# JHU generator framework: new features for Higgs boson studies

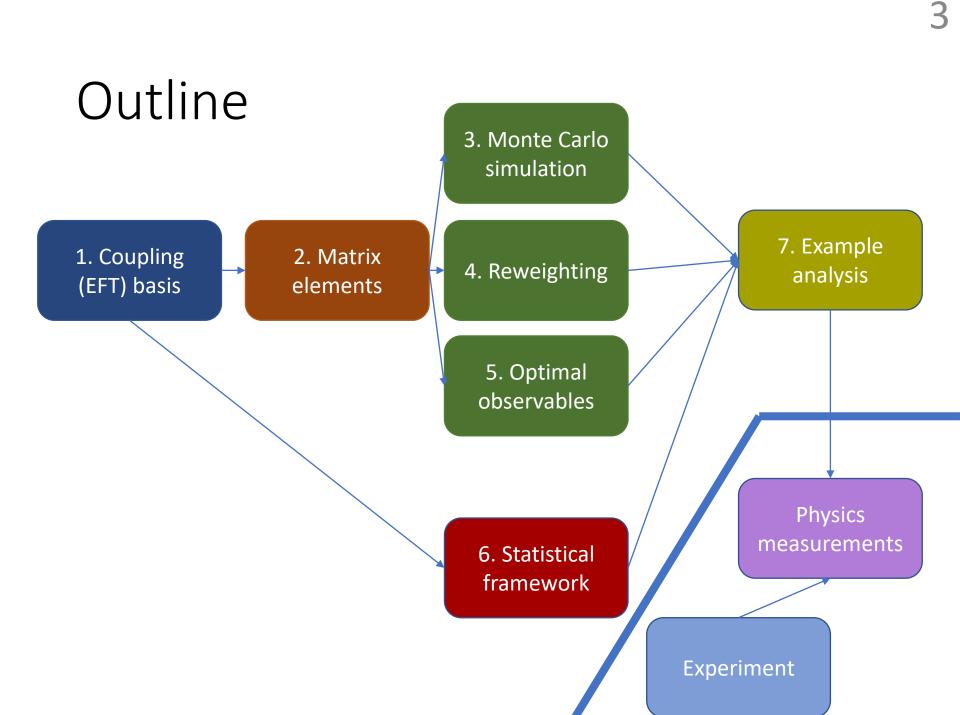
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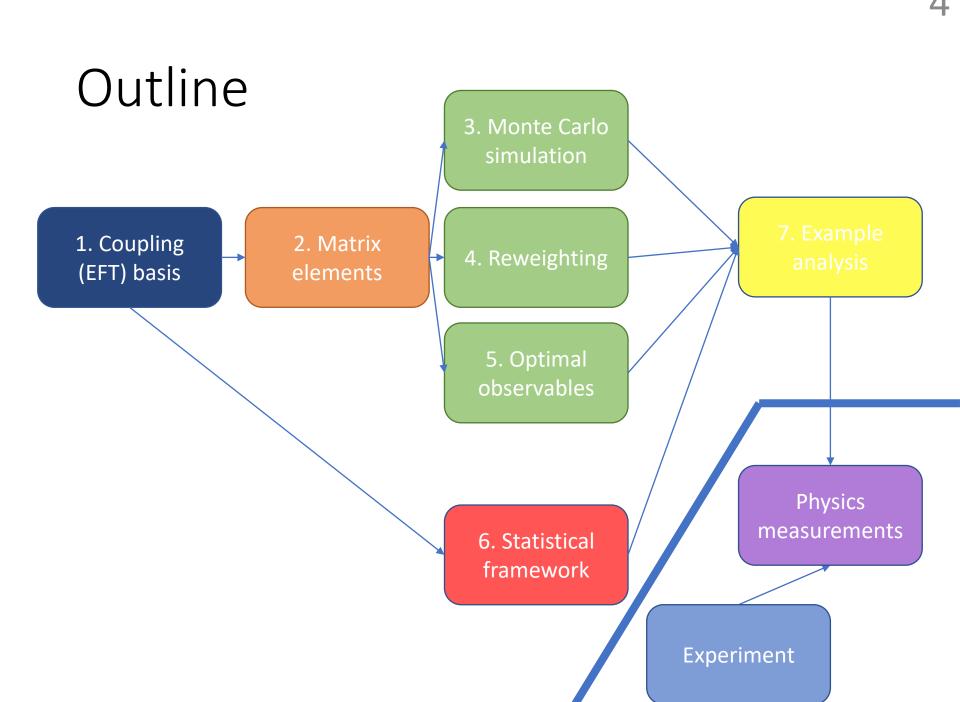
Pheno 2020 <del>Pittsburgh, PA</del> Your house May 4, 2020

https://arxiv.org/abs/2002.09888

#### JHU Generator and MELA package

- JHU Generator
  - Designed to simulate a wide range of processes involving spin 0, 1, 2 particles with a general coupling model
- JHUGen MELA—Matrix Element Likelihood Approach
  - Calculate discriminants to optimally isolate processes or operators
  - Reweight generated samples from one hypothesis to another
- <u>https://spin.pha.jhu.edu/</u>

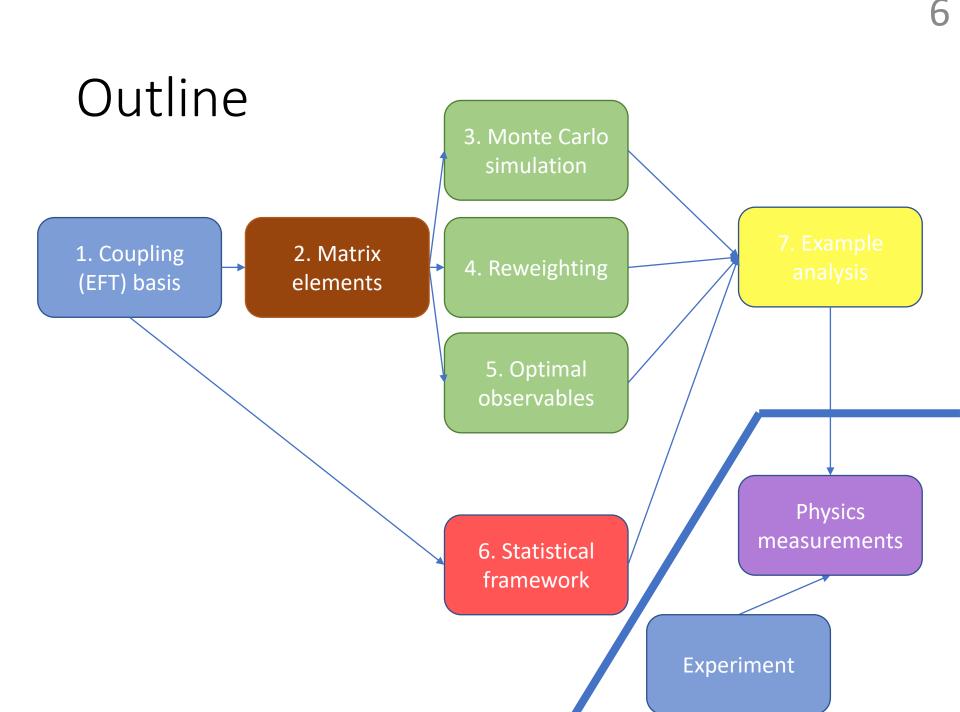




## SM Coupling parameterization CP-odd • $A(HVV) = \frac{1}{v} \left[ g_1^{VV} + \frac{\kappa_1^{VV} q_1^2 + \kappa_2^{VV} q_2^2}{(\Lambda_1^{VV})^2} + \frac{\kappa_3 (q_1 + q_2)^2}{(\Lambda_Q^{VV})^2} \right] m_{V1}^2 (\epsilon_1 \cdot \epsilon_2) + 2g_2^{VV} [(q_1 \cdot q_2)(\epsilon_1 \cdot \epsilon_2) - (\epsilon_1 \cdot q_2)(\epsilon_2 \cdot q_1)] - 2g_4 \epsilon^{\epsilon_1 \epsilon_2 q_1 q_2}$

• 
$$A(Hff) = -\frac{m_f}{v} \bar{\psi}_f (\kappa_f + i\tilde{\kappa}_f \gamma_5) \psi_f$$

- General coupling model easy to relate to other models, e.g. Higgs basis of EFT
  - See paper for details of the relationship
  - E.g.,  $c_{zz} \sim -g_2^{ZZ}$ ,  $c_{z\Box} \sim -\kappa_1^{ZZ}$ ,  $\tilde{c}_{zz} \sim -g_4^{ZZ}$
  - EFT implies relationships between couplings, e.g.  $g_2^{WW}$  is a linear combination of  $g_2^{ZZ,ZY,YY}$



#### Matrix elements

Propagator

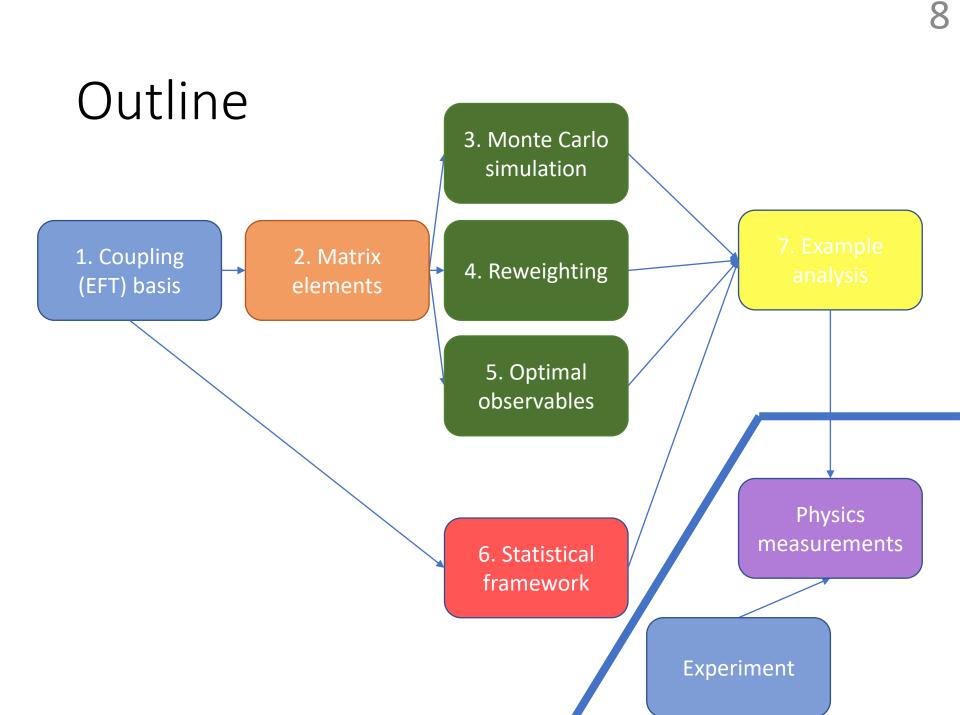
Production

Decay

- Characterize the probability of an event based on kinematics  $\Omega$ 
  - Based on couplings  $\vec{a}$
  - $\vec{a}$  includes  $g_i$ ,  $\kappa_i$  for bosons or  $c_i$  in EFT,  $\kappa$  and  $\tilde{\kappa}$  for fermions

$$\frac{d\sigma}{d\vec{\Omega}}(i \to H \to f) \\ \sim \frac{\left(\sum_{i,j} a_i a_j \alpha_{i,j}^{(i)}(\vec{\Omega}^{(i)})\right) \left(\sum_{i,j} a_i a_j \alpha_{i,j}^{(f)}(\vec{\Omega}^{(f)})\right)}{(s - m_H^2)^2 + m_H^2 \Gamma^2}$$

- Usually components factorize can calculate each one separately
- (Except when calculating interference between signal and background)

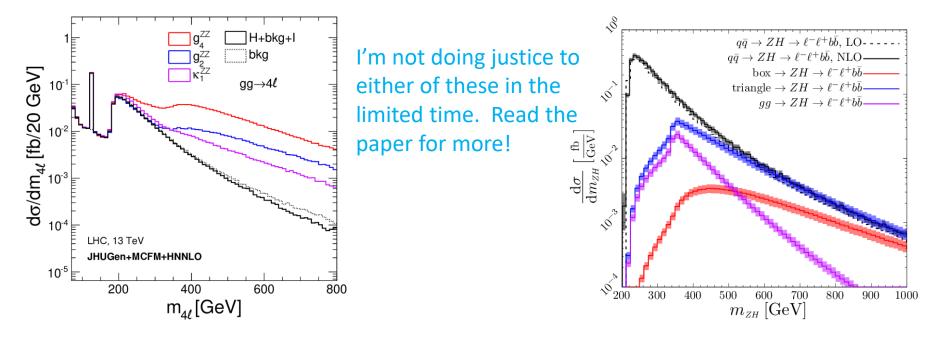


#### Sample generation and reweighting

- Generate samples according to the coupling model presented earlier
- Output in LHE format: can interface to parton showering and detector simulations
- Use the matrix elements to *reweight* samples to each other to increase statistics
  - Not an approximation: if the same matrix element is used for generation and reweighting, the results are equivalent (in the limit of infinite statistics)
  - Also needed for practical reasons when analysis needs hundreds of samples, as is the case here

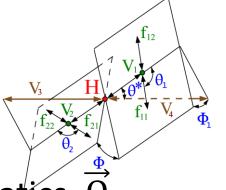
# Special features: off-shell generation and NLO VH production

- Gluon fusion (shown here) and electroweak production off-shell ( $m_{4f} \gtrsim 2m_V \sim 200 \text{ GeV}$ ) Used in analysis: 1.2
  - Sensitive to  $\Gamma_H$  and anomalous couplings



- VH production at NLO in both gg and qq initial states
  - See paper for methods to detect and analyze  $gg \rightarrow ZH$

#### Optimal observables



p<sub>int</sub>

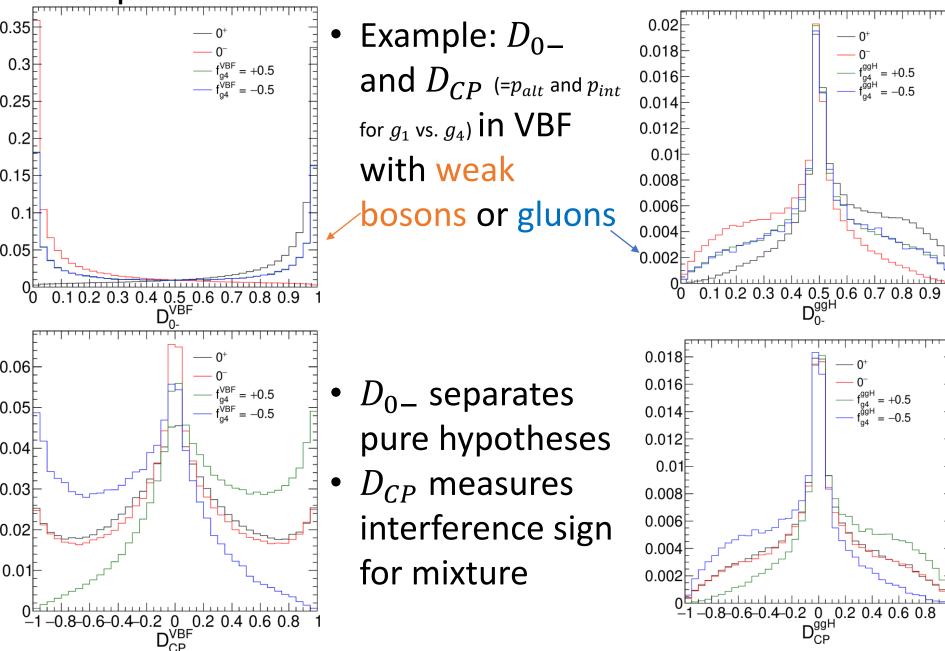
- Parameterize events in terms of kinematics  $\hat{\Omega}$
- For  $VV \rightarrow H$  with 2 jets or  $H \rightarrow VV \rightarrow 4f$ , 5 angles and 2 V invariant masses, plus H invariant mass
- *ttH*: even more angles and masses
  - also complicated because we can't tell which jet is which, or in the leptonic case we have missing neutrinos
  - <u>arXiv: 1606.03107 [hep-ex]</u>

 $D_{alt}(\Omega)$ 

• Can't deal with so many dimensions: construct observables that use all this information

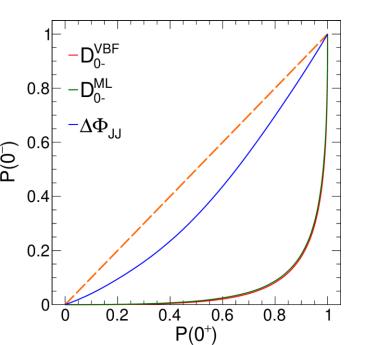
 $\frac{p_{sig}(\Omega)}{p_{sig}(\vec{\Omega}) + p_{alt}(\vec{\Omega})} D_{int}(\vec{\Omega})$ 

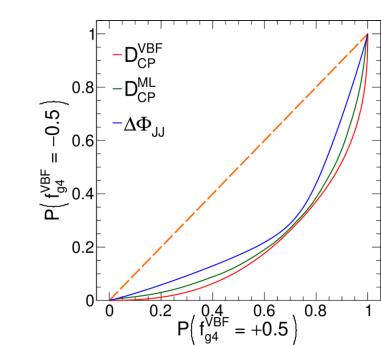
#### **Optimal observables**

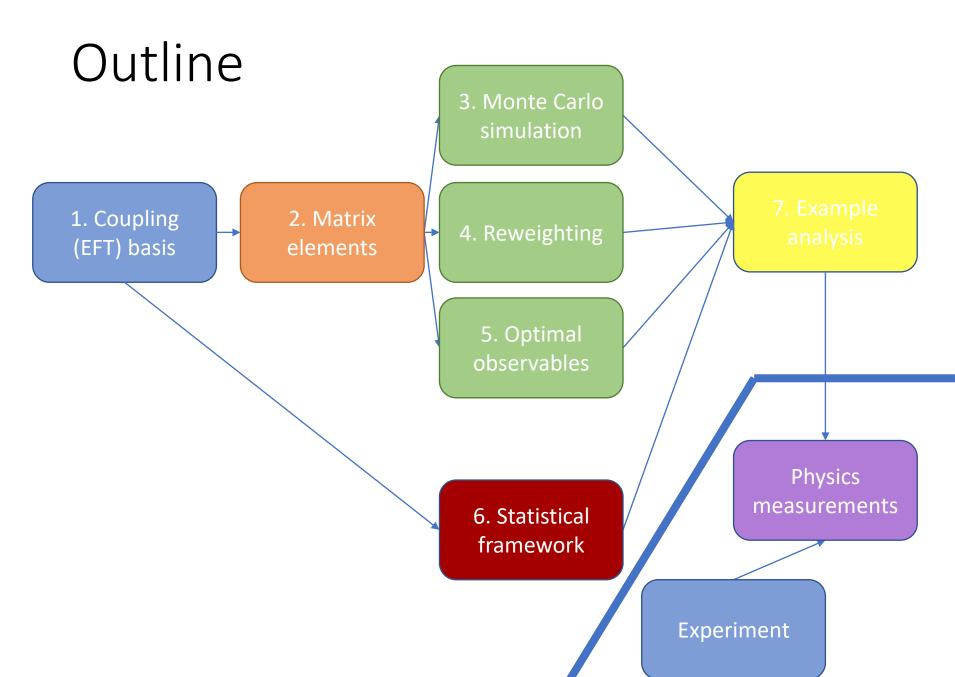


#### Machine learning

- Before NLO and detector effects, MELA contains full information: no other observable can do better
- In the limit of perfect training, machine learning (ML) can give equivalent results
  - Difficult in practice in the case of  $D_{CP}$
- If detector effects are large or some particles are lost, can e.g. integrate or use transfer functions in MELA, or use ML







#### Statistical framework

$$\mathcal{P}(\vec{\Omega}; \vec{a}) \sim \frac{\left(\sum_{i,j} a_i a_j \alpha_{i,j}^{(i)}(\vec{\Omega}^{(i)})\right) \left(\sum_{i,j} a_i a_j \alpha_{i,j}^{(f)}(\vec{\Omega}^{(f)})\right)}{(s - m_H^2)^2 + m_H^2 \Gamma^2}$$

- Polynomial in  $\vec{a}$
- When we look at only production or decay (e.g.  $gg \rightarrow H \rightarrow ZZ$ ), N = 2 powers of a; when they have the same couplings (e.g. VBF  $\rightarrow H \rightarrow ZZ$ ), N = 4
- For K couplings, have  $\binom{N+K-1}{N}$  terms to parameterize

• E.g. 
$$N = 4$$
,  $K = 5 \Rightarrow 70$  terms

• Need to either simulate each one or reweight

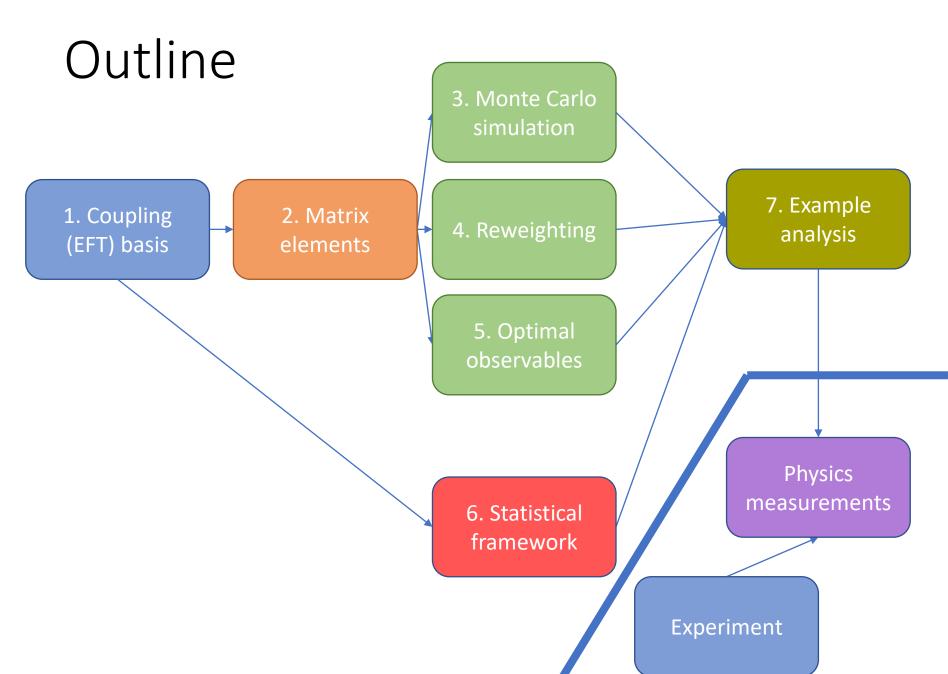
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#### Parameterization

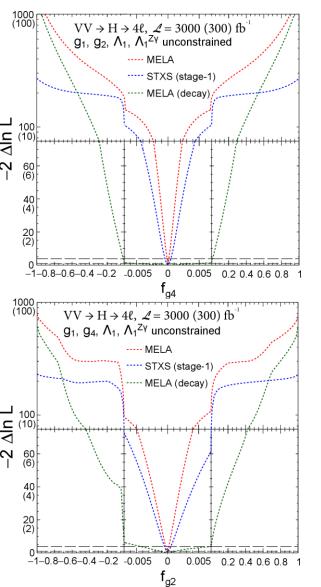
• Can either float couplings  $\vec{a}$  and width  $\Gamma_H$  directly, or reparameterize in terms of  $\vec{f}_a$  and  $\mu$ 

• 
$$f_{ai} = \frac{a_i^2 \sigma_i}{\sum_j a_j^2 \sigma_j} \operatorname{sgn} a_i$$

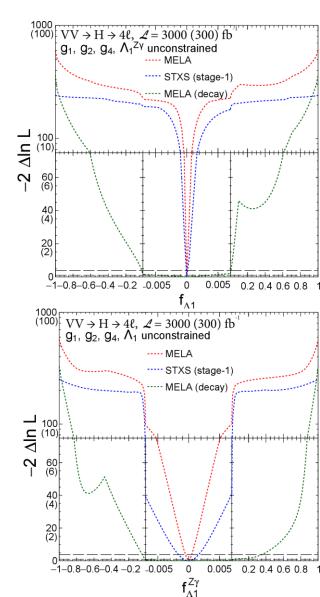
- Bounded between -1 and 1
- $\sigma_i$  depends on the process have to specify a convention
- By default for HZZ couplings we use  $H \rightarrow ZZ \rightarrow 2e2\mu$
- $\vec{f}_a$  is independent of overall scaling (including  $\Gamma_H$ ), which is absorbed by  $\mu$



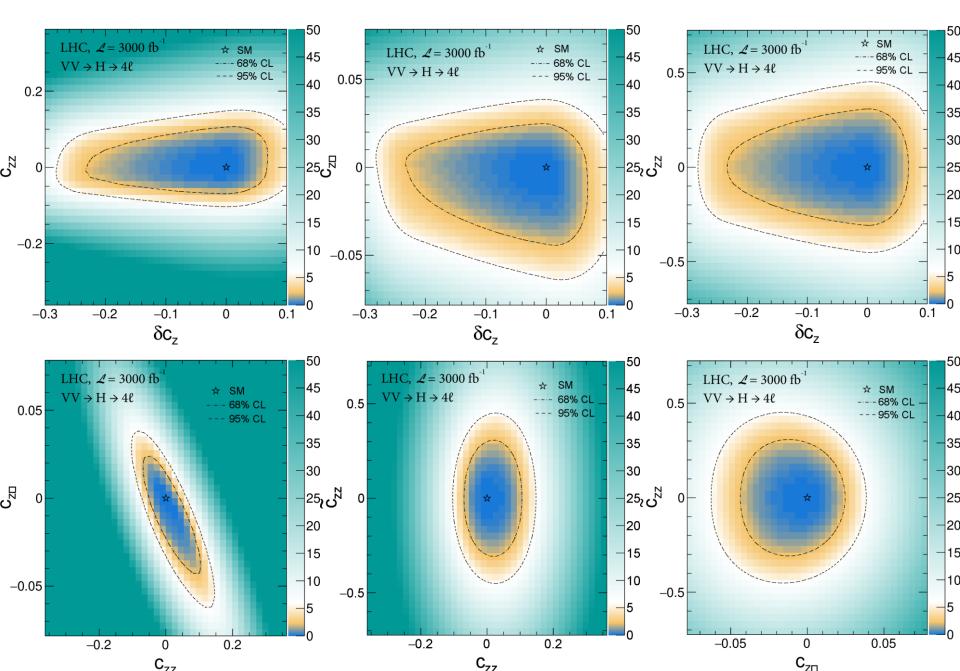
### Example analysis: search for $g_4$ , $g_2$ , $\Lambda_1$ , and $\Lambda_1^{Z\gamma}$



- Floating SM and 4 anomalous couplings simultaneously
- Expected constraints at 300 and 3000 fb<sup>-1</sup>
- Full MELA approach
- Using simplified template cross section (STXS) discriminants
- Use information from decay only
  - More model independent (no high q<sup>2</sup> regime)



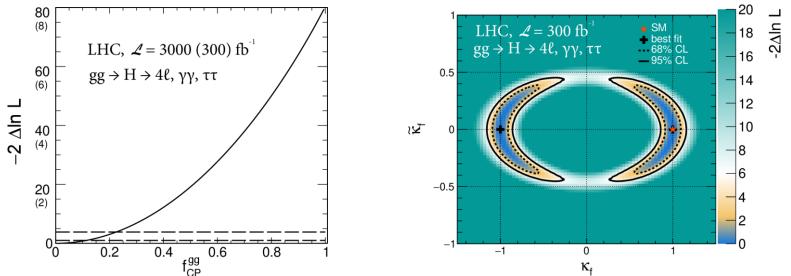
#### Example analysis: EFT: $\delta c_z$ , $c_{zz}$ , $c_{z\Box}$ , $\tilde{c}_{zz}$

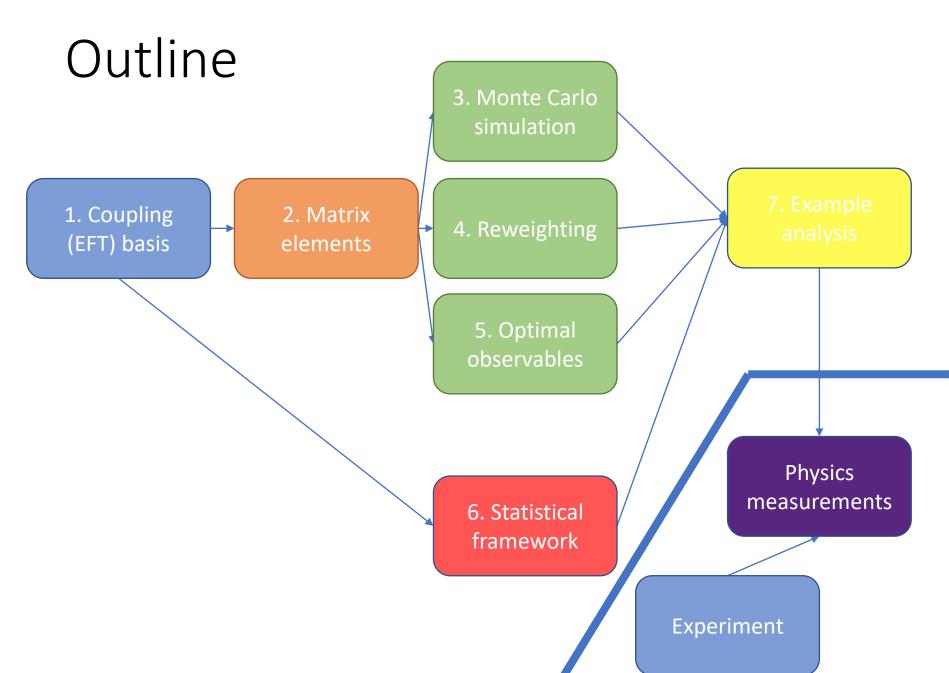


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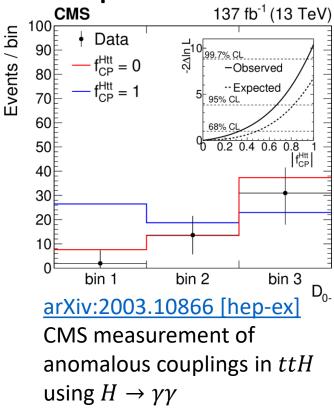
#### Example analysis: anomalous ggH

- Can measure CP in *ggH* using the MELA approach and discriminants
- Requires events with 2 jets (small fraction of *ggH*), and even then only VBF-like topology is sensitive
- Electroweak VBF forms a background
- Can start constraining now, tighter constraints at HL-LHC





#### Experimental results using JHUGen



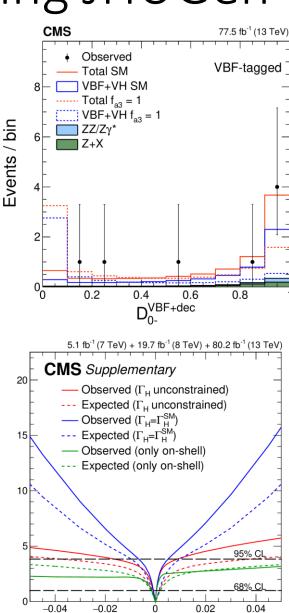
 $D_{0-}$  is calculated here using machine learning due to the complicated jet correlations and missing neutrinos. Samples are produced with JHUGen arXiv:2003.10866

#### [hep-ex]

CMS measurement of anomalous couplings in on-shell and offshell production and decay  $H \rightarrow 4\ell$ 

 $D_{0-}$  is calculated using VBF production and  $H \rightarrow 4\ell$  decay and provides strong separation between  $g_1$  and  $g_4$ 

Likelihood scans for  $\exists \forall f_{a3}(=f_{g4})$  using on-  $\heartsuit$  shell production and two possible off-shell scenarios



#### Summary

- Coherent framework for measuring Higgs boson anomalous couplings using a general model
- The tools, framework, and methods are general enough to build analyses for any convention you like – our anomalous couplings, EFT, pseudo-observables, ...
- The most general analysis will float more couplings at once, and then the results (not only the analysis methods) are independent of notation as well
  - Limited only by available data and CPU
- Sample analysis and approximate sensitivity for HL-LHC
- Has been used extensively for CMS and ATLAS analyses
- Please see <u>the paper</u> for more