



中国科学院高能物理研究所
Institute of High Energy Physics Chinese Academy of Sciences

Higgs CP measurement with EFT model in lepton collider

Qiyu Sha , Fangyi Guo, Gang Li, Yaquan Fang, Xinchou Lou

Institute of High Energy Physics CAS

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Summary

Introduction

Properties of Higgs in Standard Model: $m_H = 125.10\text{GeV}$, $J^{PC} = 0^{++}$

Related experiments in LHC:

- The hypothesis of spin-1 or spin-2 Higgs has been excluded by the ATLAS and CMS at >99% CL in $\sqrt{s} = 7\&8\text{ TeV}$, 25 fb^{-1} data. [Eur. Phys. J. C75 \(2015\) 476](#)
- The results of the study on the CP properties of the Higgs boson interactions with gauge bosons by the ATLAS and CMS show no deviations from the SM predictions.

Higgs-gauge vector boson interaction lacks precise measurement in all inclusive Higgs production mode(i.e. ggF dominant).

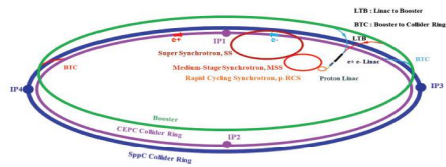
Any observation of CP violation in Higgs would be New Physics!

What we want to study is the Higgs CP mixing model aiming to find the CP-odd Higgs.

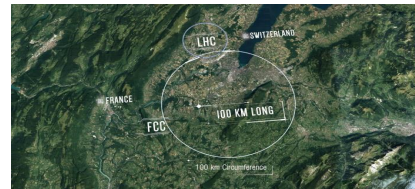
Introduction

Future e^+e^- collider experiment as Higgs factory :

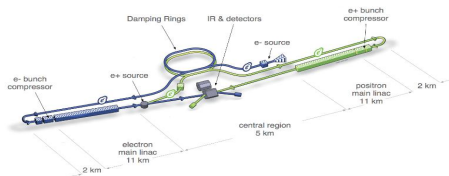
- At a center of mass energy of $\sqrt{s} \sim 240 \text{ GeV}$ which maximizes the Higgs boson production cross section through $e^+e^- \rightarrow ZH$ process.
- Cleaner environment and more events produced than (HL)-LHC.
- More precise Higgs-gauge boson coupling study.



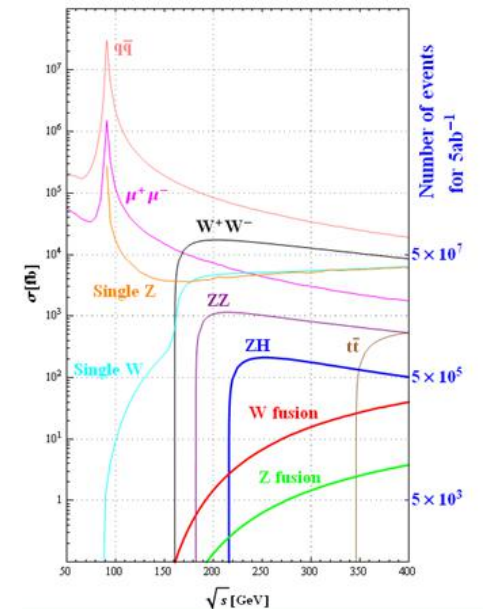
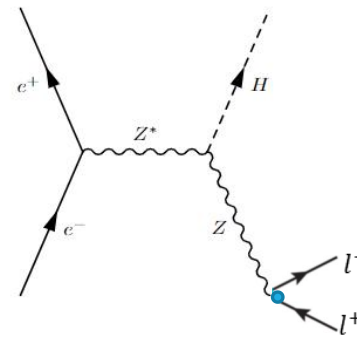
CEPC, 5.6 ab^{-1} @ 240 GeV



FCC-ee, 5 ab^{-1} @ 240 GeV



ILC, 2 ab^{-1} @ 250 GeV



Theory model

[JHEP 03\(2016\) 050](#)

[JHEP 11\(2014\) 028](#)

Consider a 6-dimension EFT model: $\mathcal{L}_{eff} = \mathcal{L}_{SM}^{(4)} + \frac{1}{\Lambda^2} \sum_{k=1}^{59} \alpha_k \mathcal{O}_k (\mathcal{L}_{BSM})$

- In this base, the experimental observables G_F, m_Z, α_{em} could be presented:

$$m_Z = m_{Z0}(1 + \delta_Z), \quad G_F = G_{F0}(1 + \delta_{G_F}), \quad \alpha_{em} = \alpha_{em0}(1 + \delta_A)$$

$$\text{where: } \delta_Z = \hat{\alpha}_{ZZ} + \frac{1}{4} \hat{\alpha}_{\Phi D}, \quad \delta_{G_F} = -\hat{\alpha}_{4l} + 2\hat{\alpha}_{\Phi l}^{(3)}, \quad \delta_A = 2\hat{\alpha}_{AA}.$$

Theory model

[JHEP 03\(2016\) 050](#)

[JHEP 11\(2014\) 028](#)

The $H \rightarrow Zll$ matrix element:

$$\mathcal{M}_{HZ\ell\ell}^\mu = \frac{1}{m_H} \bar{u}(p_3, s_3) \left[\gamma^\mu (H_{1,V} + H_{1,A}\gamma_5) + \frac{q^\mu \not{p}}{m_H^2} (H_{2,V} + H_{2,A}\gamma_5) + \frac{\epsilon^{\mu\nu\sigma\rho} p_\nu q_\sigma}{m_H^2} \gamma_\rho (H_{3,V} + H_{3,A}\gamma_5) \right] v(p_4, s_4)$$

- Where $\epsilon_{0123} = +1$ and $q = p_3 + p_4$.

And the parameters in the function are following:

$$H_{1,V} = -\frac{2m_H(\sqrt{2}G_F)^{1/2} r}{r-s} g_V \left(1 + \hat{\alpha}_1^{\text{eff}} - \frac{\kappa}{r} \hat{\alpha}_{ZZ} - \frac{\kappa}{2r} \frac{Q_\ell g_{em}(r-s)}{sg_V} \hat{\alpha}_{AZ} \right)$$

$$H_{1,A} = \frac{2m_H(\sqrt{2}G_F)^{1/2} r}{r-s} g_A \left(1 + \hat{\alpha}_2^{\text{eff}} - \frac{\kappa}{r} \hat{\alpha}_{ZZ} \right),$$

$$H_{2,V} = -\frac{2m_H(\sqrt{2}G_F)^{1/2}}{r-s} g_V \left[2\hat{\alpha}_{ZZ} + \frac{Q_\ell g_{em}(r-s)}{sg_V} \hat{\alpha}_{AZ} \right]$$

$$H_{2,A} = \frac{4m_H(\sqrt{2}G_F)^{1/2}}{r-s} g_A \hat{\alpha}_{ZZ}$$

$$H_{3,V} = -\frac{2m_H(\sqrt{2}G_F)^{1/2}}{r-s} g_V \left[2\hat{\alpha}_{ZZ} + \frac{Q_\ell g_{em}(r-s)}{sg_V} \hat{\alpha}_{AZ} \right]$$

$$H_{3,A} = \frac{4m_H(\sqrt{2}G_F)^{1/2}}{r-s} g_A \hat{\alpha}_{ZZ}$$

$$\hat{\alpha}_1^{\text{eff}} \equiv \hat{\alpha}_{ZZ}^{(1)} - \frac{m_H(\sqrt{2}G_F)^{1/2} (r-s)}{2\sqrt{r}} \frac{\hat{\alpha}_{\Phi l}^V}{g_V}$$

$$\hat{\alpha}_2^{\text{eff}} \equiv \hat{\alpha}_{ZZ}^{(1)} + \frac{m_H(\sqrt{2}G_F)^{1/2} (r-s)}{2\sqrt{r}} \frac{\hat{\alpha}_{\Phi l}^A}{g_A}$$

 : SM term
Others : EFT contribution

Theory model

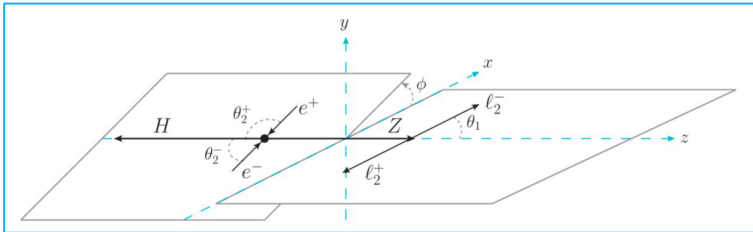
[JHEP 03\(2016\) 050](#)

[JHEP 11\(2014\) 028](#)

Differential cross section for $e^+e^- \rightarrow ZH \rightarrow llH$:

$$\frac{d\sigma}{d\cos\theta_1 d\cos\theta_2 d\phi} = \frac{\mathcal{N}_\sigma(q^2)}{m_H^2} \mathcal{J}(q^2, \theta_1, \theta_2, \phi),$$

$$\mathcal{N}_\sigma(q^2) = \frac{1}{2^{10}(2\pi)^3} \cdot \frac{1}{\sqrt{r}\gamma_Z} \cdot \frac{\sqrt{\lambda(1,s,r)}}{s^2}$$



$$\begin{aligned} \mathcal{J}(q^2, \theta_1, \theta_2, \phi) = & J_1(1 + \cos^2 \theta_1 \cos^2 \theta_2 + \cos^2 \theta_1 + \cos^2 \theta_2) \\ & + J_2 \sin^2 \theta_1 \sin^2 \theta_2 + J_3 \cos \theta_1 \cos \theta_2 \\ & + (J_4 \sin \theta_1 \sin \theta_2 + J_5 \sin 2\theta_1 \sin 2\theta_2) \sin \phi \\ & + (J_6 \sin \theta_1 \sin \theta_2 + J_7 \sin 2\theta_1 \sin 2\theta_2) \cos \phi \\ & + J_8 \sin^2 \theta_1 \sin^2 \theta_2 \sin 2\phi + J_9 \sin^2 \theta_1 \sin^2 \theta_2 \cos 2\phi. \end{aligned}$$

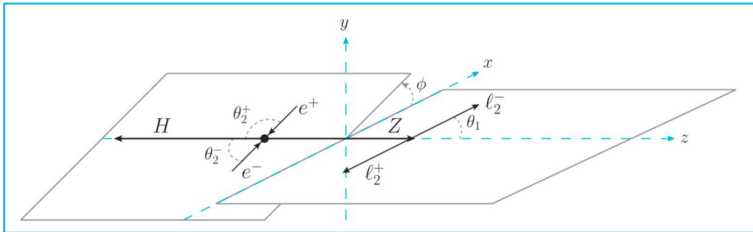
Variables for studying distribution: θ_1, θ_2, ϕ

Theory model

Differential cross section for $e^+e^- \rightarrow ZH \rightarrow llH$:

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Variables for studying distribution: θ_1, θ_2, ϕ

Assumption for simplification:

- $\delta_{GF} = \hat{\alpha}_{\phi l}^V = \hat{\alpha}_{\phi l}^A = \hat{\alpha}_{A\tilde{Z}} = \hat{\alpha}_{Z\tilde{Z}} = 10^{-3}$, others are set to 0, so $H_{2,V/A} = 0$.

$$J_1 = 2rs(g_A^2 + g_V^2)(|H_{1,V}|^2 + |H_{1,A}|^2),$$

$$J_2 = \kappa(g_A^2 + g_V^2)[\kappa(|H_{1,V}|^2 + |H_{1,A}|^2) + \lambda \text{Re}(H_{1,V}H_{2,V}^* + H_{1,A}H_{2,A}^*)],$$

$$J_3 = 32rs g_A g_V \text{Re}(H_{1,V}H_{1,A}^*),$$

$$J_4 = 4\kappa\sqrt{rs}\lambda g_A g_V \text{Re}(H_{1,V}H_{3,A}^* + H_{1,A}H_{3,V}^*),$$

$$J_5 = \frac{1}{2}\kappa\sqrt{rs}\lambda(g_A^2 + g_V^2) \text{Re}(H_{1,V}H_{3,V}^* + H_{1,A}H_{3,A}^*),$$

$$J_6 = 4\sqrt{rs}g_A g_V [4\kappa \text{Re}(H_{1,V}H_{1,A}^*) + \lambda \text{Re}(H_{1,V}H_{2,A}^* + H_{1,A}H_{2,V}^*)],$$

$$J_7 = \frac{1}{2}\sqrt{rs}(g_A^2 + g_V^2)[2\kappa(|H_{1,V}|^2 + |H_{1,A}|^2) + \lambda \text{Re}(H_{1,V}H_{2,V}^* + H_{1,A}H_{2,A}^*)],$$

$$J_8 = 2rs\sqrt{\lambda}(g_A^2 + g_V^2) \text{Re}(H_{1,V}H_{3,V}^* + H_{1,A}H_{3,A}^*),$$

$$J_9 = 2rs(g_A^2 + g_V^2)(|H_{1,V}|^2 + |H_{1,A}|^2).$$

6 of these 9 functions are independent

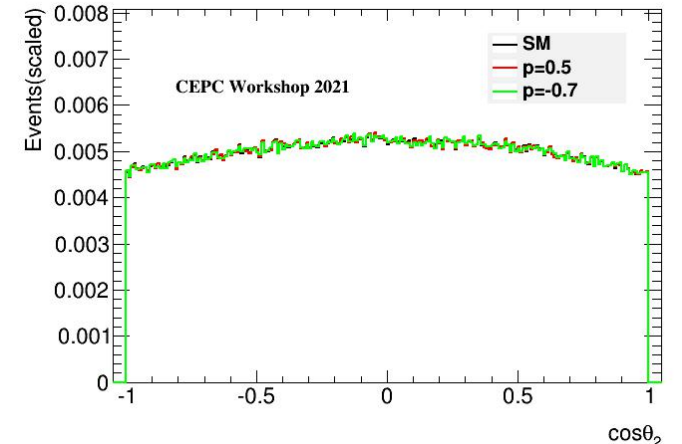
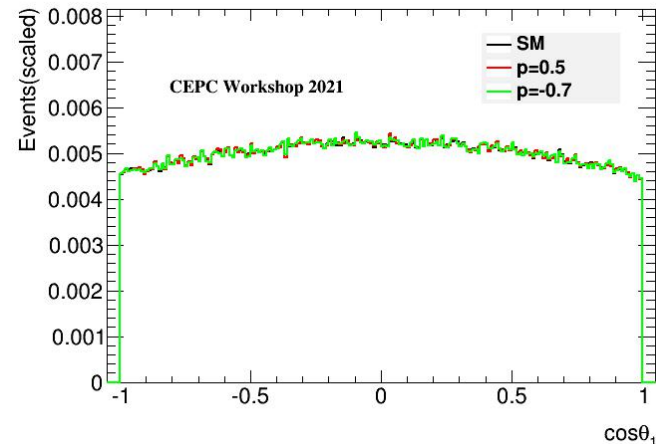
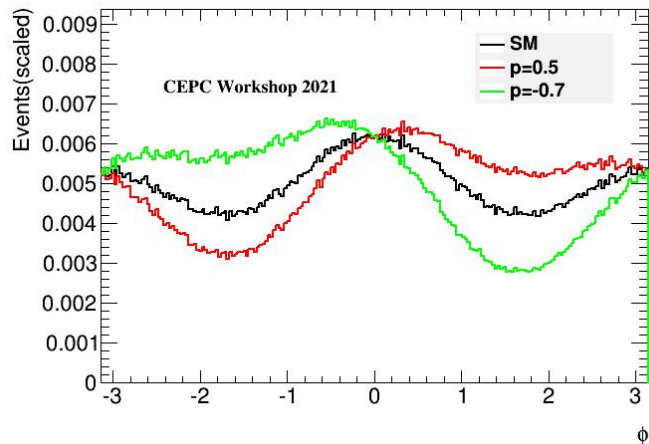
—	0 in assumption
	EFT CP-odd term
Others	CP-even contribution

Optimal variable approach

Differential cross section could be represent as:

$$\frac{d\sigma}{d\cos\theta_1 d\cos\theta_2 d\phi} = N \times (J_{CP-even}(\theta_1, \theta_2, \phi) + p \times J_{CP-odd}(\theta_1, \theta_2, \phi)).$$

where p is an additional global CP-mixing parameter.



Optimal variable approach

$$\frac{d\sigma}{d\cos\theta_1 d\cos\theta_2 d\phi} = N \times (J_{CP\text{-even}}(\theta_1, \theta_2, \phi) + p \times J_{CP\text{-odd}}(\theta_1, \theta_2, \phi)).$$

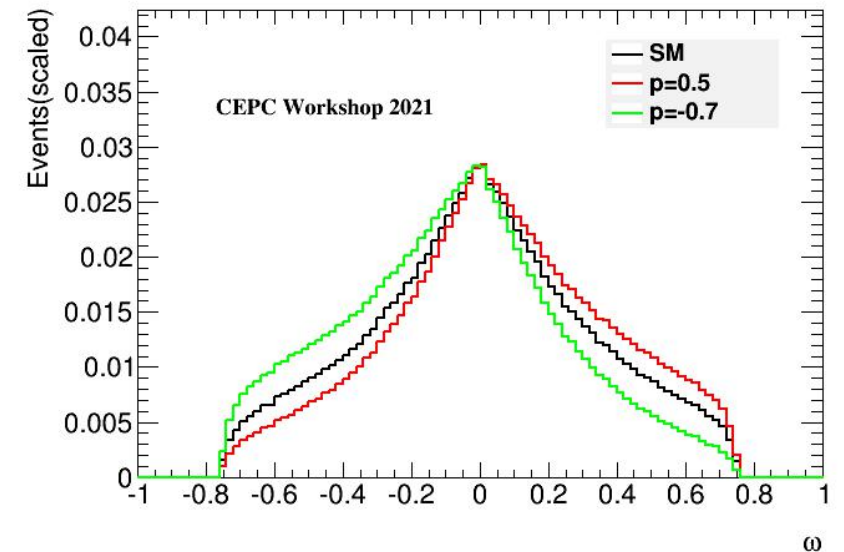
In this formation, we could define an **Optimal Variable** ω which combines the information from $\{\theta_1, \theta_2, \phi\}$:

$$\omega = \frac{J_{CP\text{-odd}}(\theta_1, \theta_2, \phi)}{J_{CP\text{-even}}(\theta_1, \theta_2, \phi)} \text{ to measure } p$$

Benefits:

- Combine the information from 3-dimension phase space
- Easier to study

[PLB 306 \(1993\) 411-417](#)



Monte Carlo samples

Two samples:

- MC Simulation: fast simulation based on CEPC-v4 detector design
- Truth level: simulate the physical process of $ee \rightarrow ZH$ directly in dynamics

MC simulation: Whizard 1.95 in CEPC, $e^+e^- \rightarrow ZH \rightarrow \mu^+\mu^-H$

- $\sqrt{s} = 250\text{GeV}$
- The mass of Higgs boson is assumed to be 125GeV and the couplings are set to the SM predictions.
- All the MC datasets are normalized to the expected yields in data with an integrated luminosity of 5.6ab^{-1} .

Event selection

(using Monte Carlo simulation sample)

- **Signal:** $e^+e^- \rightarrow ZH \rightarrow \mu^+\mu^-H(\rightarrow b\bar{b}/c\bar{c}/gg)$ channel
- **Background:** Irreducible background which contains the same final states as that in signal.
- Muon pair selection:

$$|\cos\theta_{\mu^+\mu^-}| < 0.81; \quad \text{Mass}_{\mu\mu} \in (77.5\text{GeV}, 104.5\text{GeV}); \quad M_{recoil_{\mu\mu}} \in (124\text{GeV}, 140\text{GeV}).$$

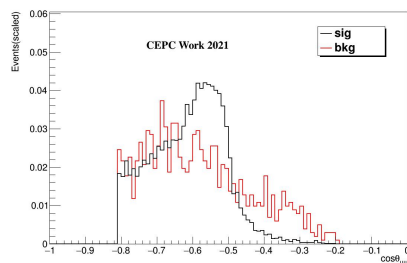
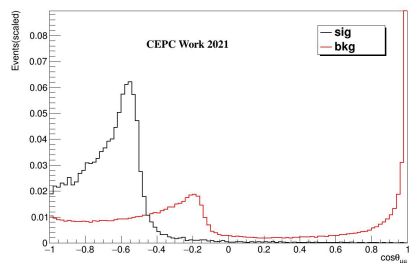
$$\text{Where } M_{recoil_{\mu\mu}}^2 = (\sqrt{s} - E_{\mu\mu})^2 - p_{\mu\mu}^2 = s - 2E_{\mu\mu}\sqrt{s} + m_{\mu\mu}^2$$

- Jets pair selection:

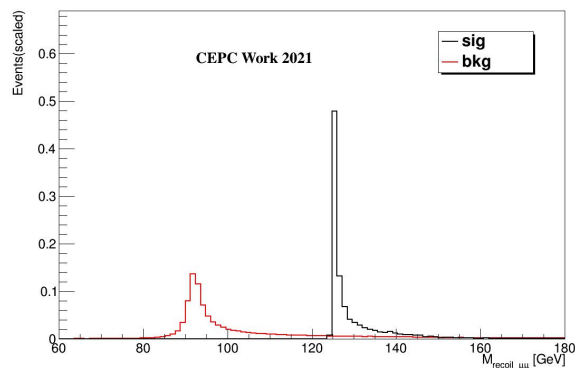
$$|\cos\theta_{jet}| < 0.96; \quad \text{Mass}_{jj} \in (100\text{GeV}, 150\text{GeV}).$$

Event selection

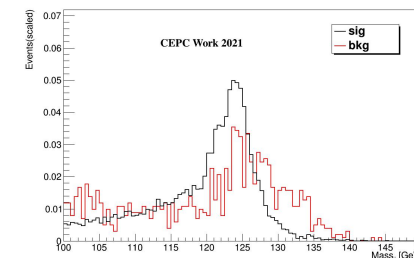
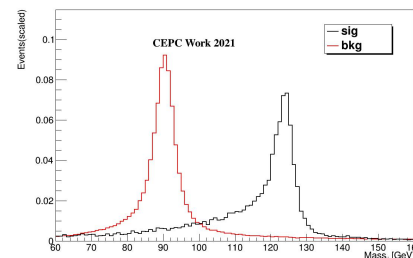
$$|\cos\theta_{\mu\mu}| < 0.81$$



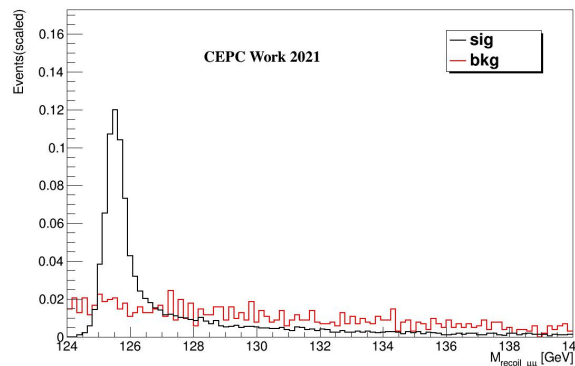
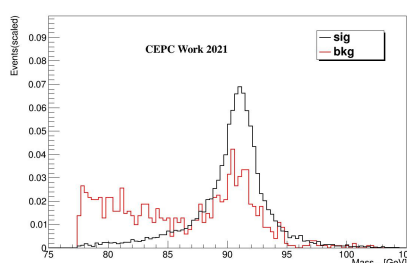
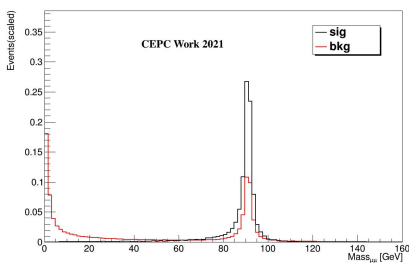
$$M_{recoil_mu\mu} \in (124\text{GeV}, 140\text{GeV})$$



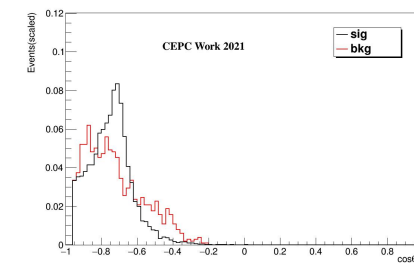
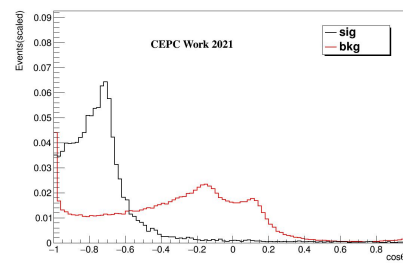
$$\text{Mass}_{jj} \in (100\text{GeV}, 150\text{GeV})$$



$$\text{Mass}_{\mu\mu} \in (77.5\text{GeV}, 104.5\text{GeV})$$



$$|\cos\theta_{jet}| < 0.96$$



Event selection

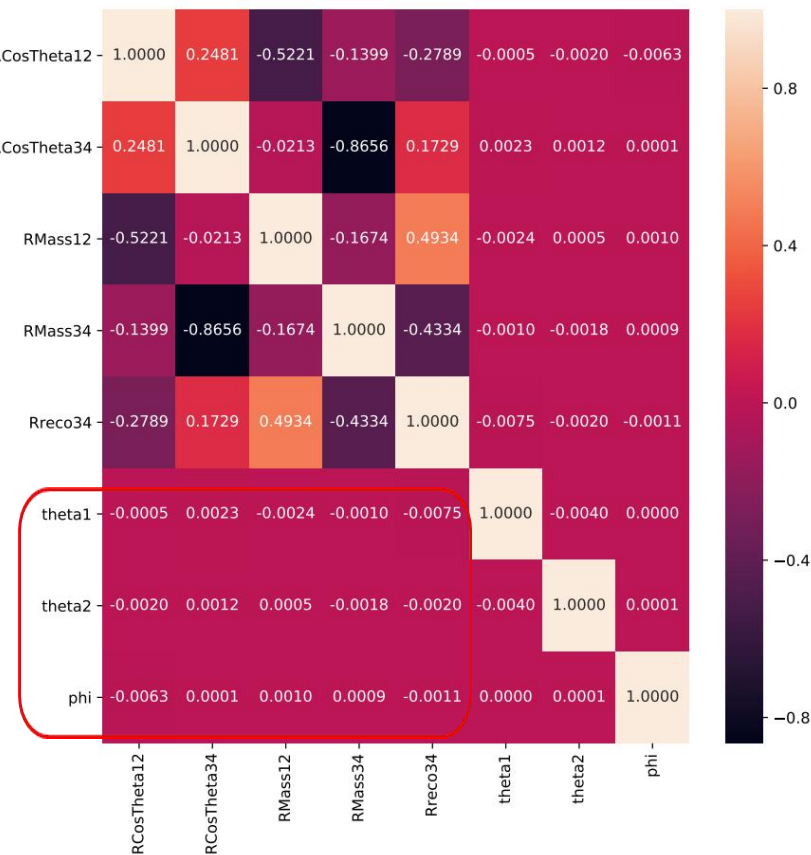
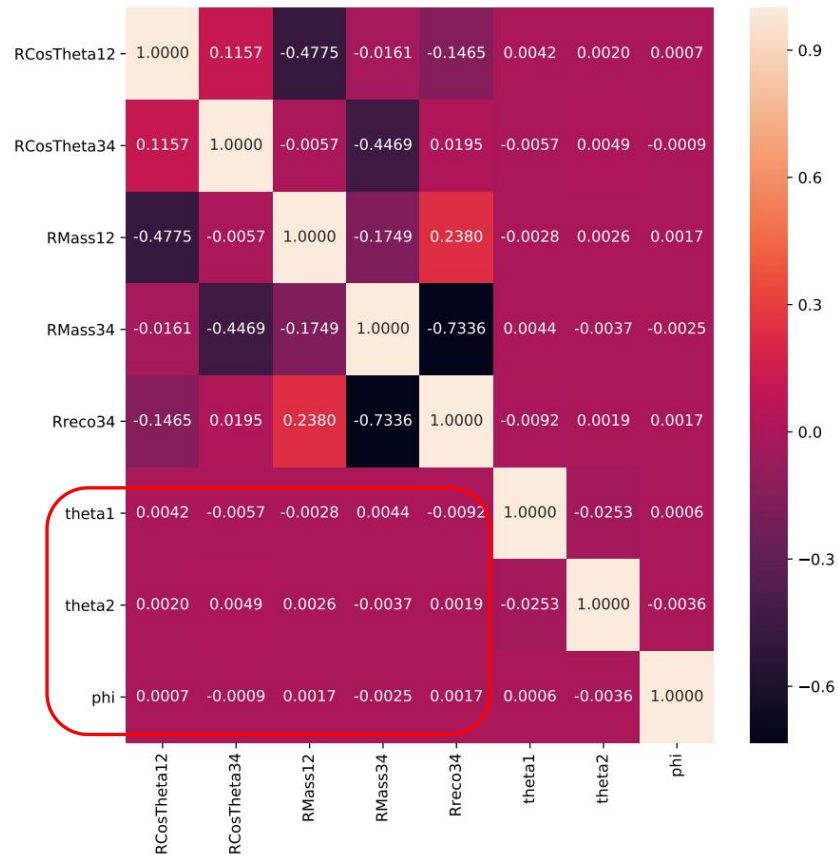
Cut Flow:

$ZH \rightarrow \mu^+ \mu^- + b\bar{b}/c\bar{c}/gg$ channel		
	Signal	Irreducible Background
Original	28627	1251768
Muon pair selection	18555 (efficiency:64.82%)	11198 (efficiency:0.89%)
All selection	13405 (efficiency:46.83%)	3610 (efficiency:0.29%)

Higgs CP-mixing measurement

Correlation:

- We can see that θ_1, θ_2, ϕ have little correlation with $\cos\theta_{\mu^+\mu^-}, \text{Mass}_{\mu\mu}, M_{recoil_{\mu\mu}}, \cos\theta_{jet}, \text{Mass}_{jj}$.

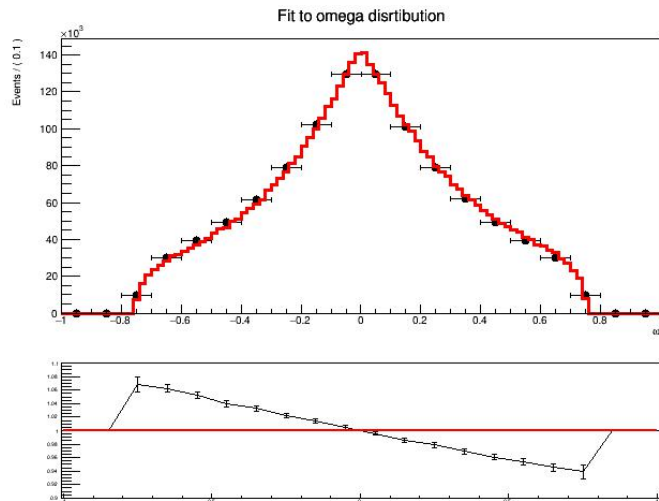


- So we can ignore the impact of event selections to $\theta_1, \theta_2, \text{ and } \phi$.

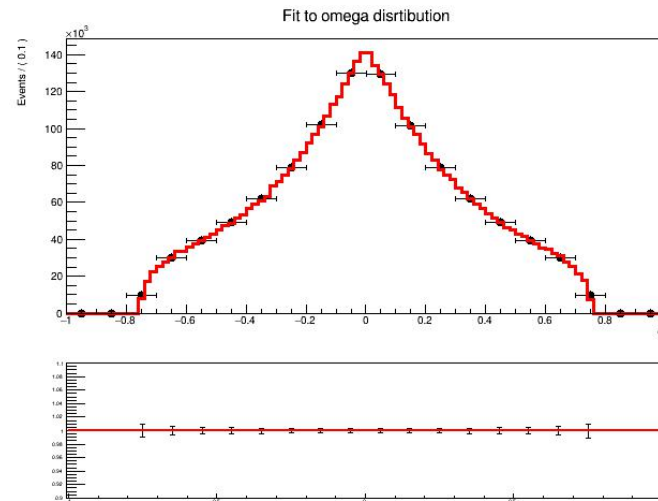
Fitting strategy and result

Fit strategy: Maximum-likelihood fit

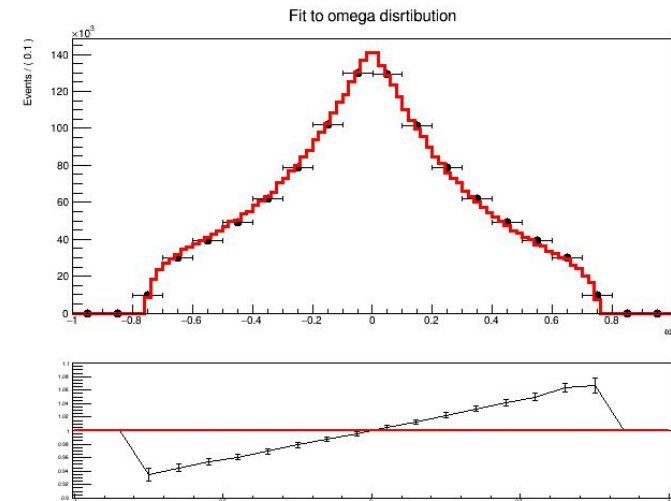
- $f^p(\omega) = N_{sig} * f_{sig}^p(\omega) + N_{bkg} * f_{bkg}^p(\omega)$
- **Fit ω to get $f_{sig}^p(\omega)$ and $f_{bkg}^p(\omega)$**
- **Fit $M_{recoil_{\mu\mu}}$ to get N_{sig} and N_{bkg}**
- Evaluate likelihood function for each p value hypothesis, and construct a ΔNLL as a function of p.



P= 0.09 compare p= 0



P= 0 compare p= 0

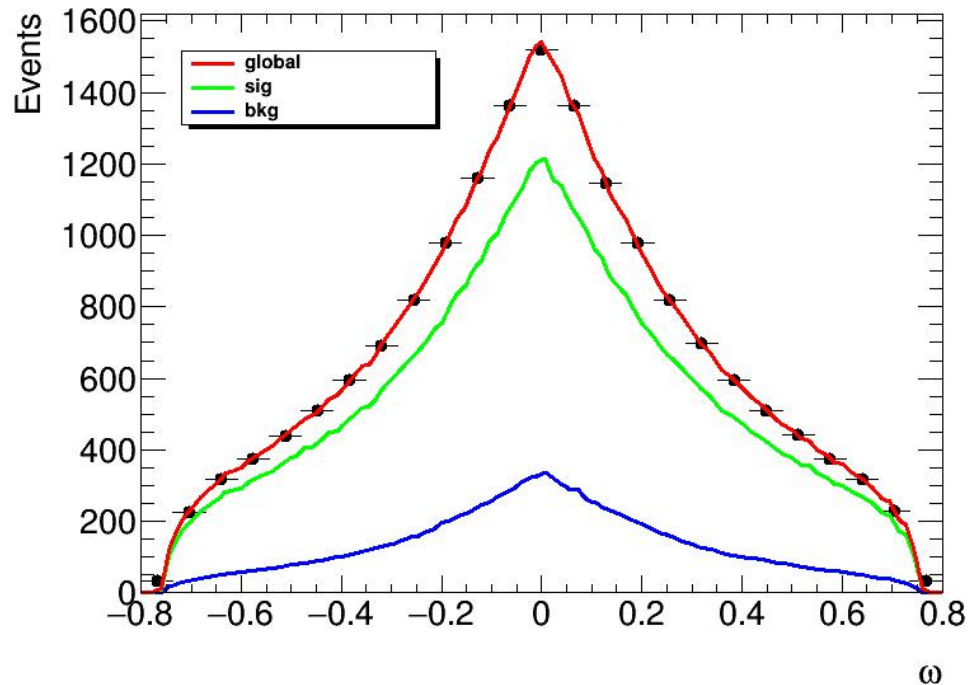


P= -0.09 compare p= 0

Fitting strategy and result

Fit ω :

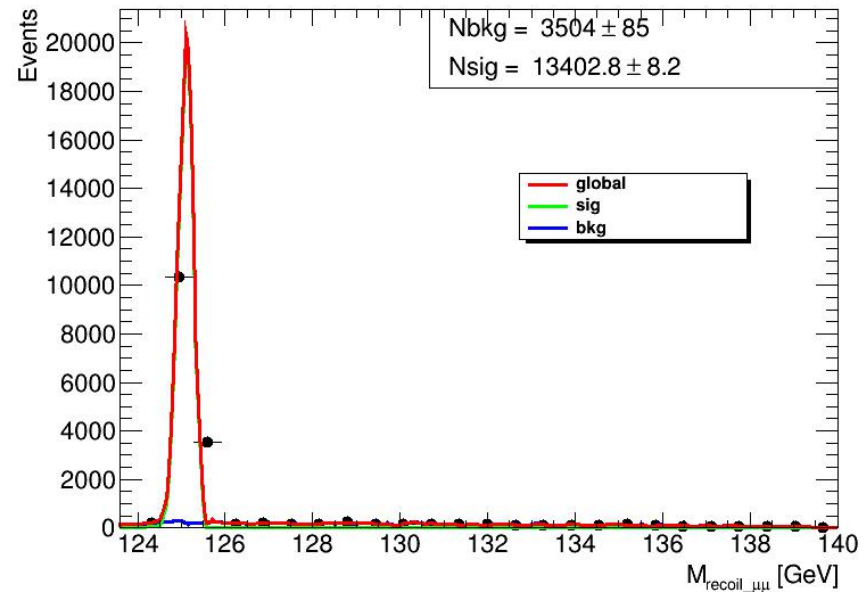
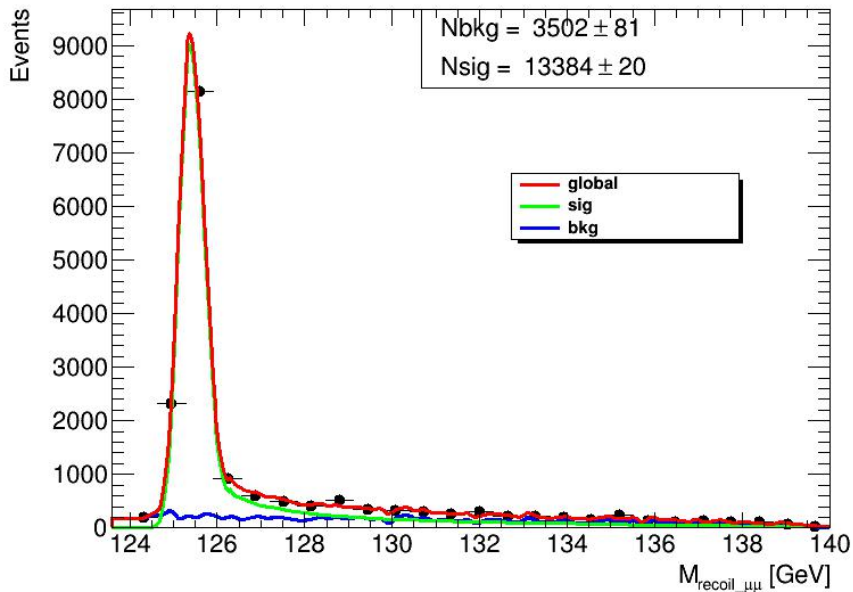
- Use histogram pdf to fit **MC signal and background sample**.
- The red curve is global fit, the green curve is signal events, the blue curve is background events.



Fitting strategy and result

Fit $M_{recoil_{\mu\mu}}$:

- Use histogram pdf to fit the $M_{recoil_{\mu\mu}}$ of signal (the signal is **on truth level** after **doing the smear to simulate detector error**).
- The $M_{recoil_{\mu\mu}}$ distribution of the combinatorial background (dominated by the $e^+e^- \rightarrow ZZ \rightarrow llq\bar{q}$) is modeled by a second order polynomial.
- Fit plots using truth level sample considering ISR(left) and not considering ISR(right).
- Using **ISR sample** can simulate the small exponential tail (which corresponding to the expected distribution.)



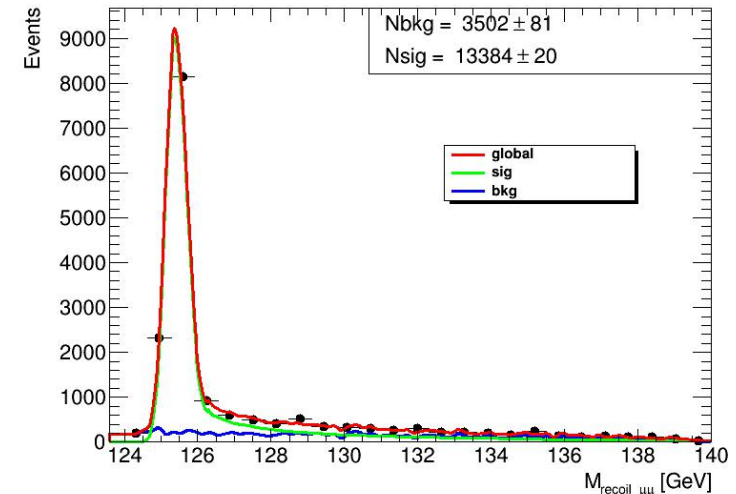
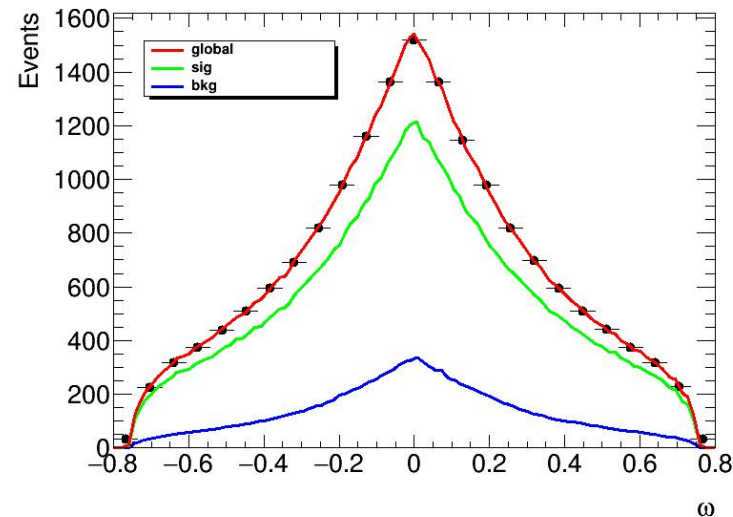
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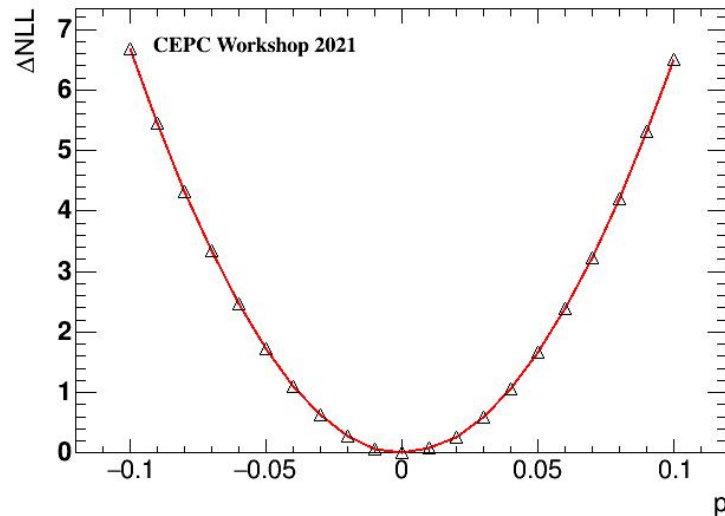
- Use histogram pdf to fit the $M_{recoil_{\mu\mu}}$ of signal (the signal is **on truth level** after **doing the smear to simulate detector error** and **considering the ISR** to simulate the small exponential tail).
- The $M_{recoil_{\mu\mu}}$ distribution of the combinatorial background (dominated by the $e^+e^- \rightarrow ZZ \rightarrow llq\bar{q}$) is modeled by a second order polynomial.



Fitting strategy and result

Extract maximum-likelihood fit p-value and interval

- Fit ΔNLL curve with a quadratic function $\Delta NLL(p) = a \cdot (p - p_0)^2$
- 68%(95%) CL interval corresponds to $\Delta NLL=0.5(1.96)$.



$$\Delta NLL(p|\omega) = 659.6(p - 5.6 \times 10^{-4})^2$$

For p :

$$68\% \text{ CL: } [-2.8 \times 10^{-2}, 2.7 \times 10^{-2}]$$

$$95\% \text{ CL: } [-5.5 \times 10^{-2}, 5.4 \times 10^{-2}]$$

Summary

An EFT based Higgs CP-mixing test is performed.

- Set up some basic assumptions to have a simplest CP-mixing model.
- Introduced optimal variable with better performance.
- Used ML-fit in ω distribution to extract p .
- Result: 95% CL $p \in [- 5.5 \times 10^{-2}, 5.4 \times 10^{-2}]$, corresponding to $\delta G_F, \hat{\alpha}_{\phi l}^V, \hat{\alpha}_{\phi l}^A, \hat{\alpha}_{AZ}, \hat{\alpha}_{Z\tilde{Z}} < 10^{-5}$.

Backup

Maximum likelihood fit

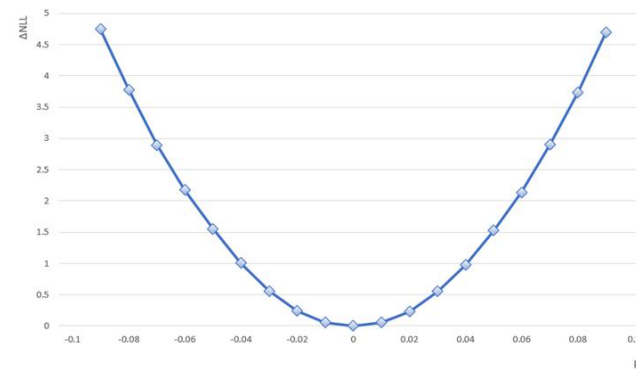
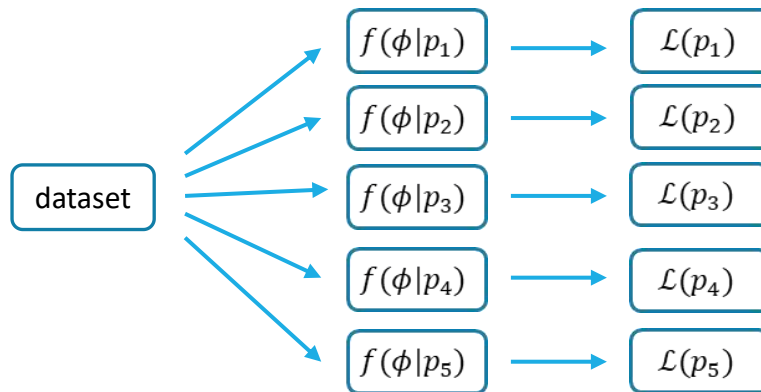
Construct a likelihood function

- $\mathcal{L}(\vec{x}|p, \vec{\theta}) = \prod_{data} f(x_i | p, \vec{\theta})$

$\vec{\theta}$: nuisance parameter. p : POI, CP-mixing parameter. x_i : dataset (ω).

- When statistics is large enough, we suppose $\mathcal{L}(\vec{x}|p, \vec{\theta}) \sim Gaus(\hat{p}, \sigma_p^2)$, so $\ln\mathcal{L}(p) = \ln\mathcal{L}_{max} - \frac{1}{2} \left(\frac{p-\hat{p}}{\sigma_p} \right)^2$

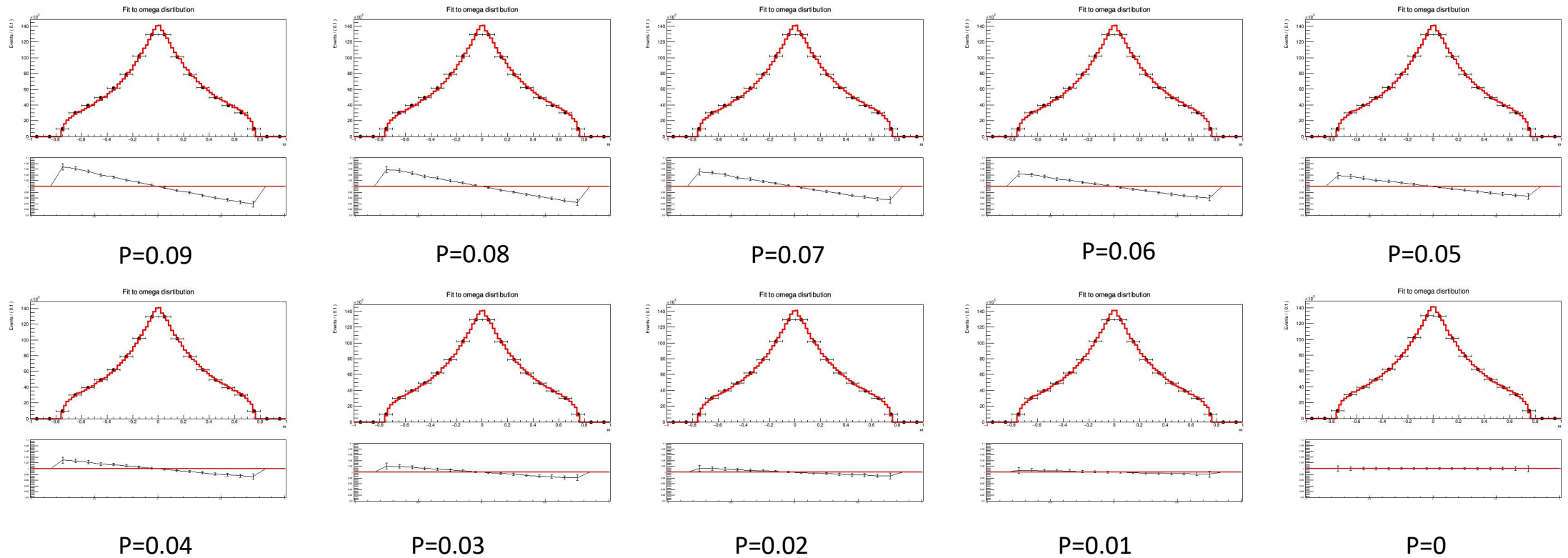
- From $\Delta NLL = NLL - NLL_{min}$ (negative log likelihood) we can extract maximum likelihood estimate \hat{p} and its CL interval.



Maximum likelihood fit

Sample modelling

- ω modelling: Histogram pdf. Highly depends on the sample statistics used to build histogram and HistPdf.



Maximum likelihood fit

Sample modelling

- ω modelling: Histogram pdf. Highly depends on the sample statistics used to build histogram and HistPdf.

