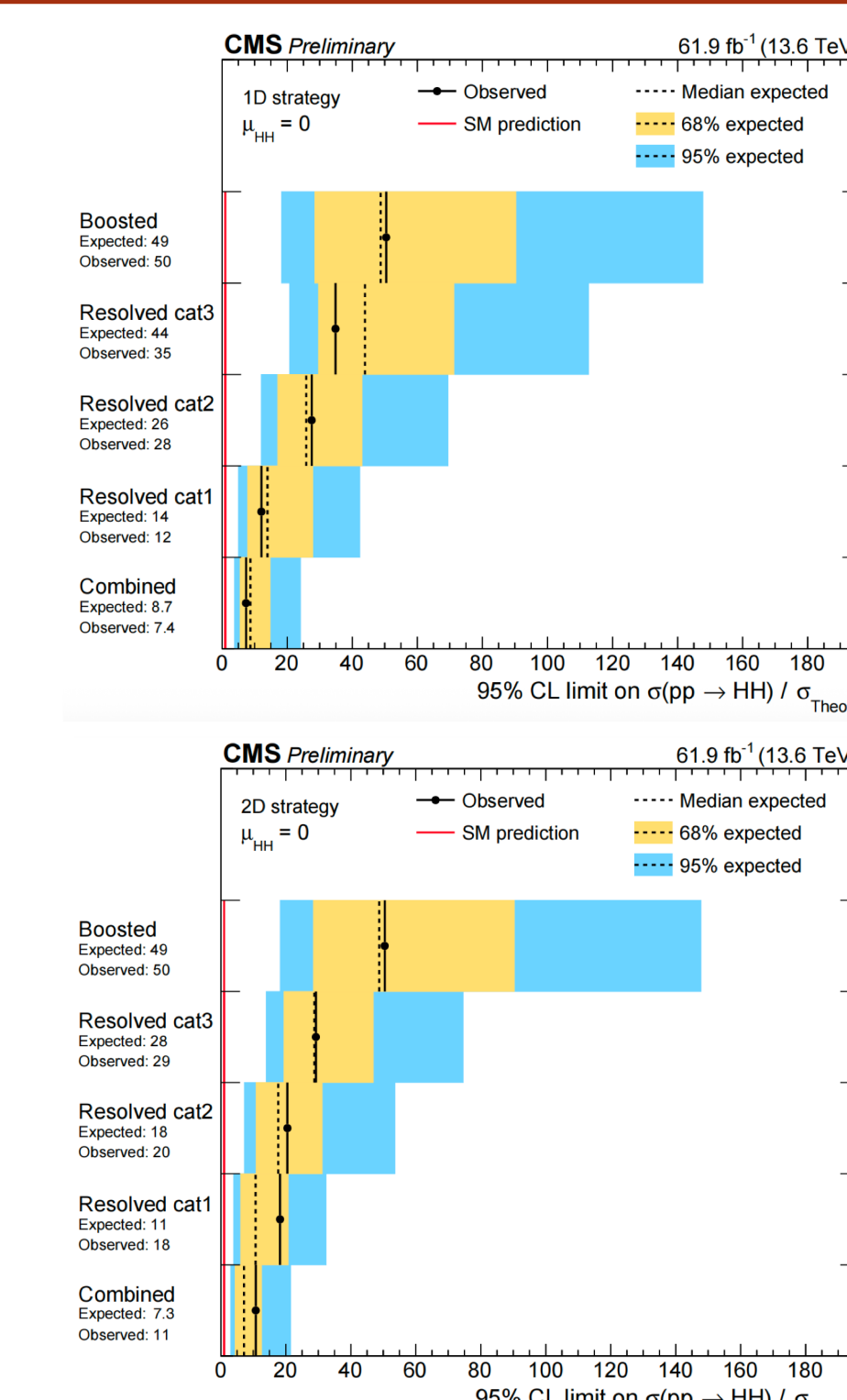
- Simultaneous fit on **$(m_{\gamma\gamma}, m_{bb})$** ;
- Signal**: sum of Gaussians;
- Resonant backgrounds**: DCB for $m_{\gamma\gamma}$; DCB or a combination of falling and peaking functions for m_{bb} .
- Nonresonant background**: discrete profiling as in 1D.



Results and Conclusion

- Search for non-resonant $HH \rightarrow b\bar{b}\gamma\gamma$** using **61.9 fb⁻¹** of Run 3 data (2022–2023, $\sqrt{s} = 13.6$ TeV).
- Two complementary fit strategies are employed: 1D ($m_{\gamma\gamma}$) and 2D ($m_{\gamma\gamma} \otimes m_{bb}$).
 - Observed (expected) 95% CL limits: **7.4 (**8.7**) $\times \sigma_{\text{SM}}$ (1D), **11.0 (**7.3**) $\times \sigma_{\text{SM}}$ (2D).****
 - Constraints on the Higgs self-coupling: **$-3.9 < \kappa_\lambda < 10.4$ (1D), $-5.0 < \kappa_\lambda < 12.0$ (2D).**
 - Both strategies give **consistent and robust results**.
 - The boosted events** improve sensitivity by $\sim 5\%$.

- The analysis is **statistics-limited**.
- Dominant uncertainties come from **background normalization**, **signal theory**, and **heavy-flavor modeling** in ggH.
- A **bias uncertainty** is included in the 2D fit to cover $m_{\gamma\gamma}$ – m_{bb} correlations.
- The analysis demonstrates the maturity of the CMS $HH \rightarrow b\bar{b}\gamma\gamma$ framework and its readiness for the full Run 3 dataset.



Reference: [1] The CMS Collaboration. A portrait of the Higgs boson by the CMS experiment ten years after the discovery. *Nature* **607**, 60–68 (2022). <https://doi.org/10.1038/s41586-022-04892-x>

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