Higgs to dimuon discovery using quark / gluon tagging of ISR

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Collaboration with Won Sang Cho, Hyung Do Kim, Dongsub Lee







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- H $\rightarrow bb$: 5.6 σ
- ttH : 5.8σ

- PRD 99(2019) 072001
- PRL 121(2018) 1218081
- PLB 784(2018) 173





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Higgs ~ 2nd generation fermions @LHC? : $h \rightarrow \mu \mu$!





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Cross-section : ggH >> VBF Significance : $VBF \gtrsim ggH$



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dominant background



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large $\Delta \eta_{jj}$, m_{jj}



enhanced significance



enhanced significance



Intro / Motivation \rightarrow Saving ggH !



want to save ggH and exploit it well !

ex) ggH leading jets





ex) DY leading jets





H

ex) ggH leading jets





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Quark Jet Gluon Jet



ex) ggH leading jets





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Backup; kinematic consideration



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- $h \rightarrow \mu\mu$ significance can be enhanced
- by exploiting ggH channel more efficiently,
 - by quark / gluon tagging of ISR jet,
 - by including jet-substructure variables



Thanks

Backup; Roguh Estimation of Quark/Gluon ratio



DY



 $\propto 2C_F \mathcal{L}_{qq}$



$\mathcal{L}_{F}\mathcal{L}_{gq}$	$rac{\sigma_g}{\sigma_q} \propto rac{(2C_A + 3C_F)\mathcal{L}_{gg}}{C_F\mathcal{L}_{gq}} \simeq 7rac{\mathcal{L}_{gg}}{\mathcal{L}_{gq}} (\sim 3-4)$
$-\mathcal{L}_{gq}$	$\frac{\sigma_g}{\sigma_q} \propto \frac{2C_F \mathcal{L}_{gg}}{C_F \mathcal{L}_{gq}} \sim \frac{1}{7} - \frac{1}{1} + \text{next order gluon jets through glue} \\ \rightarrow \text{ quark triangle loop diagram}$



Backup; CMS variables

4 Dimuon system

 $p_T^{\mu\mu}$, $y^{\mu\mu}$

 ϕ_{CS} , cos θ_{CS} : azimuthal and polar angle in Collins-Soper rest frame.

• 4 Each muon $\eta^{\mu}, \frac{p_T^{\mu}}{m^{\mu\mu}}$

• 3 (+2) Each jet p_T^j , η^j $\Delta R(\mu\mu, j_1)$: distance between the leading jet and dimuon in $\eta - \phi$ space

- 3 Dijet system m^{jj} : dijet invariant mass $\Delta \eta^{jj}$ $\Delta \phi^{jj}$
- 5 Dimuon ~ jets $\Delta \eta(\mu\mu, j_1), \Delta \eta(\mu\mu, j_2)$ $\Delta \phi(\mu\mu, j_1), \Delta \phi(\mu\mu, j_2)$ $z_* = \frac{y^{\mu\mu} - (y^{j_1} + y^{j_2})/2}{|y^{j_1} - y^{j_2}|}$: Zepenfeld variable



Backup; jet-substructure variables



Backup; Multi-Variate Analysis (MVA)

Score Plots



Backup; Mutual information

