

Search for $H \rightarrow aa \rightarrow \mu\mu bb$

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On behalf of the CMS collaboration

Isfahan University of Technology

The 19th Workshop of the LHC Higgs Working Group



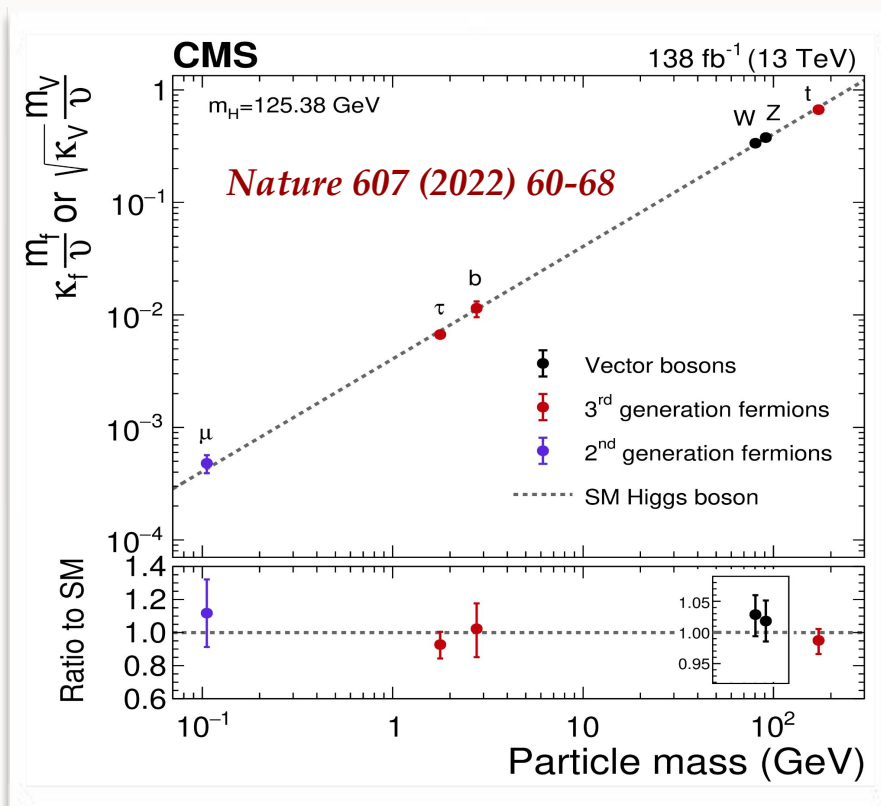
Mon, 28 Nov 2022

Outline

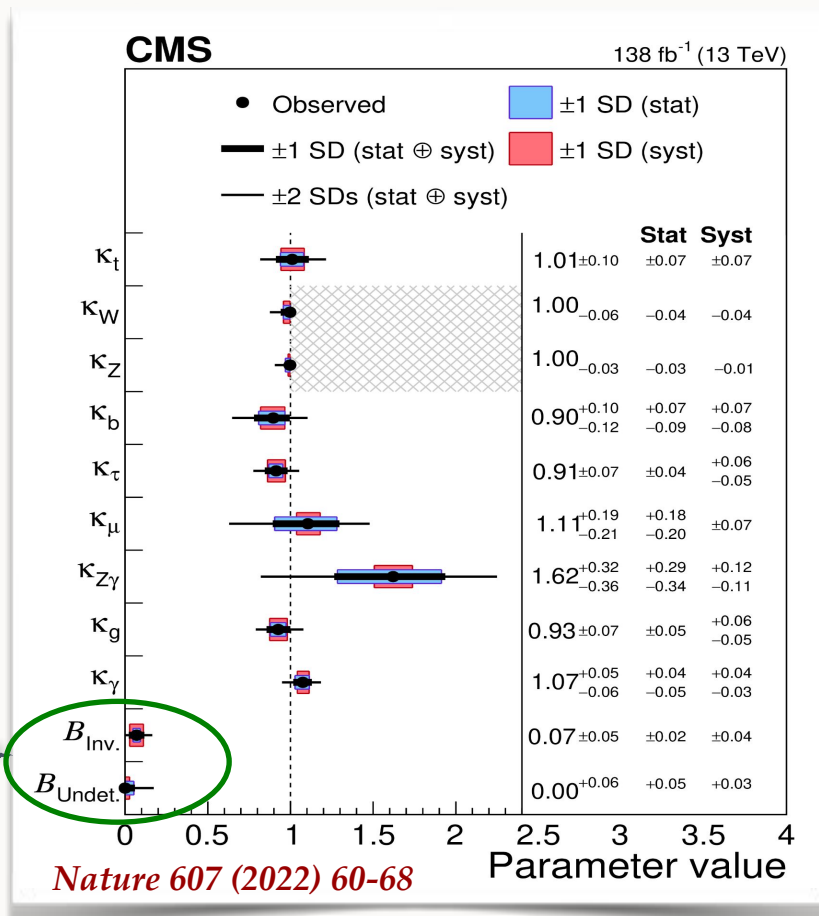
- General motivation of the search
- Analysis strategy
- Results
- Summary

General motivation to search for exotic Higgs decays

- **Journal Nature:** most comprehensive overview of the Higgs boson published by CMS and ATLAS
 - Data in agreement with SM predictions, so far
 - $\mathcal{B}_{\text{Undet.}} < 0.16$ and $\mathcal{B}_{\text{Inv.}} < 0.16$
 - There are rooms for the Higgs boson to connect with BSM



Search for new physics

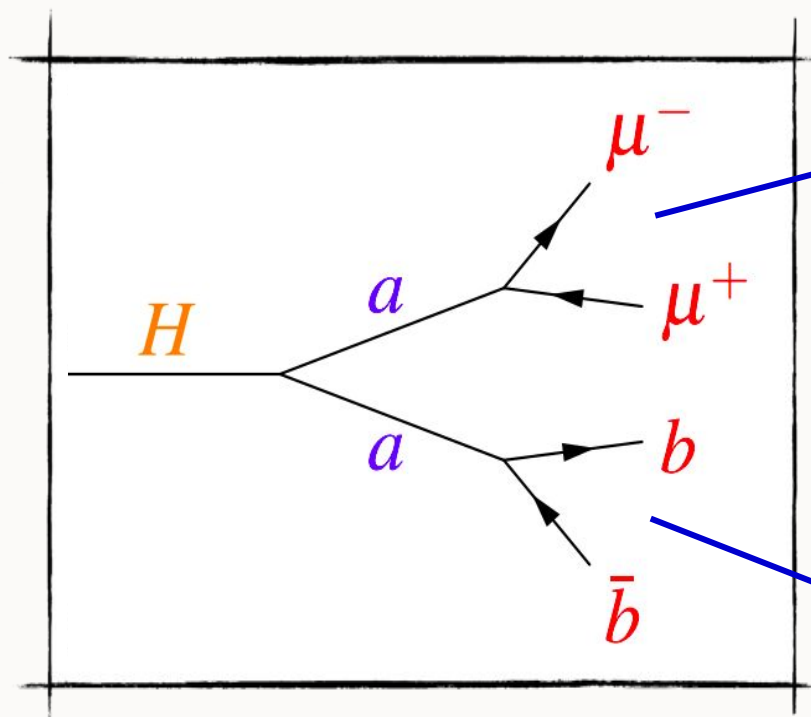


Exotic decays of the Higgs boson provide a natural and efficient way for probing new physics

Exotic Higgs decays : $H \rightarrow$ (pseudo)scalars

- Some extensions of the SM include Higgs decays via a pair of on-shell (pseudo)scalars, e.g. **2HDM+S**
- The (pseudo)scalar a decays into fermions via mixing with the Higgs boson
- $\mathcal{B}(a \rightarrow xx)$ depend on the model
- Focus especially on the decay of the pseudoscalar $aa \rightarrow \mu\mu bb$
 - **Main backgrounds:** $t\bar{t}$ and Drell-Yan
 - **Background estimation :** Fully data driven approach

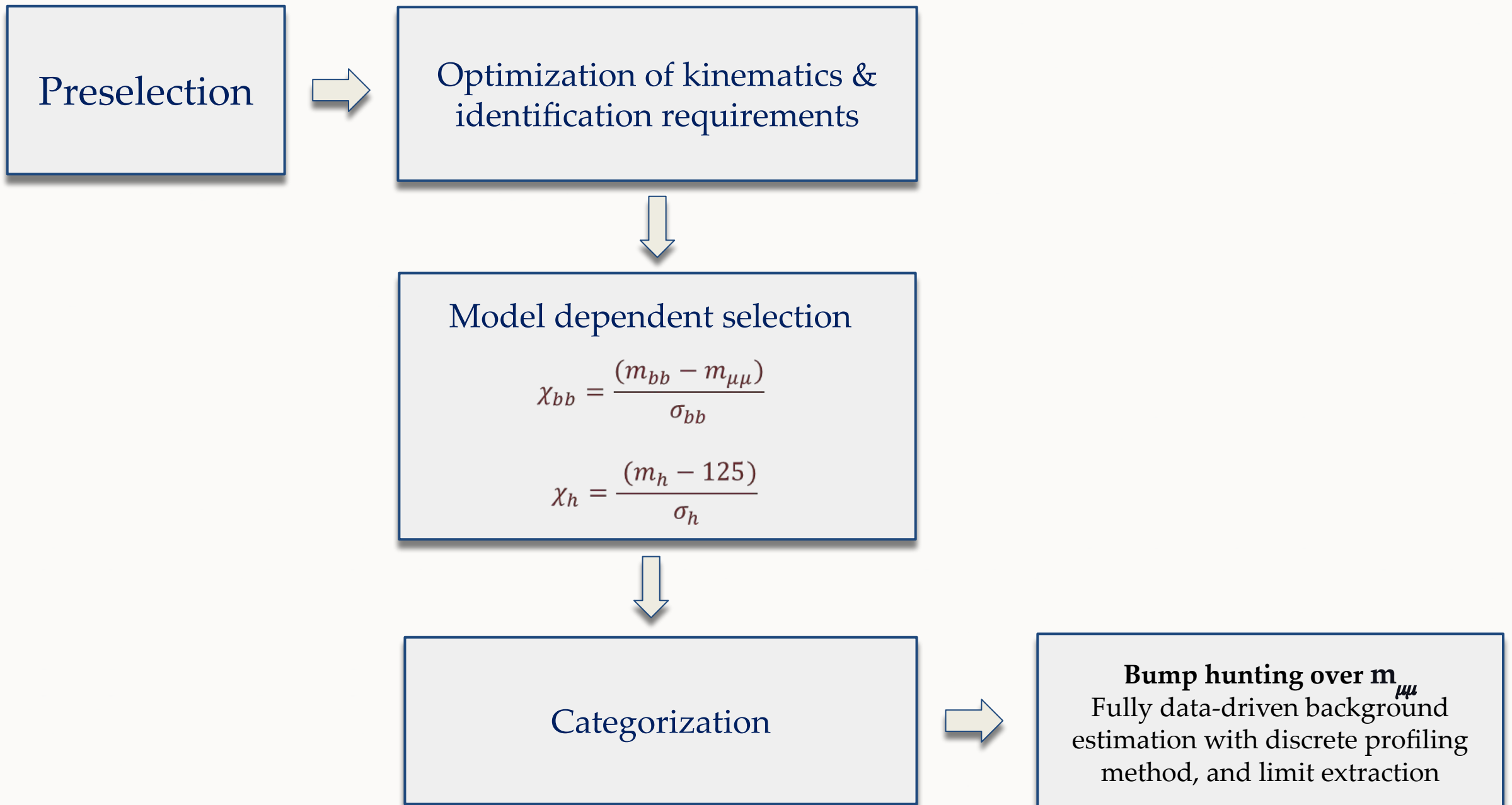
$H \rightarrow aa \rightarrow \mu\mu bb$



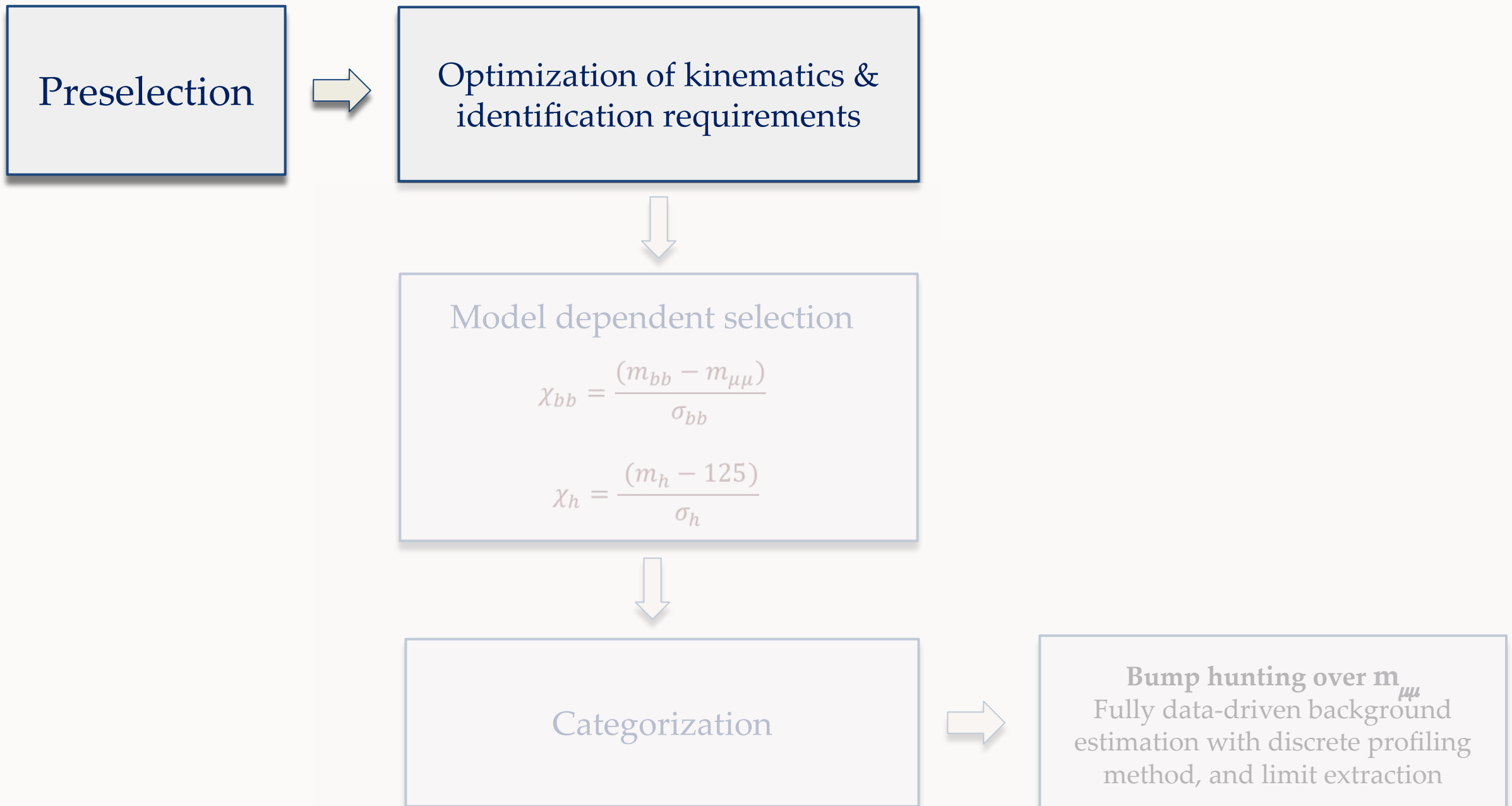
✓ Excellent mass resolution
 ✓ Easy to trigger
 ✗ Low BR

✓ Large BR
 ✗ Hard to trigger
 ✗ Large jets-backgrounds

Analysis strategy



Analysis strategy



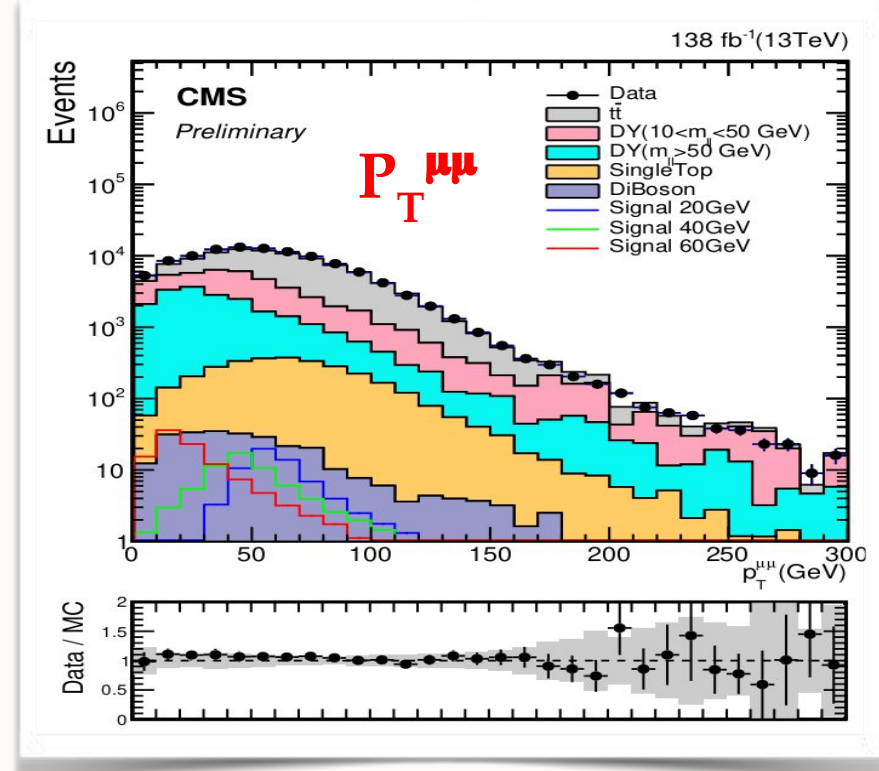
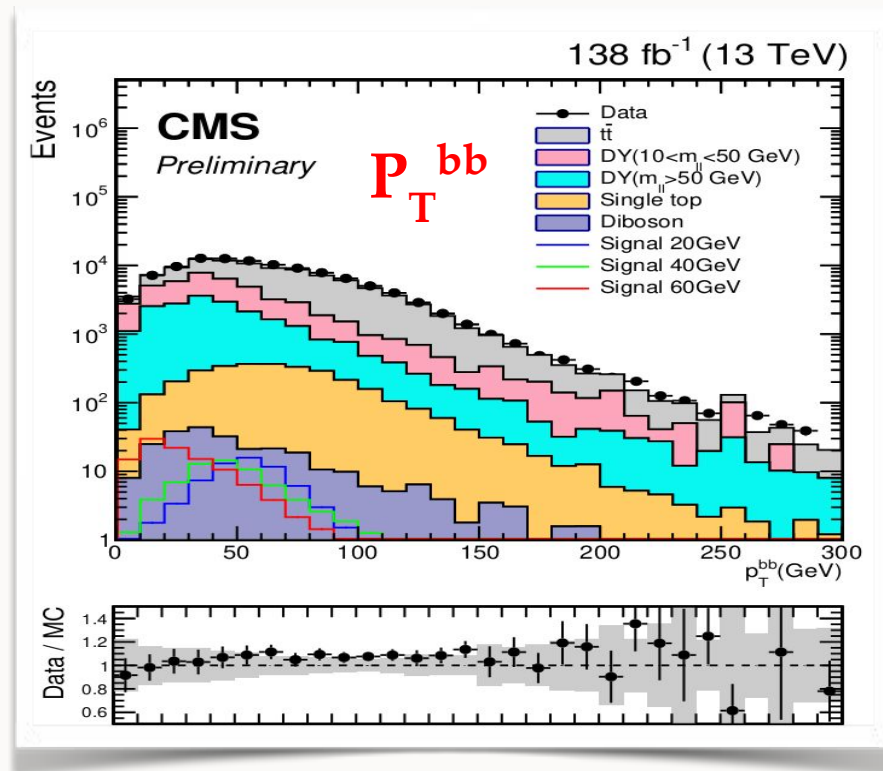
Pre-Selection :

- Required at least two b-jets :
 - $P_T > 15$ GeV
- Required 2 opposite sign Muons :
 - $P_T > (15,17)$ GeV
 - $14 \text{ GeV} < m_{\mu\mu} < 70$ GeV
- MET < 60 GeV

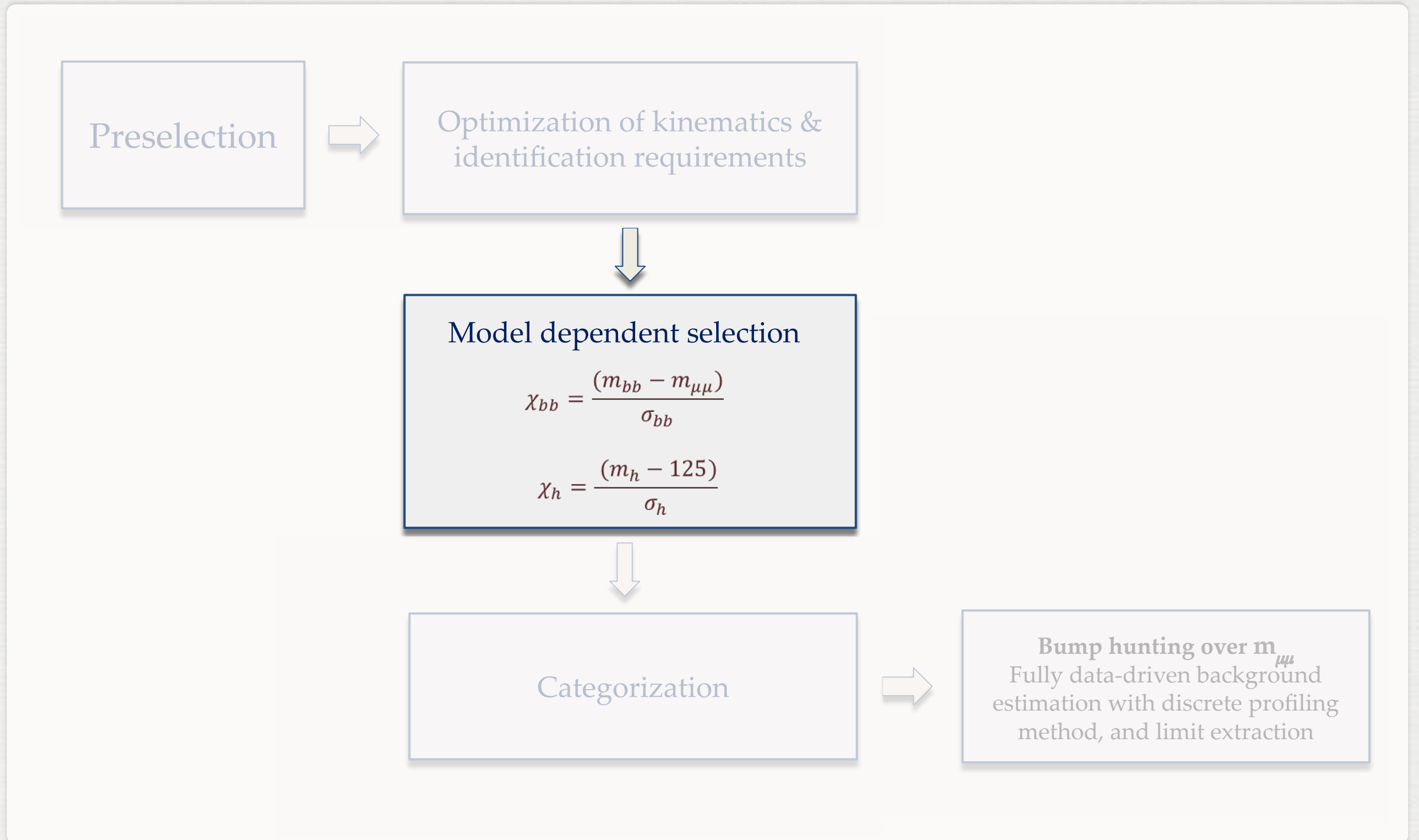
Optimization :

- Figure of merit for optimization: **significance** (S/\sqrt{B})
- Optimization:
 - **B-tagging algorithm and working points, muon Id/Iso**
- Try to have a uniform selection vs. m_a
- **Optimal selection:**
 - **Mu:** TightIdLooseIso , **balgo:** deepjet , **bwp:** TL

Data/MC comparison after optimization procedure

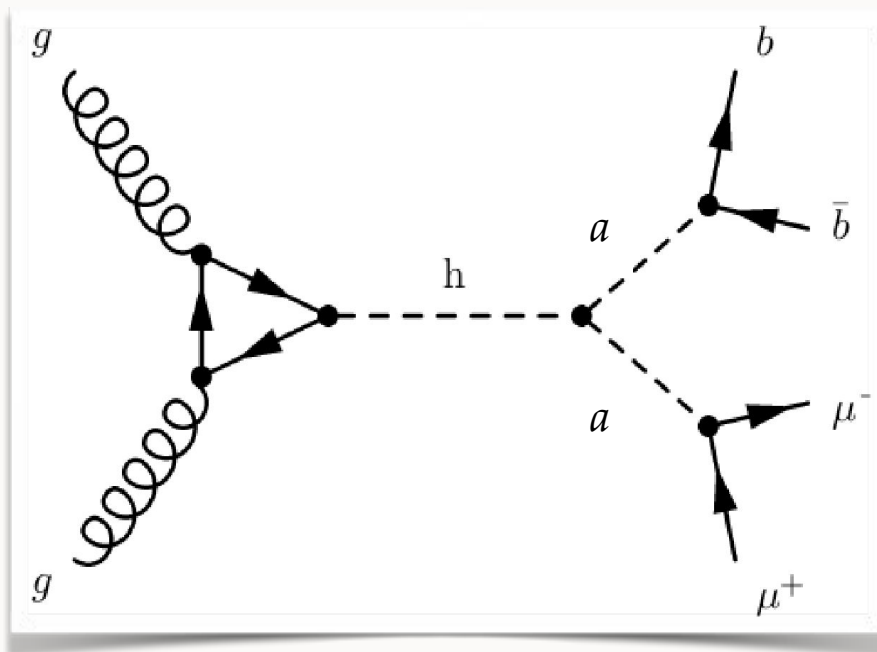


Analysis strategy



χ^2 definition and selection

- Model dependent selection : exploit two features in signal
 - $m_{bb} = m_{\mu\mu}$
 - $m_{\mu\mu bb} = 125 \text{ GeV}$

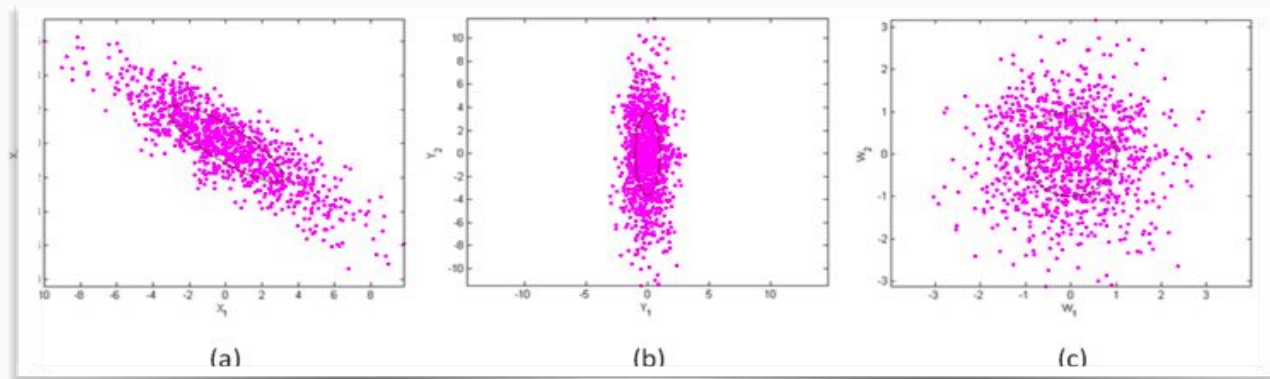


$$\chi_{bb} = \frac{(m_{bb} - m_{\mu\mu})}{\sigma_{bb}} \quad , \quad \chi_h = \frac{(m_h - 125)}{\sigma_h}$$

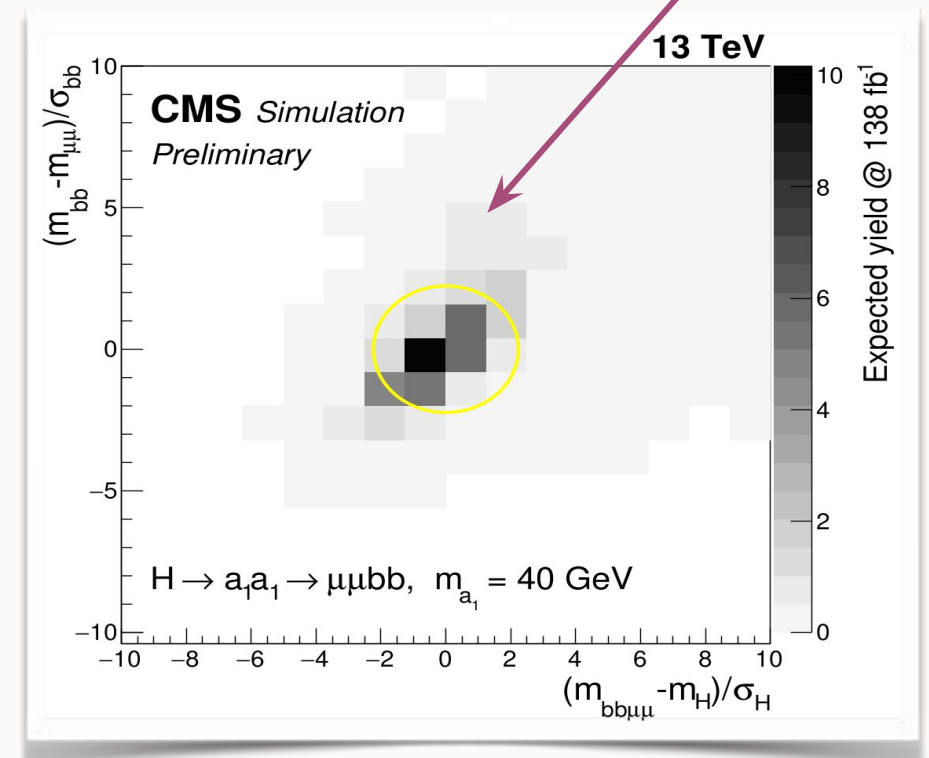
$$\chi^2 = \chi_{bb}^2 + \chi_h^2$$

- Resolutions derived from a Gaussian fits to m_{bb} and $m_{\mu\mu bb}$, where objects are matched to the MC truth
- σ_{bb} changes as a function of m_a , consider this change in χ_{bb} calculation
- χ^2 : discriminator variable between signal and backgrounds

- Looks like an ellipse and depends on m_a
 - placing a circular cut ($\chi^2 < X$) is suboptimal
- χ_{bb} and χ_h must be decorrelated using PCA method
- New variable χ_d^2 is defined : $\chi_d^2 = \chi_{b,d}^2 + \chi_{h,d}^2$



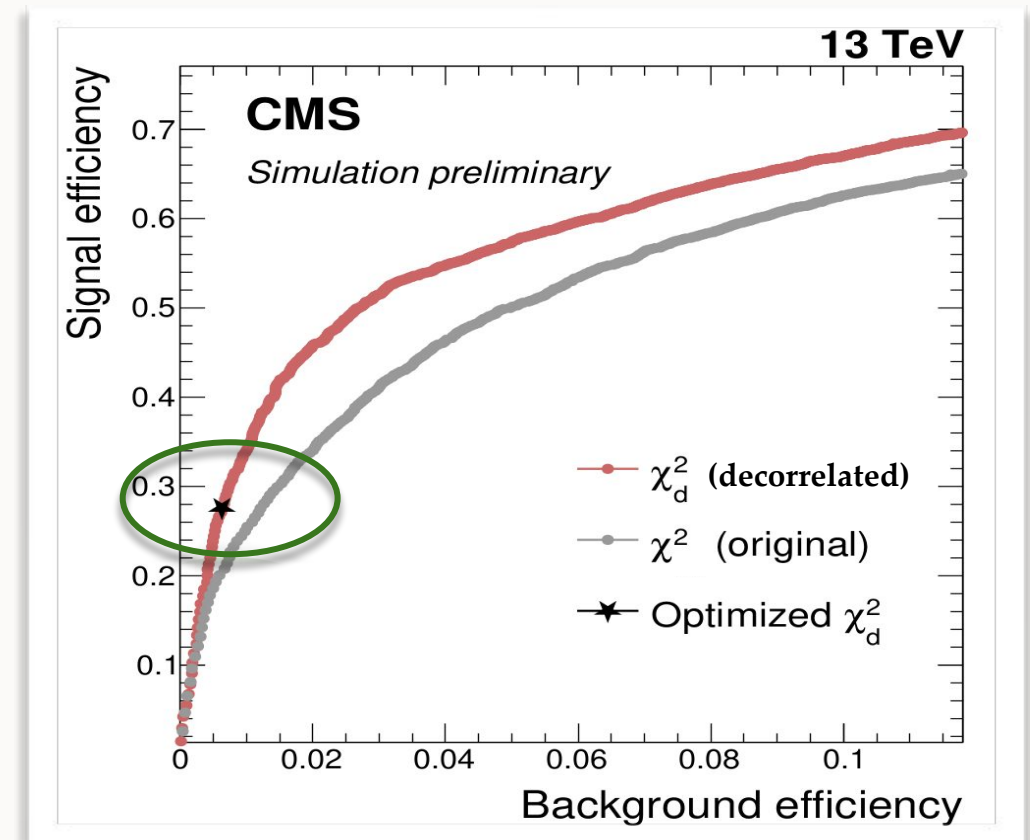
Two variables are correlated



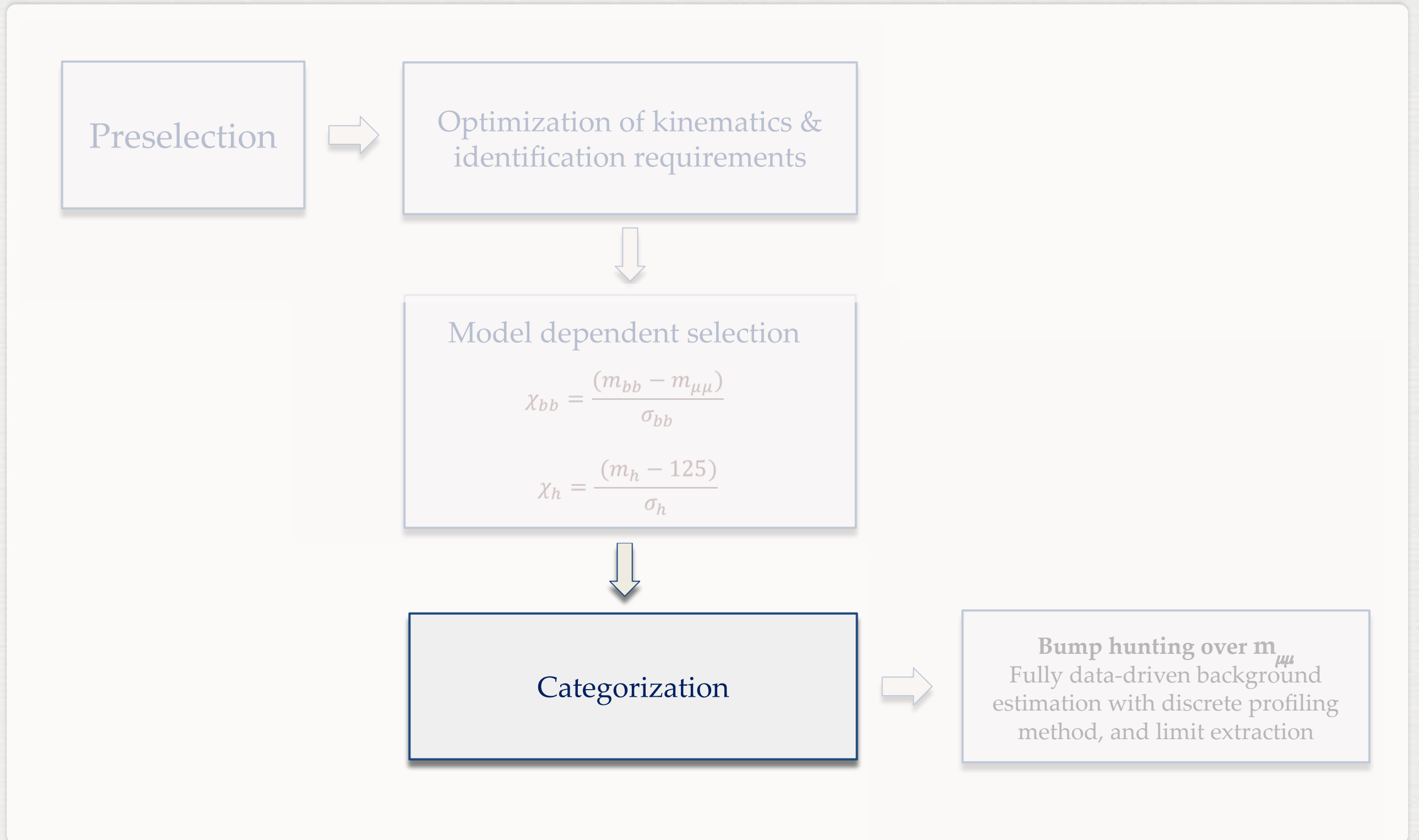
Improved performance of the χ^2_d and χ^2

- Optimized cut value : $\chi^2_d < 1.5$
- A large fraction of the backgrounds rejected, while most of the signal remains
- Comparison of the selection performance of the χ^2_d and χ^2 in terms of efficiencies for signal ($m_a = 40$ GeV) and backgrounds

A clear improvement is visible after decorrelation



Analysis strategy



Categorization

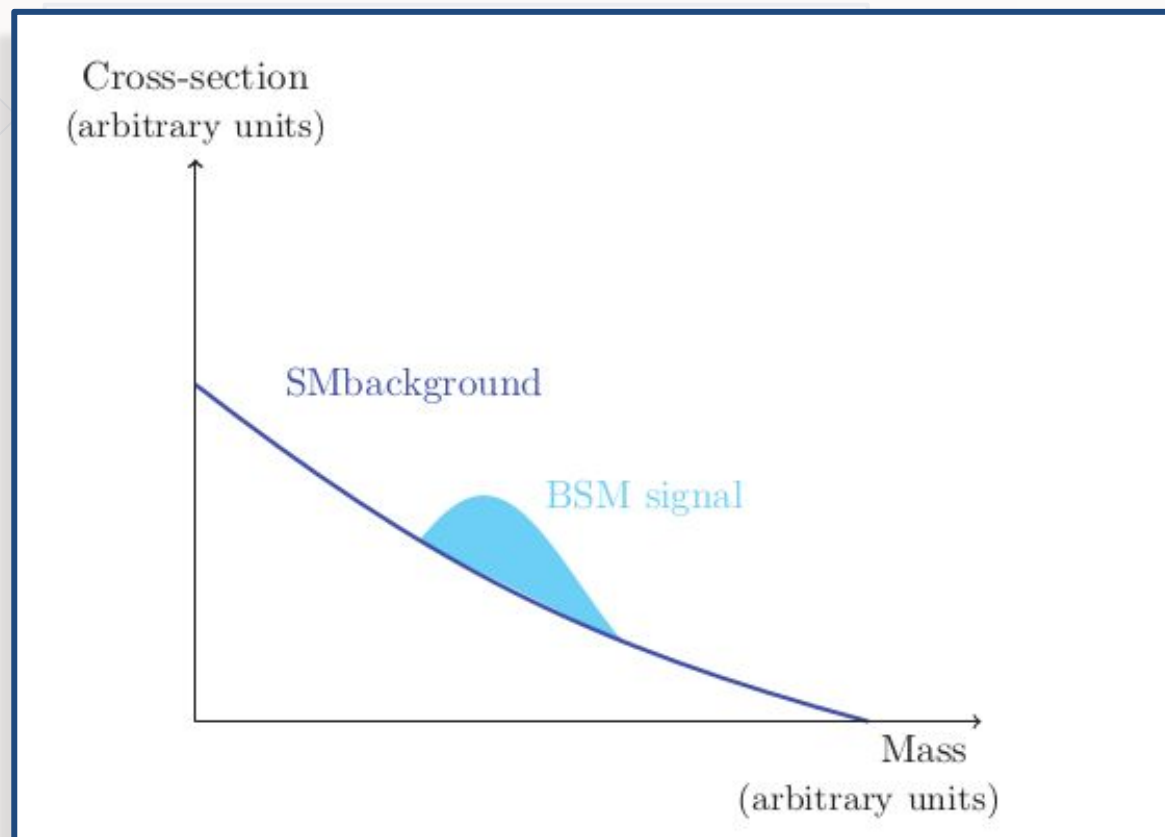
- A categorization of events is performed, to increase the sensitivity of the analysis
- Events are categorized according to the **b-jet P_T , event's compatibility with VBF Higgs production, and the score of the b-jets**
- The **Low P_T** category brings extra sensitivity to the signal with lower m_a values.

Categories for selected events	
Low P_T	At least one b-jet with $P_T < 20$ GeV
VBF	Two add. Jets with $P_T > 30$ GeV, $ \eta < 4.7$, and $m_{jj} > 250$ GeV
TL	Looser b-jet passes L but fails M
TM	Looser b-jet passes M but fails T
TT	Looser b-jet passes T

- The majority of background events ($\approx 70\%$) fall into the **Low P_T** category
- For signal (ggH): $\sim 40\%$ in **Low P_T**

Input models and limits

Preselection



While blind, defining a control data region to model the background for expected limits

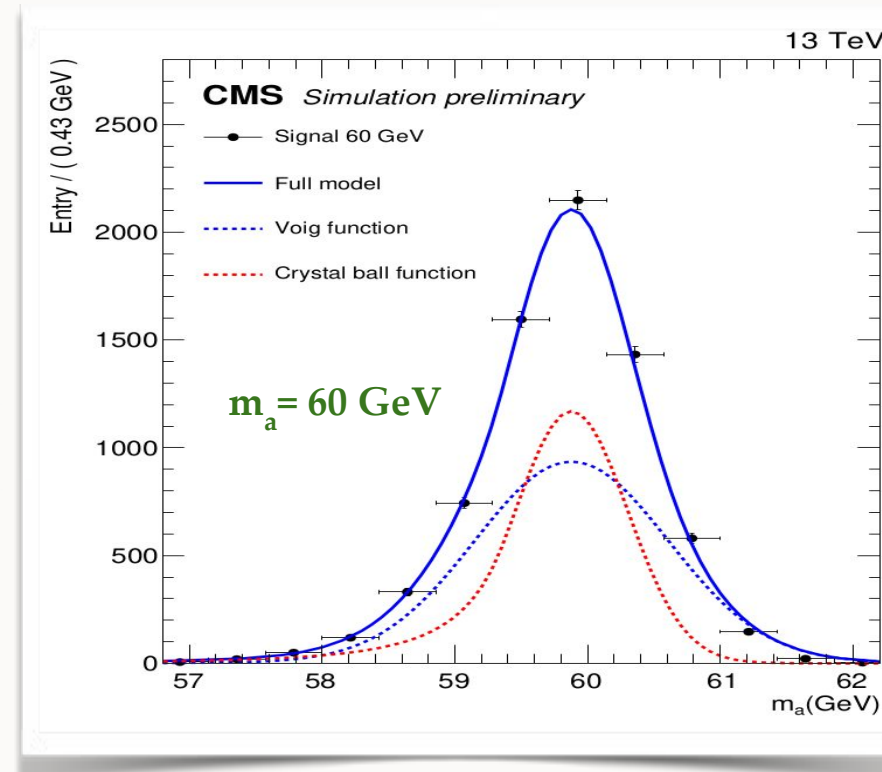
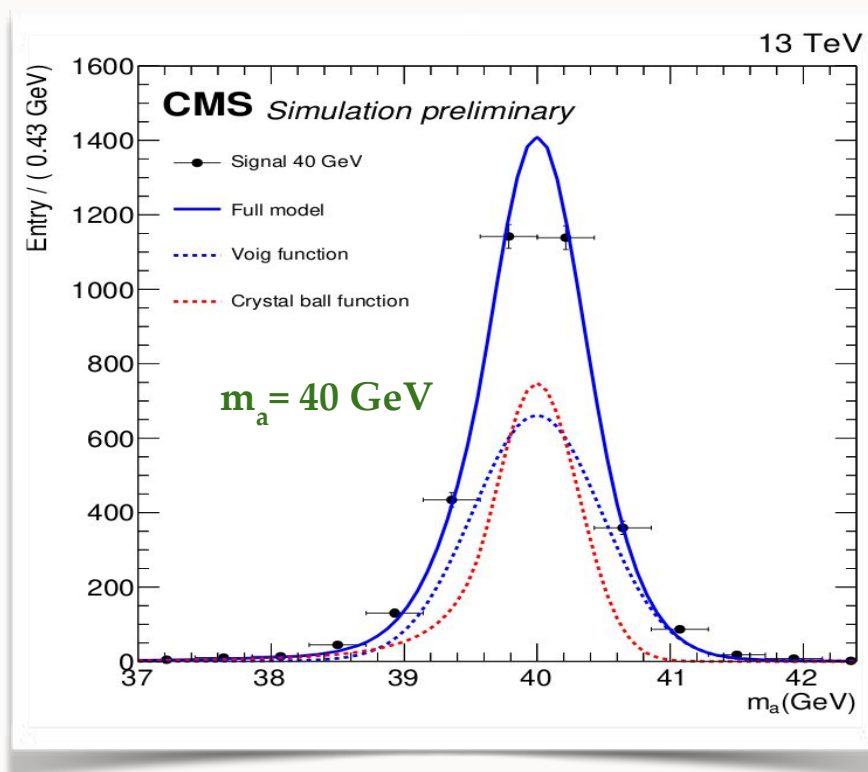
Categorization based on b-tagging score of b-jets

Bump hunting over $m_{\mu\mu}$
Fully data-driven background estimation with discrete profiling method, and limit extraction

Signal Shape Modeling (*a la* HIG-18-011)

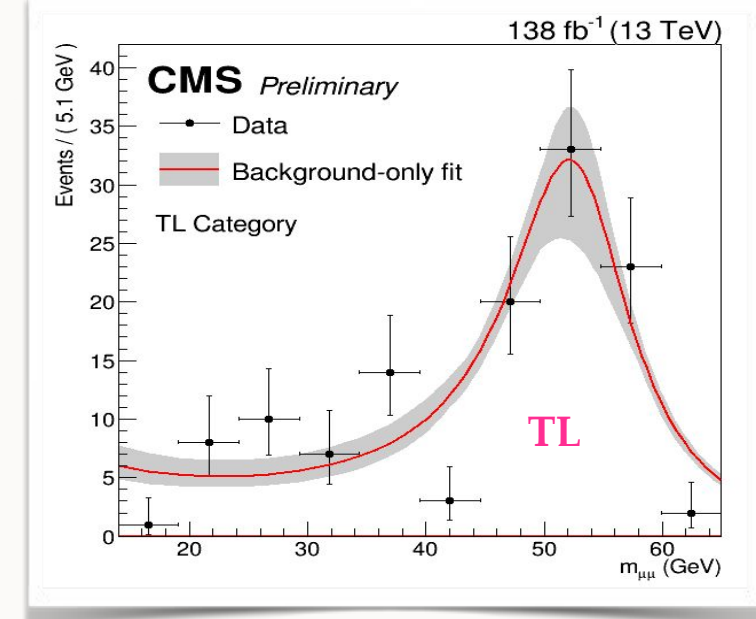
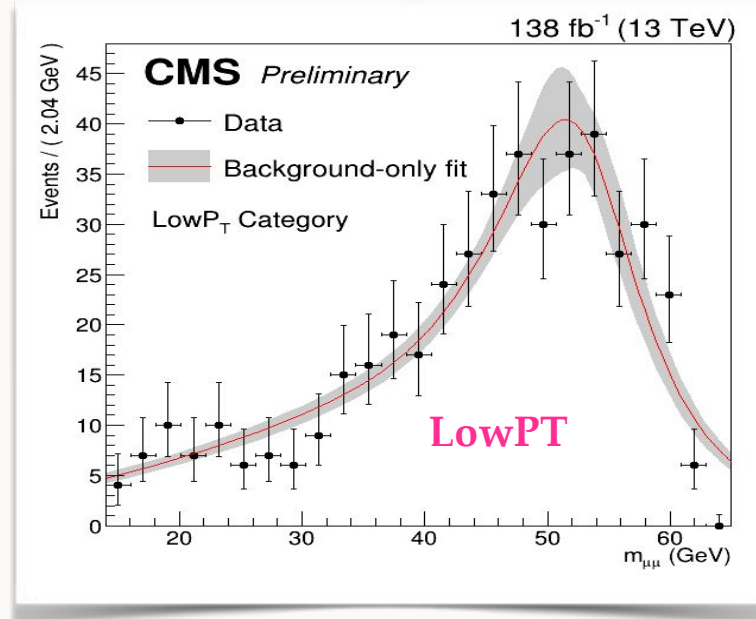
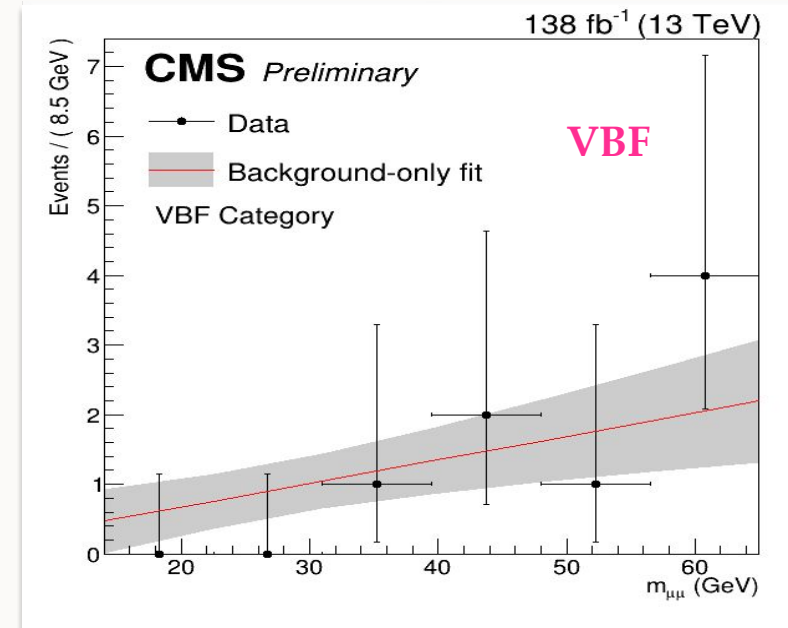
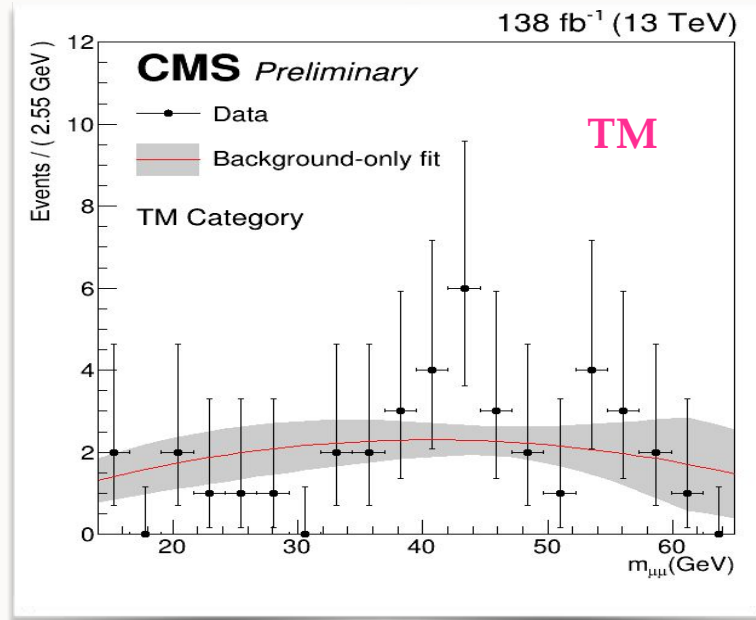
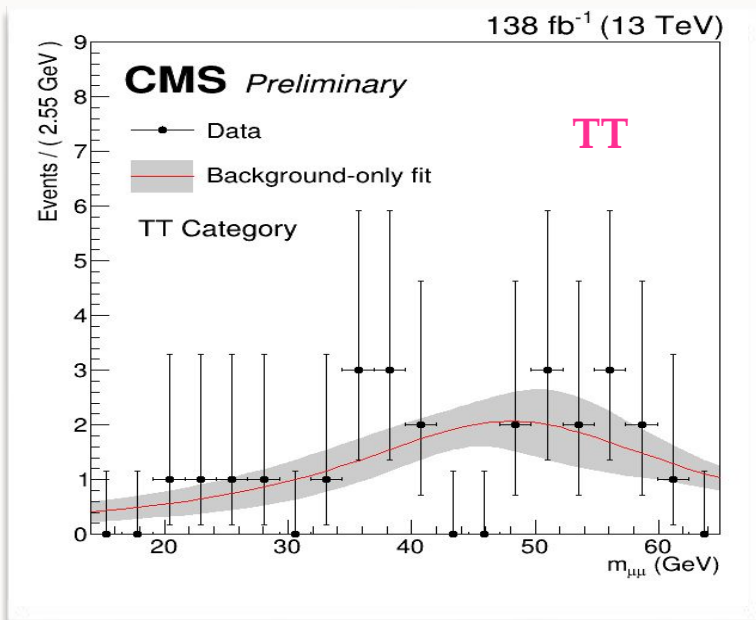
- The shape of signal is estimated from simulation
- Signal is modeled with a combination of a **Voigt** and a **CrystalBall (CB)** profile.

$$S(m_{\mu\mu}|f, p_V, p_{cb}) \equiv f \cdot V(m_{\mu\mu}|p_V) + (1 - f) \cdot CB(m_{\mu\mu}|p_{cb})$$



Background Modeling

background-only fit results

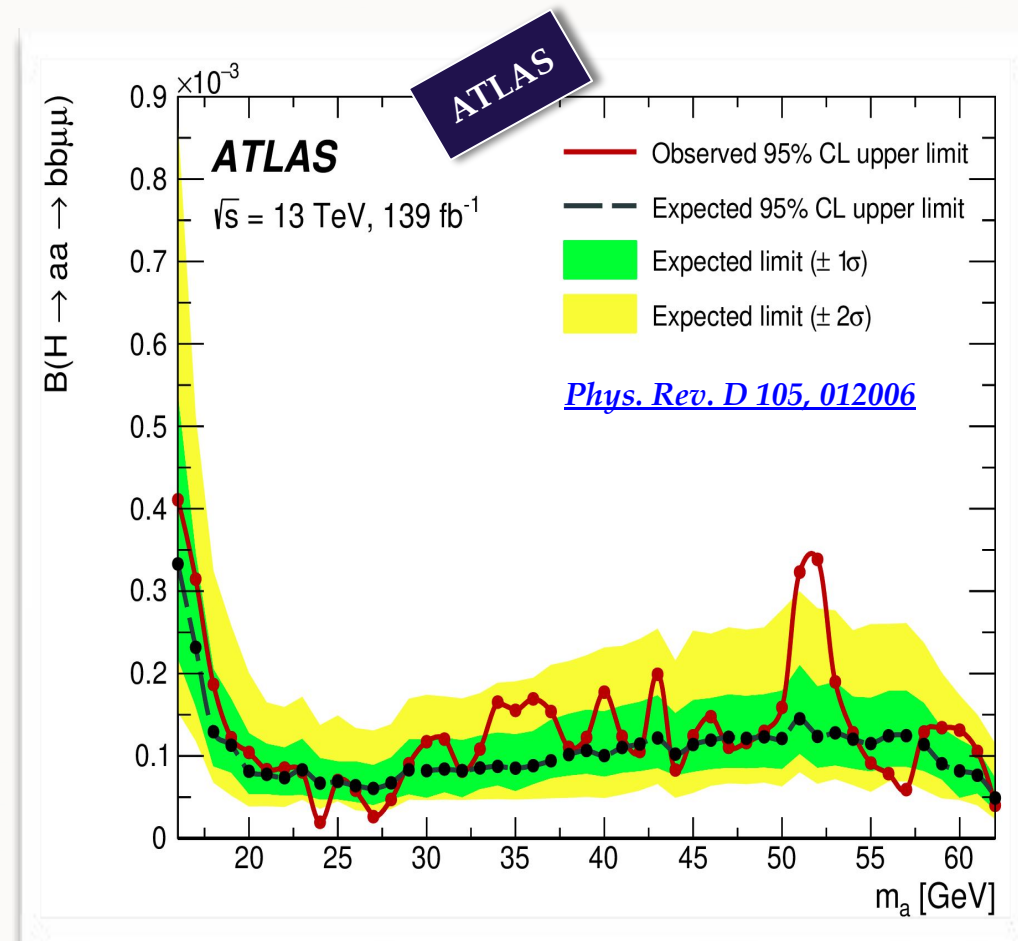
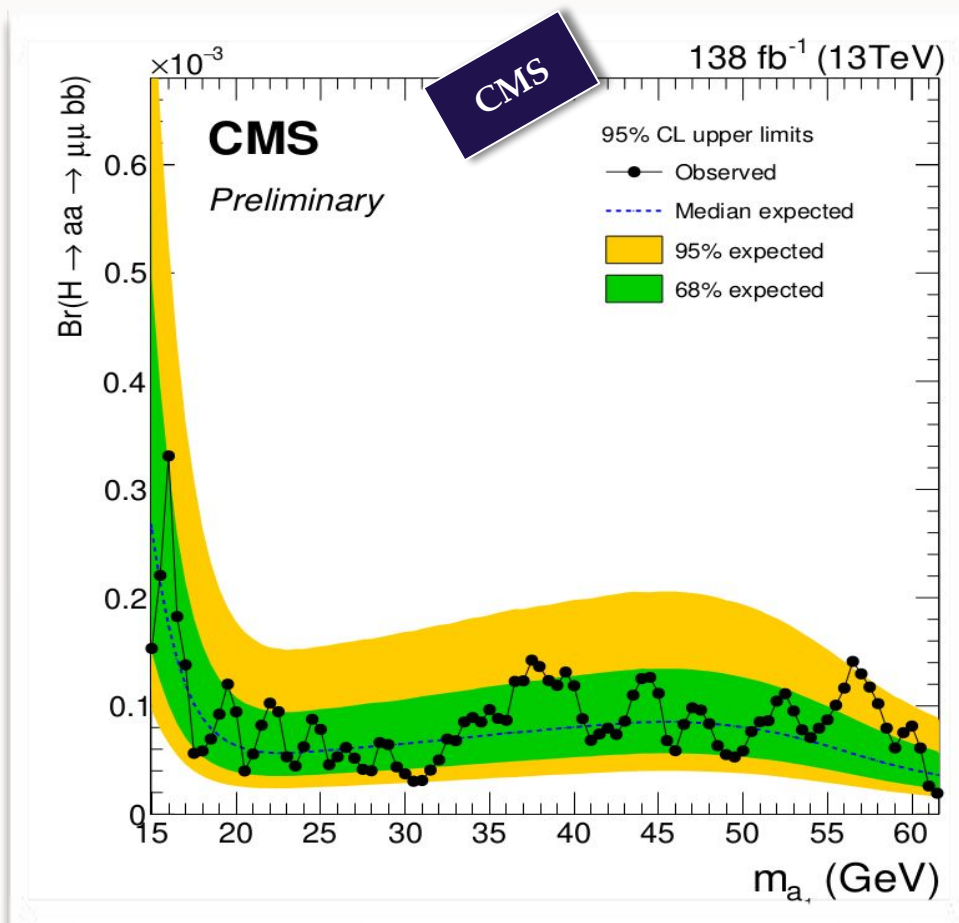


- Estimated by a fully data-driven approach, using envelope method
- Results on background only fit for LowP_T, TM, TT, TL and VBF categories
- The uncertainty bands correspond to the best-fit background model uncertainty as extracted from the fit to the data

Results

Observed/ Expected Limits

- At 95% CL, upper limits on $\mathcal{B}(H \rightarrow aa \rightarrow \mu\mu bb)$ for the mass range 15 to 62.5 GeV:
 - **Observed:** $(0.17 - 3.3) \times 10^{-4}$
 - **Expected:** $(0.35 - 2.6) \times 10^{-4}$
- Large improvement with respect to 2016 beyond the increase of luminosity



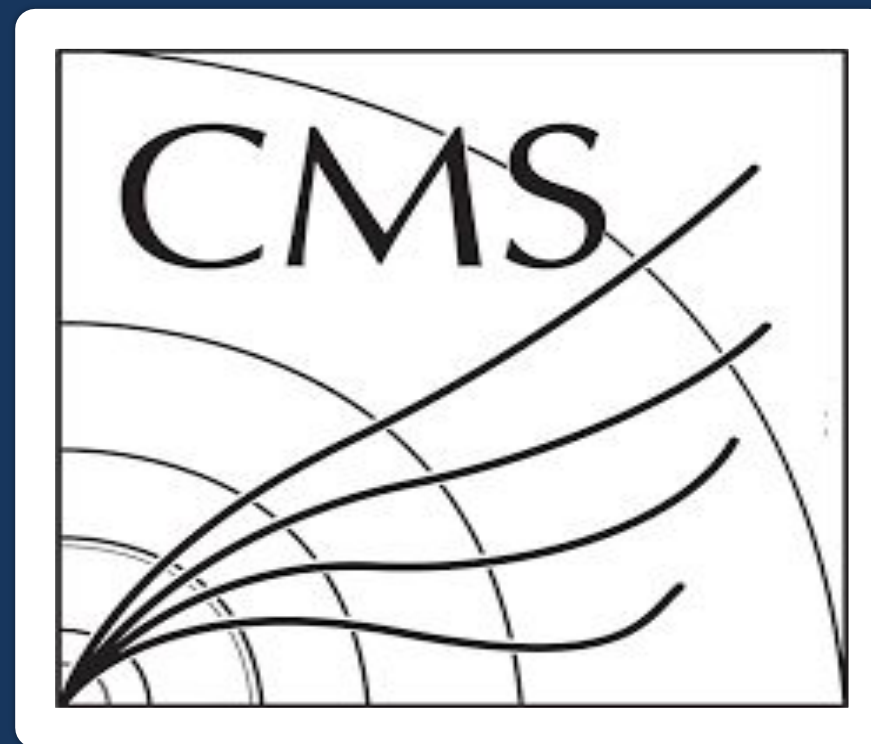
Latest ATLAS results :

$$\mathcal{B}(H \rightarrow aa \rightarrow \mu\mu bb) < \text{obs. } (0.22 - 4) \times 10^{-4} \text{ exp. } (0.5-5) \times 10^{-4}$$

Summary

- Search for exotic Higgs decays in $\mu\mu b\bar{b}$ final state
- Full Run 2 dataset is analyzed
- No excess is found over the SM backgrounds
 - Upper limits are reported on $\text{Br}(H \rightarrow a\bar{a} \rightarrow \mu\mu b\bar{b})$
- Improved analysis strategy resulted in better expected and observed limits compared with 2016 analysis, beyond the effect of luminosity increase
- Looking forward to surprises in LHC Run 3!

“Everything is a mathematical trick,
except what you measure in the lab.”
(Prof. Dr. Armin Scrinzi)



Thank you for your attention !

BACKUP

2HDM:

- 2HDM is an extensions of the SM in the Higgs sector
- Separated in 4 types, depending on how the SM fermions interact with the two Higgs doublets
- After EWKSB, the 2HDM leads to 5 physical states:
 - Charge scalar pair H_{\pm} ,
 - Neutral pseudoscalar A ,
 - Neutral scalar H_0 ,
 - Neutral scalar h , $m_h < m_{H_0}$, can be SM-like

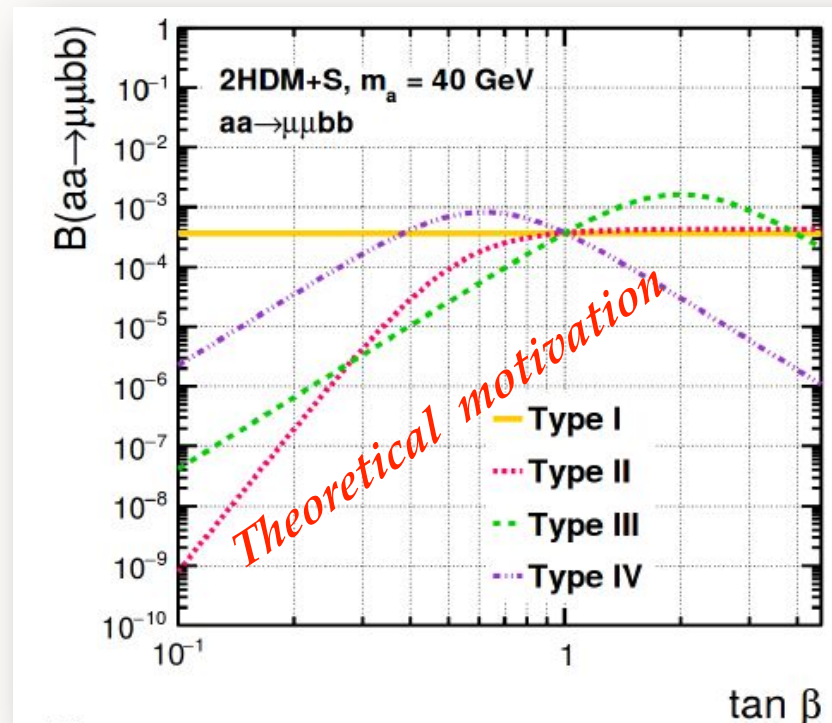
	Type-I	Type-II	Type-III	Type-IV
l	φ_2	φ_1	φ_1	φ_2
u	φ_2	φ_2	φ_2	φ_2
d	φ_2	φ_1	φ_2	φ_1

2HDM+S:

- 2HDM has been constrained from experimental data, one way is to extend the 2HDM
- We consider the extensions of 2HDM (**2HDM+S**) that has **NMSSM** as a special case
- A complex scalar singlet is added to the already present scalar doublets.
- Because of the additional singlet, two new bosons are introduced: $s(a)$ (*pseudo*)scalar
- h boson can decay to fermions through $s(a)$

LHC data are used to search for this exotic decay and set limit on $\text{Br}(h \rightarrow aa \rightarrow ff)$

The $\mu\mu bb$ final state



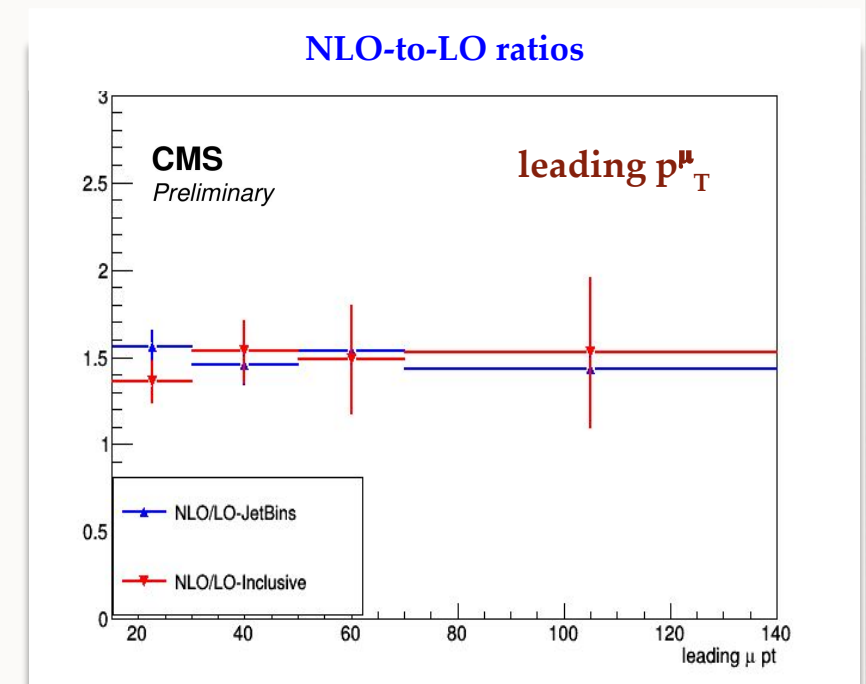
$$aa \rightarrow \mu\mu bb$$

The largest BR($aa \rightarrow \mu\mu bb$) is obtained for $\tan \beta > 2$ in 2HDM+S type III.

- Summary of simulated background samples :

Samples	2016 Dataset	2017 Dataset	2018 Dataset
Drell-Yan ($10 < m_{\mu\mu} < 50$ GeV)	NLO, LO sample exclusive in number of additional partons (up to four)	Leading Order	Leading Order
Drell-Yan ($m_{\mu\mu} > 50$)	NLO, exclusive in number of additional partons (up to two)		
$t\bar{t}$	Next to Leading Order		
Single Top	Next to Leading Order		
Diboson	Leading Order		

- The NLO low-mass Drell-Yan samples are only available in 2016
 - A k-factor of 1.5 is extracted from 2016 samples and applied in all years.
 - An uncertainty of 30% is considered on the k-factor.
- **Note:** Background samples are used only for the selection optimization purposes
- The final background contribution is determined fully based on data with no reference to MC



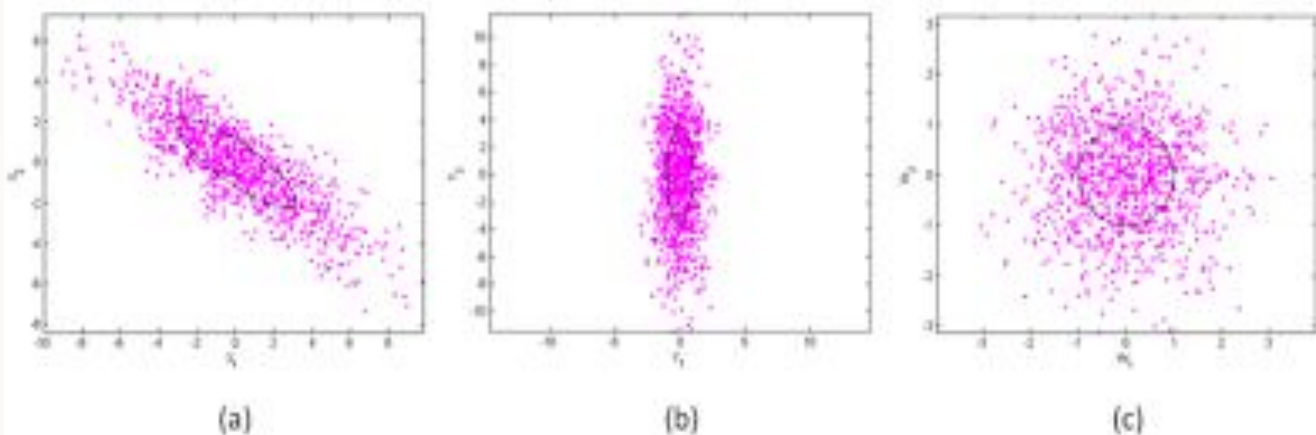
Principal component analysis (PCA) as a decorrelation method

$$\begin{pmatrix} \chi_h \\ \chi_{bb} \end{pmatrix} \longrightarrow \Sigma = \begin{pmatrix} v(\chi_h) & \text{cov}(\chi_h, \chi_{bb}) \\ \text{cov}(\chi_{bb}, \chi_h) & v(\chi_{bb}) \end{pmatrix} \longrightarrow \begin{cases} E0 = \begin{pmatrix} a \\ b \end{pmatrix}, \lambda_0 \\ E1 = \begin{pmatrix} -b \\ a \end{pmatrix}, \lambda_1 \end{cases}$$

- (Σ) decomposed as a sequence of rotation and scaling operations on uncorrelated data (white)
 - **(E)**: Rotation matrix defined by the **eigenvectors** of Σ .
 - **(C)**: Scaling matrix defined by the **eigenvalues** of Σ

$$\begin{cases} C = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \\ E = \begin{pmatrix} a & -b \\ b & a \end{pmatrix} \end{cases}$$

$$W = C^{-1/2} E^T \quad W = \begin{pmatrix} a/\sqrt{\lambda_1} & b/\sqrt{\lambda_1} \\ -b/\sqrt{\lambda_2} & a/\sqrt{\lambda_2} \end{pmatrix}$$



$$\begin{cases} \chi_{hd} = \frac{a}{\sqrt{\lambda_1}} (\chi_h + \chi_{bb} \tan \alpha) \\ \chi_{bbd} = \frac{a}{\sqrt{\lambda_2}} (-\chi_h \tan \alpha + \chi_{bb}) \end{cases}$$

Refers to decorrelated variables

- **Signal yields:** for non-simulated mass points are estimated by extrapolation, using SPLine
 - Benchmark: $\text{Br}(H \rightarrow aa) = 10\%$, $2\text{Br}(a \rightarrow bb)\text{Br}(a \rightarrow \mu\mu) = 1.7 \times 10^{-3}$
- **Background** is ultimately determined from **data** (envelope method)

Categorization

- For signal (ggH): **~40%** in **LowP_T**
- Different ratios of signal samples fall into each category.
- The majority of background events (**~70%**) fall into the **LowP_T** category
- (**~20%**) of background events fall into the **TL** category, about **10%** pass the **TM** category, less than **10%** can meet the **TT** criteria
- For signal (vbF) : about **40%** in this category

Background:

- Estimated by a fully data-driven approach, using discrete profiling method (envelope method)
- Systematic uncertainties on background modeling are taken into account

Signal:

- **Signal shape:**
 - Estimated by a parametric signal model (MC-based)
 - Uncertainties on signal model parameters are found to be negligible (2016, RunI paper)
- **Signal normalization:**
 - Is affected by various sources of systematic uncertainties:

- **Modeling**
 - PS
 - QCD Scales
 - PDF
 - **Muon**
 - ID/Iso SF
 - HLT: 1%
 - **Jets**
 - btag SF
 - JES/JER
 - **Other sources**
 - Prefire
 - PU
 - Lumi: Followed LUM POG recipe
 - Xsection : ggH+VBF, considered for the limit on BR

Uncertainties on signal shape are negligible, their effects on yields are taken into account as nuisance parameters in the fit

Background Modeling

- **Estimated by a fully data-driven approach, using envelope method**
- Evaluated through a fit to the $\mu\mu$ mass distribution in data w/o any reference to MC
- Different types of functional forms are used on data in SR
 - RooPolynomial , RooBernstein, RooChebychev, Inverse polynomial ($1/P_n$)
 - All of them are considered in the extracted limit
- A **F-test** is used to determine the collection of pdfs for each family
- Profile likelihood method is performed using combine
- Results on background only fit for Low P_T , TM, TT, TL and VBF categories.
- The uncertainty band corresponds to the best-fit background model uncertainty as extracted from the fit to the data.

Optimization procedure/results

- Muon working point: **Tight ID/ Loose Iso**
- b-tagging algorithm: **DeepJet** (It outperforms the other taggers)
- The most optimum b-tagging working point :
 - **One Tight/ One Loose**
 - Similar to 2016 published paper
 - Similar working point is optimized for 2016/2017

Optimum selection: Mu: TightIdLooseIso , balgo: deepjet , bwp: TL