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Search for Higgs boson pair production in $\gamma\gamma$ bb final state in *pp* collisions at $\sqrt{s} = 13$ TeV with the ATLAS detector

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Outline

- Motivation
- Analysis overview
- Event preselection
- Event Categorization
- Signal and backgound modeling
- Results
- Summary

arXiv:2112.11876 (this talk)

<u>JHEP 11 (2018) 040</u> (ATLAS 36 *fb*⁻¹)

JHEP 03 (2021) 257 (CMS)

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ATLAS-PHYS-PUB-2021-031 (HH summary)
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Motivation – Higgs self-coupling

- The Higgs boson completes the Standard Model of Particle Physics.
- However, the shape of **the Higgs potential** has yet to be measured.
- We can probe the Higgs potential by measuring the Higgs self-coupling (λ) .



HH production



HHyybb Analysis overview

Search for Non-resonant and Resonant HH production in $\gamma\gamma bb$ channel (full Run2 data, 139 fb^{-1}).

One of the most sensitive HH final states:

- $H \rightarrow bb$: large branching ratio
- $H \rightarrow \gamma \gamma$: excellent photon resolution, clean smoothly falling di-photon background for signal extraction



HHyybb Analysis overview

Search for Non-resonant and Resonant HH production in $\gamma\gamma bb$ channel (full Run2 data, 139 fb^{-1}).

Main backgrounds

- Non-resonant $\gamma\gamma$ backgrounds
- Single Higgs production

Common Preselection

- 2 identified and isolated photons
- 2 b-tagged jets (77% DL1r b-tagging efficiency)
- < 6 central jets (reject $t\bar{t}H$ events)
- Veto events containing an electron or muon

Multivariate method designed to reject background processes

Statistical results obtained from a fit of $m_{\gamma\gamma}$ distribution



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Event preselection

Photon identification (Tight WP)

Calorimeter- and track-based **isolation** within a cone of $\Delta R = 0.2$ $E_T^{iso} < 0.065 \cdot E_T$ and $p_T^{iso} < 0.05 \cdot E_T$

- ▶ $105 < m_{\gamma\gamma} < 160 \, GeV$
- $\succ p_T^{\gamma 1}/m_{\gamma \gamma} > 0.35, p_T^{\gamma 2}/m_{\gamma \gamma} > 0.25$
- DL1r b-tagging (a deep-learning neural network)
 WP: 77% efficiency
- Energy correction
- muon-in-jet correction: muons from semileptonic
 b-hadron decays
- *p_T*-reco correction: *p_T* loss due to neutrinos and objects outside of the jet cone



 m_{bb} resolution improved by about 22%



Event categorization

Non-resonant analysis: target SM HH $\rightarrow \gamma\gamma bb$ processes, and possible modifications to κ_{λ} .

Target only ggF HH production, but VBF HH events also considered as signal

Signal regions defined using $m^*_{\gamma\gamma bb}$ and BDT score

> Modified invariant mass $m^*_{\gamma\gamma bb} = m_{\gamma\gamma bb} - m_{\gamma\gamma} - m_{bb} + 250 \ GeV$

Provides cancellation of experimental resolution effects

Spliting into 2 mass regions provide enhanced sensitivity to κ_{λ}

Step one = 2 HH mass regions



Event categorization

Non-resonant analysis: target SM HH $\rightarrow \gamma\gamma bb$ processes, and possible modifications to κ_{λ} .

Target only ggF HH production, but VBF HH events also considered as signal

Signal regions defined using $m^*_{\gamma\gamma bb}$ and BDT score

Boosted Decision Tree, one BDT in each mass region

Against $\gamma\gamma$ and single Higgs backgrounds

BDT trained on photon, jet and missing transverse energy variables



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Step two = 4 BDT categories

Event categorization

Resonant analysis: target $X \rightarrow HH \rightarrow \gamma\gamma bb$ processes, with $m_X \in [251, 1000]$ GeV.

Non-resonant SM HH production included as **background**

Signal regions defined using $m^*_{\gamma\gamma bb}$ and BDT score

Modified invariant mass

1 category for each m_X

2σ window cut around each m_X

 σ = the standard deviation parameter of the Crystal Ball function that best fits the $m^*_{\gamma\gamma bb}$ Relaxed to 4σ for 900 GeV and 1000 GeV mass hypotheses

Boosted Decision Tree

Two BDTs against $\gamma\gamma$ +tt $\gamma\gamma$ and single Higgs backgrounds respectively

For each m_X , cut on the **combined BDT score**

Shared by all resonance masses to avoid lack of background at high mass BDTs trained on photon, jet and missing transverse energy variables



Signal and background modeling

The signal and backgrounds are extracted by fitting **analytic functions** to $m_{\gamma\gamma}$ distribution.

Signal parameterization

Modeled with double sided crystal ball (DSCB) function derived from MC

Non-resonant

- Fit to SM *HH* signal, model shared with *H* background
- No sizable dependence on κ_{λ} is observed

Resonant

• Fit to resonance signals, model shared with SM HH and H background

Background parameterization

- Modeled with exponential function
 - Function form determined from **spurious signal** studies
 - **spurious signal** = systematic uncertainty assigned to the function **choice**





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Observed (Expected) Results

No signal is observed. Exclusion limits at 95%CL are set.

Non-resonant

- $\sigma_{ggF+VBF}(HH)$ upper limit: 4.2 x SM (5.7 x SM)
- κ_{λ} interval: [-1.5, 6.7] ([-2.4, 7.7])



ATLAS 36 fb^{-1} JHEP 11 (2018) 040 $\sigma_{ggF+VBF}(HH)$ limit: 22 (28) x SM κ_{λ} interval: [-8.2, 13.2] ([-8.3, 13.2]) CMS JHEP 03 (2021) 257

 $σ_{ggF+VBF}(HH)$ limit: 7.7 (5.2) x SM $κ_{\lambda}$ interval: [-3.3, 8.5] ([-2.5, 8.2])

 $\sigma_{ggF+VBF}(HH)$ upper limit improved by a factor of 5 w.r.t 36 fb^{-1} ~2 from increase of luminosity

~rest from analysis strategy

 κ_{λ} interval shrinks by a factor of ~2

Observed (Expected) Results

No signal is observed. Exclusion limits at 95%CL are set.

Resonant

• $\sigma(X \rightarrow HH)$ upper limits vary between 610 fb and 47 fb (360 fb and 43 fb) in $m_X \in [251, 1000]$ GeV



ATLAS 36 *f b*⁻¹ <u>JHEP 11 (2018) 040</u>

 $\sigma(X \rightarrow HH)$ upper limits vary between 1.1 *pb* and 0.12 *pb*

 $(0.9 \ pb \text{ and } 0.15 \ pb) \text{ in } m_X \in [260, 1000] \text{ GeV}$

$\sigma(X \rightarrow HH)$ upper limits

improved by a factor of 2-3 depending on the m_X value

~2 from increase of luminosity

~1.2 from analysis strategy

The analyzed mass range expanded to lower values

Summary

Searches for non-resonant and resonant HH production are performed in the $bb\gamma\gamma$ final state (139 fb^{-1}). No significant excess with respect to the SM background expectation is observed.

Improvement compared to the previous ATLAS result based on **36** fb^{-1} data

- Extends the **data set** by more than a factor of 4
- Incorporates a **categorization** based on $m^*_{\gamma\gamma bb}$ and multivariate event selections
- More precise object reconstruction and calibration

Status: submitted to PRD



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Thanks!

HH production



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$HH \rightarrow b\bar{b}\gamma\gamma$ analysis in a nutshell



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Background Samples

Table 1: Summary of single Higgs boson background samples, split by production modes, and continuum background samples. The generator used in the simulation, the PDF set, and tuned parameters (tune) are also provided.

Process	Generator	PDF set	Showering	Tune
ggF	NNLOPS [65–67] [68, 69]	PDFLHC [42]	Рутніа 8.2 [70]	AZNLO [71]
VBF	Powheg Box v2 [39, 66, 72–78]	PDFLHC	Рутніа 8.2	AZNLO
WH	Powheg Box v2	PDFLHC	Рутніа 8.2	AZNLO
$qq \rightarrow ZH$	Powheg Box v2	PDFLHC	Рутніа 8.2	AZNLO
$gg \rightarrow ZH$	Powheg Box v2	PDFLHC	Рутніа 8.2	AZNLO
tŦH	Powheg Box v2 [73–75, 78, 79]	NNPDF3.0nlo[80]	Рутніа 8.2	A14 [<mark>81</mark>]
bbH	Powheg Box v2	NNPDF3.0nlo	Рутніа 8.2	A14
tHqj	MadGraph5_aMC@NLO	NNPDF3.0nlo	Рутніа 8.2	A14
tHW	MadGraph5_aMC@NLO	NNPDF3.0nlo	Рутніа 8.2	A14
$\gamma\gamma$ +jets	Sherpa v2.2.4 [56]	NNPDF3.0nnlo	Sherpa v2.2.4	_
$t\bar{t}\gamma\gamma$	MadGraph5_aMC@NLO	NNPDF2.31o	Рутніа 8.2	_

Non-resonant BDT variables

Table 2: Variables used in the BDT for the non-resonant analysis. The *b*-tag status identifies the highest fixed *b*-tag working point (60%, 70%, 77%) that the jet passes. All vectors in the event are rotated so that the leading photon ϕ is equal to zero.

Variable	Definition
Photon-related kin	ematic variables
$p_{\mathrm{T}}/m_{\gamma\gamma}$	Transverse momentum of the two photons scaled by their invariant mass $m_{\gamma\gamma}$
η and ϕ	Pseudo-rapidity and azimuthal angle of the leading and sub-leading photon
Jet-related kinema	tic variables
<i>b</i> -tag status	Highest fixed <i>b</i> -tag working point that the jet passes
p_{T},η and ϕ	Transverse momentum, pseudo-rapidity and azimuthal angle of the two jets with the highest <i>b</i> -tagging score
$p_{\mathrm{T}}^{bar{b}},\eta_{bar{b}}$ and $\phi_{bar{b}}$	Transverse momentum, pseudo-rapidity and azimuthal angle of <i>b</i> -tagged jets system
$m_{b\bar{b}}$	Invariant mass built with the two jets with the highest <i>b</i> -tagging score
$H_{ m T}$	Scalar sum of the $p_{\rm T}$ of the jets in the event
Single topness	For the definition, see Eq. (1)

$$\chi_{Wt} = \min \sqrt{\left(\frac{m_{j_1 j_2} - m_W}{m_W}\right)^2 + \left(\frac{m_{j_1 j_2 j_3} - m_t}{m_t}\right)^2},$$

Missing transverse momentum-related variables

$E_{\rm T}^{\rm miss}$ and $\phi^{\rm miss}$	Missing transverse momentum and its azimuthal	angle
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Non-resonant Categorization

$$Z = \sqrt{2} * [(s+b) * \log(1 + s/b) - s]$$

Low mass region



High mass region





Events / 2.5 GeV

25

15

10

5

0

Resonant BDT variables

Table 4: Variables used in the BDT for the resonant analysis. For variables depending on *b*-tagged jets, only jets *b*-tagged using the 77% working point are considered as described in Section 4.1.

Variable	Definition
Photon-related kinematic variab	les
$p_{\rm T}^{\gamma\gamma}, y^{\gamma\gamma}$	Transverse momentum and rapidity of the di-photon system
$\Delta \phi_{\gamma\gamma}$ and $\Delta R_{\gamma\gamma}$	Azimuthal angular distance and ΔR between the two photons
Jet-related kinematic variables	
$m_{b\bar{b}}, p_{\rm T}^{b\bar{b}}$ and $y_{b\bar{b}}$	Invariant mass, transverse momentum and rapidity of the <i>b</i> -tagged jets system
$\Delta \phi_{bar{b}}$ and $\Delta R_{bar{b}}$	Azimuthal angular distance and ΔR between the two <i>b</i> -tagged jets
$N_{\rm jets}$ and $N_{b-\rm jets}$	Number of jets and number of <i>b</i> -tagged jets
H_{T}	Scalar sum of the $p_{\rm T}$ of the jets in the event
Photons and jets-related kinema	tic variables
$m_{b\bar{b}\gamma\gamma}$	Invariant mass built with the di-photon and <i>b</i> -tagged jets system
	Distance in rapidity, azimuthal angle and ΔR between the

$$BDT_{tot} = \frac{1}{\sqrt{C_1^2 + C_2^2}} \sqrt{C_1^2 \left(\frac{BDT_{\gamma\gamma} + 1}{2}\right)^2 + C_2^2 \left(\frac{BDT_{SingleH} + 1}{2}\right)^2}$$

2-stage optimization

- 1. Maximize significance for each resonance
 - Different coefficients and BDT scores
- 2. Select coefficients providing a significance within 5% from the maximum value, for each resonance
 - A common $C_1 = 0.65$ coefficient is found, individual BDT cuts are used

144 -	Invariant mass built with the di-photon and <i>b</i> -tagger
mbbγγ	system
$\Delta y_{\gamma\gamma,b\bar{b}}, \Delta \phi_{\gamma\gamma,b\bar{b}}$ and $\Delta R_{\gamma\gamma,b\bar{b}}$	Distance in rapidity, azimuthal angle and ΔR betwee di-photon and the <i>b</i> -tagged jets system

Signal modeling - DSCB

<u>A Gaussian core + asymmetric power law tails</u>



where N is a normalization factor and the six parameters are

- μ_{CB} and σ_{CB} describe the mean and the width of the Gaussian core, which are combined in $t = (m_{\gamma\gamma} \mu_{CB}) / \sigma_{CB};$
- α_{low} and α_{high} are the positions of the transitions with respect to μ_{CB} from the Gaussian core to power-law tails, in unit of σ_{CB} , on the low and high mass sides respectively;
- n_{low} and n_{high} are the exponents of the low and high mass tails. With the α 's, they define $R_{low} = \frac{n_{low}}{\alpha_{low}}$ and R_{high} similarly.

non-Gaussian tails can arise from experimental effects, such as photon energy mismeasurements.

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Diphoton background decomposition

- Reconstructed $\gamma\gamma$ events is mainly composed of $\gamma\gamma$, γ -jets and jet-jet events, where **the jet(s) fake(s) a real photon**.
- The 2x2D sideband method is developed using the discriminating power of **photon identification and isolation criteria**.
- The event yields in the signal region and the 15 sidebands can be expressed as **functions** of the photon efficiencies, jet fake rates and correlation coefficients.



Reference

Suffers from low statistics, not used in constructing the background templates for the spurious signal procedure.

Spurious signal

Spurious signal: a bias estimated from a signal + background fit to a background-only MC template.

$$N_{sp} = \max_{121 < m_H < 129 \, GeV} |N_s(m_H)|$$

Selection criteria:

□ $N_{sp} < 20\%$ of the data's statistical uncertainty + 2 × the MC background template statistical uncertainty □ must satisfy a simple χ^2 requirement in a background-only fit to the MC template: $p - value(\chi^2) > 1\%$

- The least number of parameters is preferred.
- The **smaller systematic uncertainty** (spurious signal) is preferred.

Wald tests show that the data do not prefer a higher degree functional form with respect to the exponential form.

Systematic uncertainties

In general the analysis is almost completely statistically dominated with the Run 2 dataset

		Relative impact of the sy	stematic uncertainties [%]
Source	Туре	Nonresonant analysis HH	Resonant analysis $m_X = 300 \text{ GeV}$
Experimental			
Photon energy resolution	Norm. + Shape	0.4	0.6
Jet energy scale and resolution	Normalization	< 0.2	0.3
Flavor tagging	Normalization	< 0.2	0.2
Theoretical			
Factorization and renormalization scale	Normalization	0.3	< 0.2
Parton showering model	Norm. + Shape	0.6	2.6
Heavy-flavor content	Normalization	0.3	< 0.2
$\mathcal{B}(H \to \gamma \gamma, b \bar{b})$	Normalization	0.2	< 0.2
Spurious signal	Normalization	3.0	3.3

Statistical framework

> The results of the analysis are obtained from a **maximum-likelihood fit** of the $m\gamma\gamma$ distribution.

Likelihood

$$\mathcal{L} = \prod_{c} \left(\operatorname{Pois}(n_{c} | N_{c}(\boldsymbol{\theta})) \cdot \prod_{i=1}^{n_{c}} f_{c}(m_{\gamma\gamma}^{i}, \boldsymbol{\theta}) \cdot G(\boldsymbol{\theta}) \right)$$

Event parameterization

$$N_{c}(\boldsymbol{\theta}) = \mu \cdot N_{HH,c}(\boldsymbol{\theta}_{HH}^{\text{yield}}) + N_{\text{bkg,c}}^{\text{res}}(\boldsymbol{\theta}_{\text{res}}^{\text{yield}}) + N_{\text{SS,c}} \cdot \boldsymbol{\theta}^{\text{SS,c}} + N_{\text{bkg,c}}^{\text{non-res}}$$

Model PDF

$$\frac{f_c(m_{\gamma\gamma}, \boldsymbol{\theta})}{f_c(m_{\gamma\gamma}, \boldsymbol{\theta})} = \left[\mu \cdot N_{HH,c}(\boldsymbol{\theta}_{HH}^{\text{yield}}) \cdot f_{HH,c}(m_{\gamma\gamma}, \boldsymbol{\theta}_{HH}^{\text{shape}}) + N_{\text{bkg,c}}^{\text{res}}(\boldsymbol{\theta}_{\text{res}}^{\text{yield}}) \cdot f_{\text{bkg,c}}^{\text{res}}(m_{\gamma\gamma}, \boldsymbol{\theta}_{\text{res}}^{\text{shape}}) + N_{\text{SS,c}}^{\text{son-res}} \cdot \boldsymbol{\theta}_{HH}^{\text{shape}} \cdot f_{HH,c}(m_{\gamma\gamma}, \boldsymbol{\theta}_{HH}^{\text{shape}}) + N_{\text{bkg,c}}^{\text{non-res}} \cdot f_{\text{bkg,c}}^{\text{non-res}}(m_{\gamma\gamma}, \boldsymbol{\theta}_{\text{non-res}}^{\text{shape}}) \right] / N_c(\boldsymbol{\theta}_{\text{non-res}}^{\text{yield}}).$$

Statistical framework

The measurement of the parameter of interest is carried out using a statistical test based on the profile likelihood ratio

the profile likelihood ratio

the profile-likelihood-ratio-based test statistic

$$\Lambda(\mu) = \frac{\mathcal{L}(\mu, \hat{\hat{\theta}}(\mu))}{\mathcal{L}(\hat{\mu}, \hat{\theta})}$$

$$\tilde{q}_{\mu} = \begin{cases} -2\ln\frac{\Lambda(\mu,\hat{\hat{\boldsymbol{\theta}}}(\mu))}{\Lambda(0,\hat{\hat{\boldsymbol{\theta}}}(0))} & \hat{\mu} < 0, \\ -2\ln\frac{\Lambda(\mu,\hat{\hat{\boldsymbol{\theta}}}(\mu))}{\Lambda(\hat{\mu},\hat{\boldsymbol{\theta}}(\mu))} & 0 \le \hat{\mu} \le \mu, \\ 0 & \hat{\mu} > \mu. \end{cases}$$

κ_{λ} reweighting for ggF HH samples

Common HH procedure. The method derives the scale factors as a function of κ_{λ} in bins of m_{HH} by performing a linear combination of samples generated at $\kappa_{\lambda} = 0, 1, 20$.



$$\mathcal{A}(\kappa_t, \kappa_{\lambda}) = \kappa_t^2 \mathcal{A}_1 + \kappa_t \kappa_{\lambda} \mathcal{A}_2 \qquad \qquad H$$
$$\sigma_{ggF}(pp \to HH) \propto \int \kappa_t^4 \left[|\mathcal{A}_1|^2 + 2\left(\frac{\kappa_{\lambda}}{\kappa_t}\right) \Re(\mathcal{A}_1^* \mathcal{A}_2) + \left(\frac{\kappa_{\lambda}}{\kappa_t}\right)^2 |\mathcal{A}_2|^2 \right]$$

$$\sigma(\kappa_t = 1, \kappa_{\lambda} = 0) \sim |\mathcal{A}_1|^2$$

$$\sigma(\kappa_t = 1, \kappa_{\lambda} = 1) \sim |\mathcal{A}_1|^2 + 2\Re \mathcal{A}_1^* \mathcal{A}_2 + |\mathcal{A}_2|^2$$

$$\sigma(\kappa_t = 1, \kappa_{\lambda} = 20) \sim |\mathcal{A}_1|^2 + 2 \cdot 20\Re \mathcal{A}_1^* \mathcal{A}_2 + 20^2 |\mathcal{A}_2|^2$$

$$\sigma(\kappa_{t},\kappa_{\lambda}) \sim \kappa_{t}^{2} \left[\left(\kappa_{t}^{2} + \frac{\kappa_{\lambda}^{2}}{20} - \frac{399}{380} \kappa_{\lambda} \kappa_{t} \right) |S(1,0)|^{2} + \left(\frac{40}{38} \kappa_{\lambda} \kappa_{t} - \frac{2}{38} \kappa_{\lambda}^{2} \right) |S(1,1)|^{2} + \left(\frac{\kappa_{\lambda}^{2} - \kappa_{\lambda} \kappa_{t}}{380} \right) |S(1,20)|^{2} \right]$$

$$+ \left(\frac{\kappa_{\lambda}^{2} - \kappa_{\lambda} \kappa_{t}}{380} \right) |S(1,20)|^{2} \right]$$

$$U_{t}^{2} = \frac{1}{2022/7/25}$$

HH summary



good sensitivity at low resonant masses