

Isolated lepton tagging & New (CS) jet clustering

Junping Tian (KEK)

The 43 General Meeting of ILC Physics Subgroup, Sep. 5, 2015 @ KEK

idea to select a isolated lepton (e/μ)

i) lepton ID: electron or muon

- ▶ energy ratio deposited in ECal, HCal, Yoke
- ▶ general PID: dE/dx + shower profile

ii) vertex: prompt or secondary

- ▶ d_0, z_0 significance: $d_0/\delta d_0, z_0/\delta z_0$

iii) isolation: not from jets

- ▶ relatively high P
- ▶ almost empty around

status of available tools

i) lepton ID: electron or muon

- ▶ energy ratio deposited in ECal, HCal, Yoke ✓
- ▶ general PID: dE/dx + shower profile to be added

ii) vertex: prompt or secondary

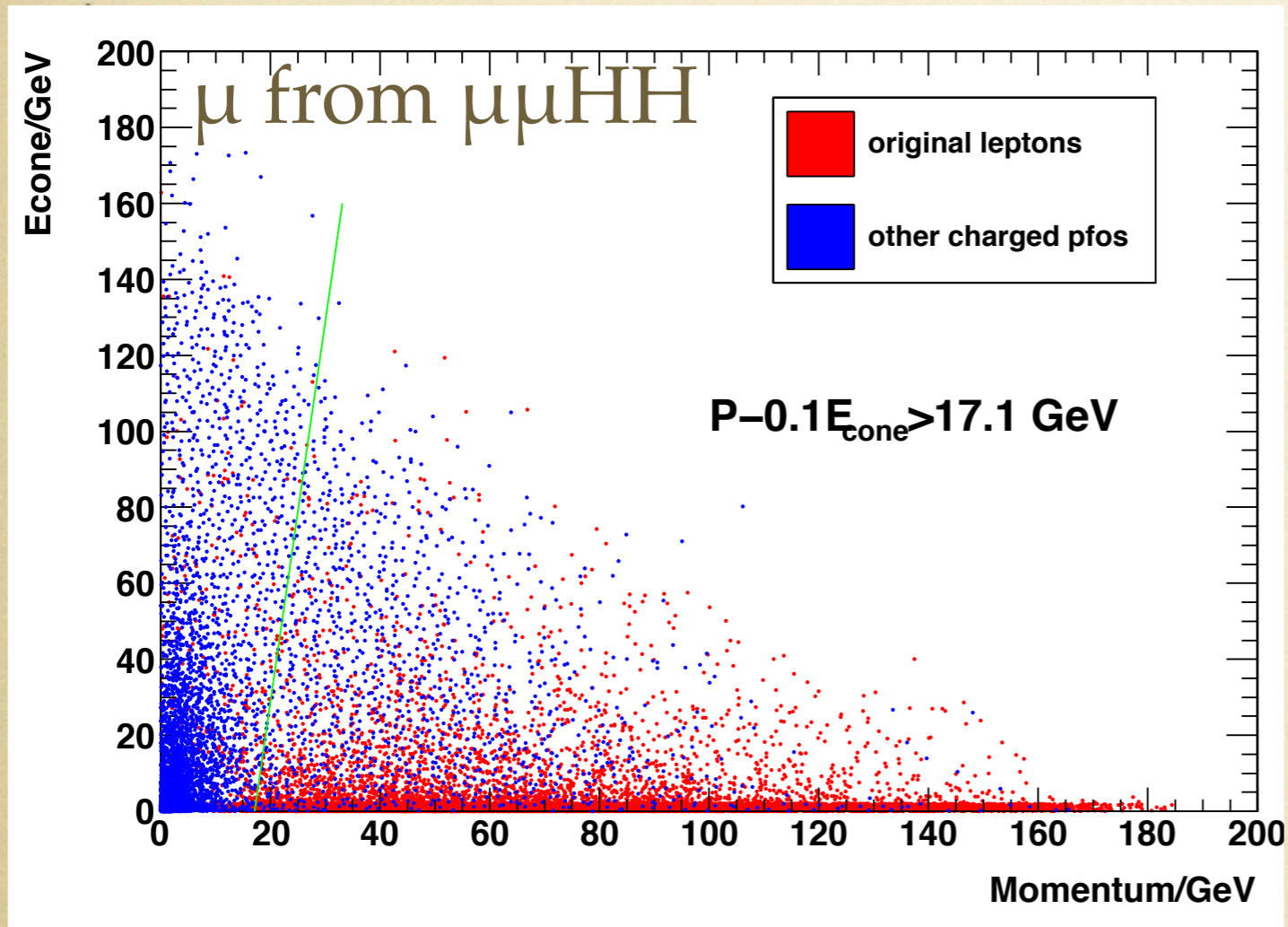
- ▶ d_0, z_0 significance: $d_0 / \delta d_0, z_0 / \delta z_0$ ✓

iii) isolation: not from jets ✓

- ▶ relatively high P several algorithms exist
- ▶ almost empty around

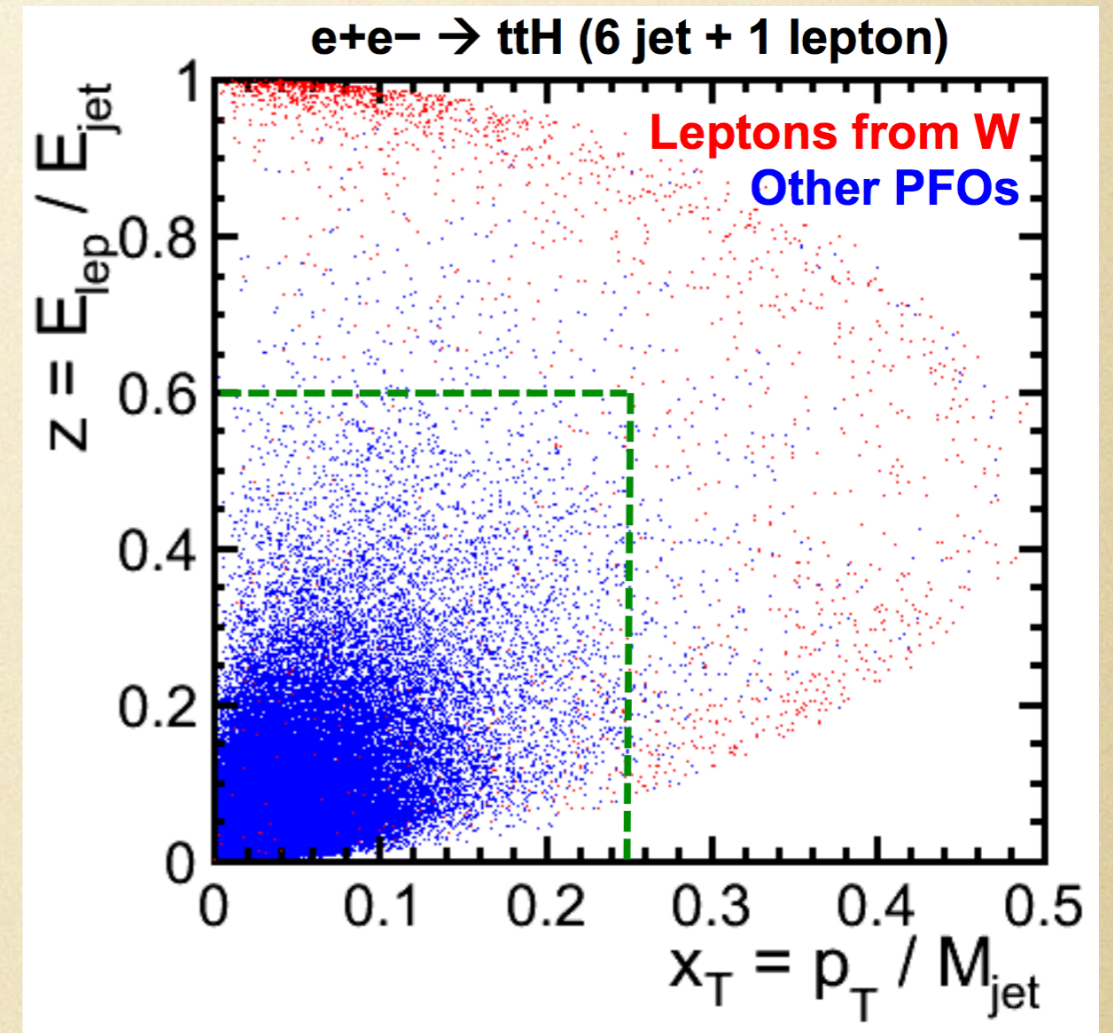
two isolation algorithms in DBD

cone based



(Ryo / Tomohiko; Junping)

jet based



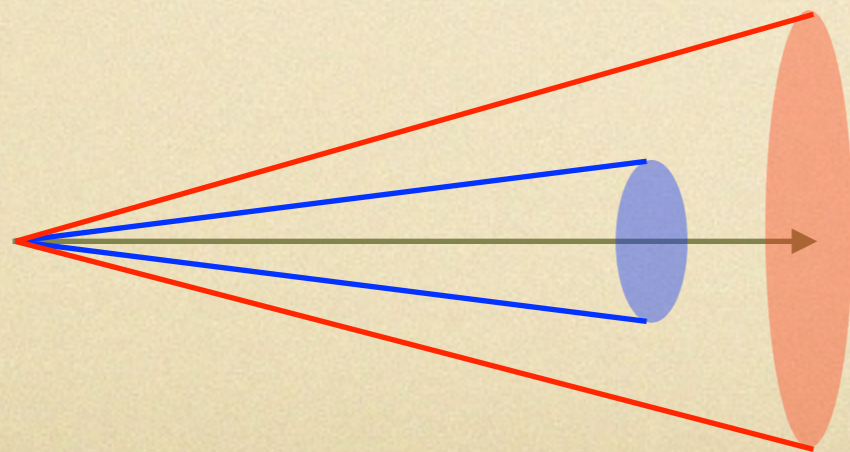
(M. Amjad / LAL group)

both available during DBD, in MarlinReco / Analysis / IsolatedLeptonFinder

New idea for further improvement:

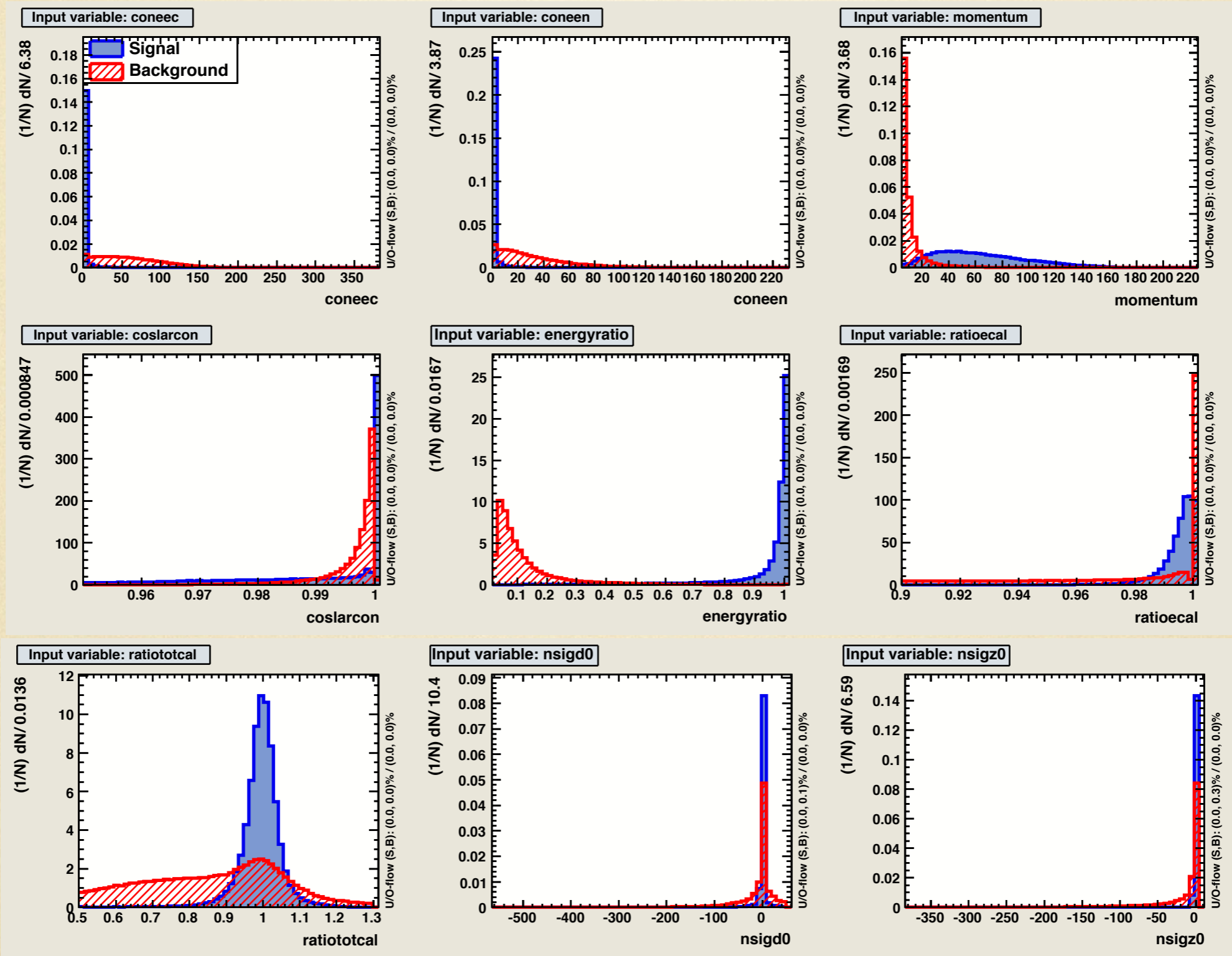
(developed by Claude/Junping for Higgs self-coupling analysis)

- utilise d_0 , z_0 , cone energy, momentum, E_{ecal} / E_{hcal} more effectively \rightarrow **MVA**
- rethink of “what is isolation?” in terms of separation with “non-isolated” ones from jets \rightarrow **utilise information of neighbour particles in every layer of cone**
- as a first trial, introduce double cones to magnify the jet influence \rightarrow $E(\text{lep}) / E(\text{jet})$ and angle between them



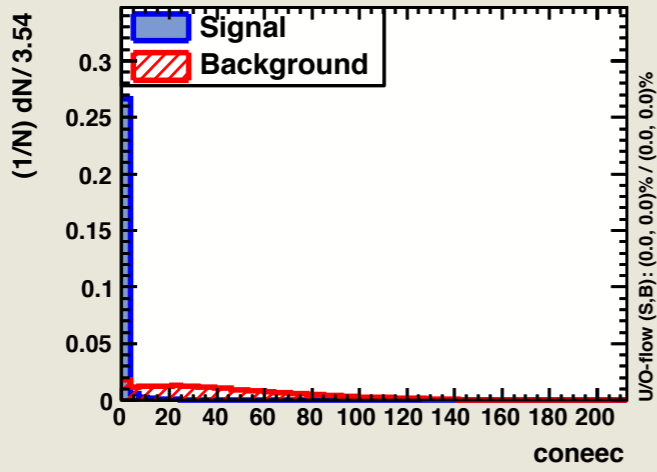
$$\cos=0.98, 95$$

input variables: electron

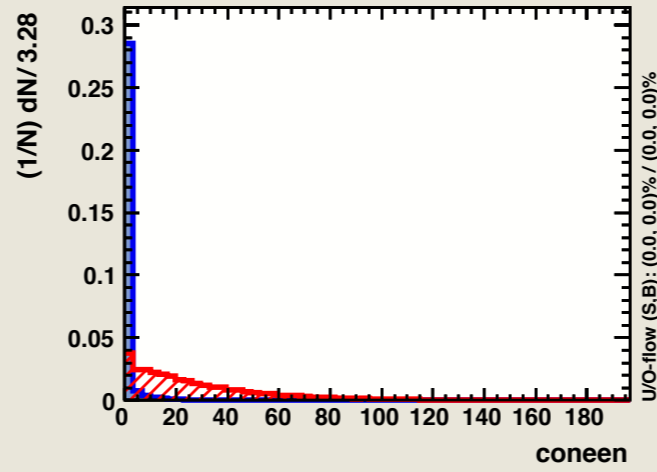


input variables: muon

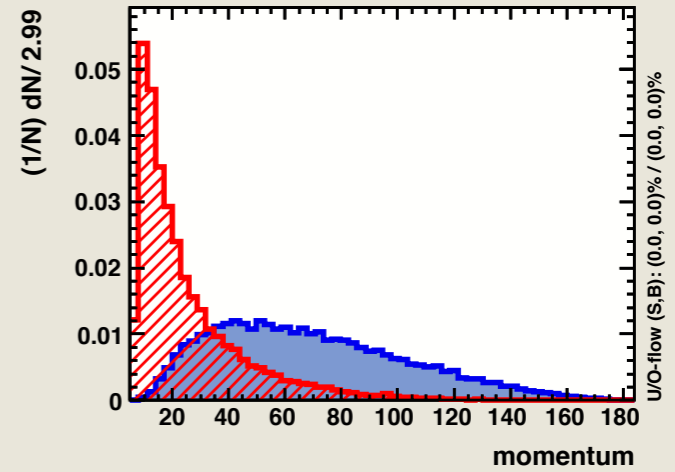
Input variable: coneec



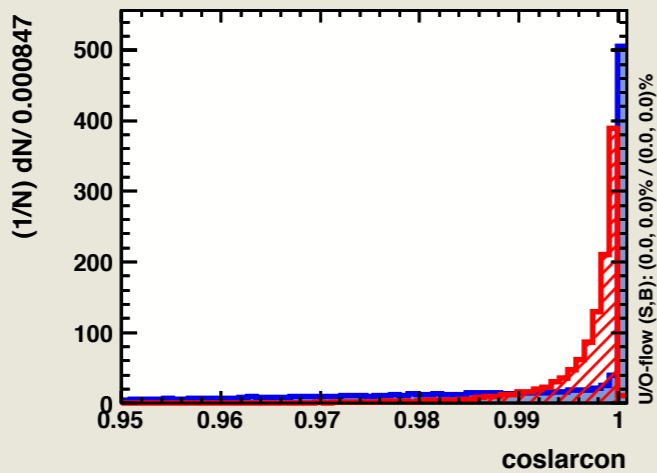
Input variable: coneen



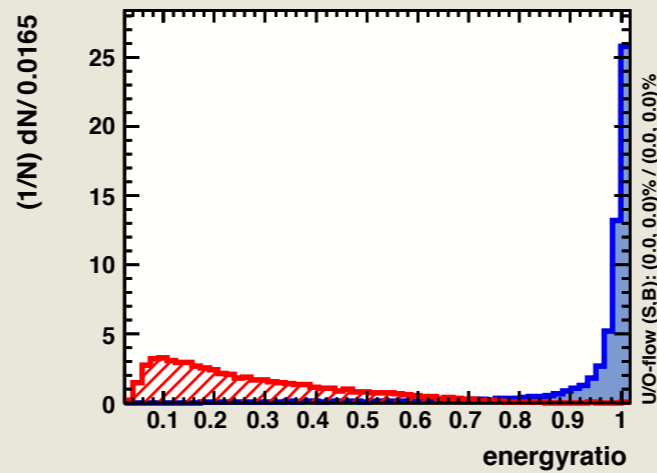
Input variable: momentum



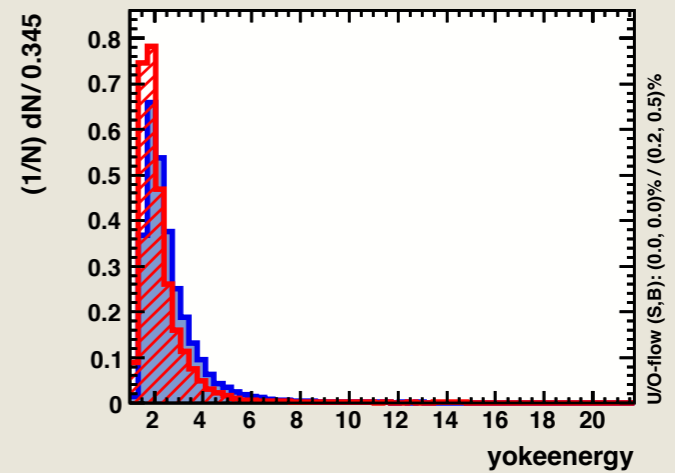
Input variable: coslarcon



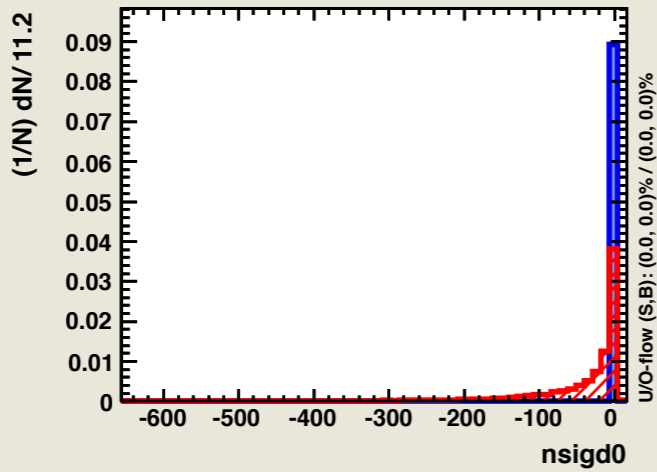
Input variable: energyratio



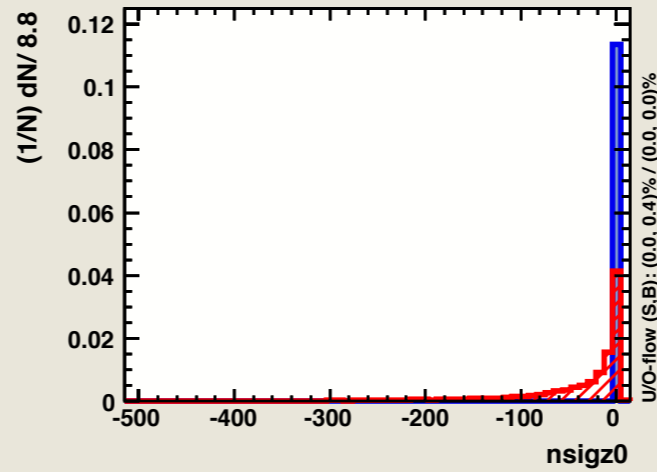
Input variable: yokeenergy



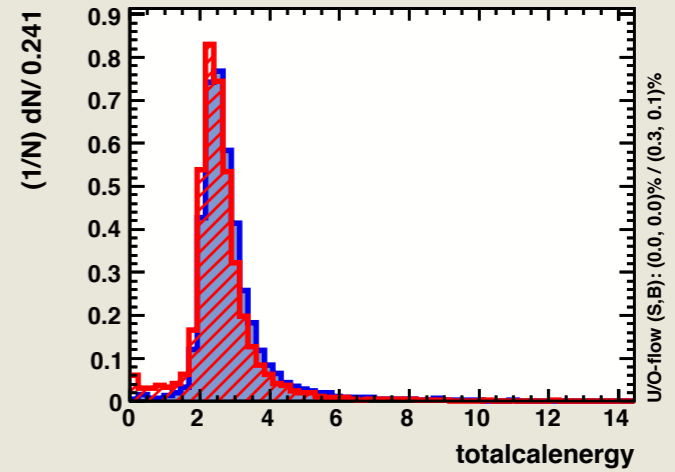
Input variable: nsigz0



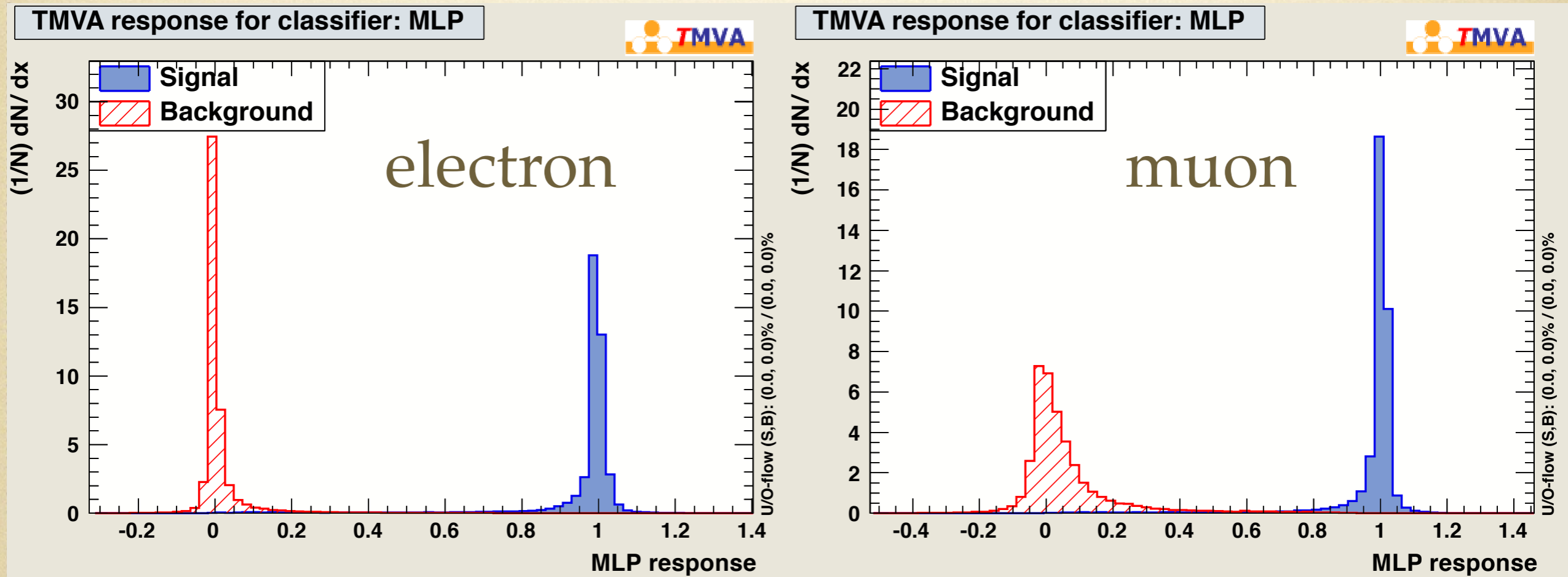
Input variable: nsigz0



Input variable: totalcalenergy



neural-net output (tagging)



lepton tagging is associated to the selected lepton collection, can be optimised in final selection

Available Processor in ilcsoft-v01-17-08

MarlinReco / Analysis / IsolatedLeptonTagging

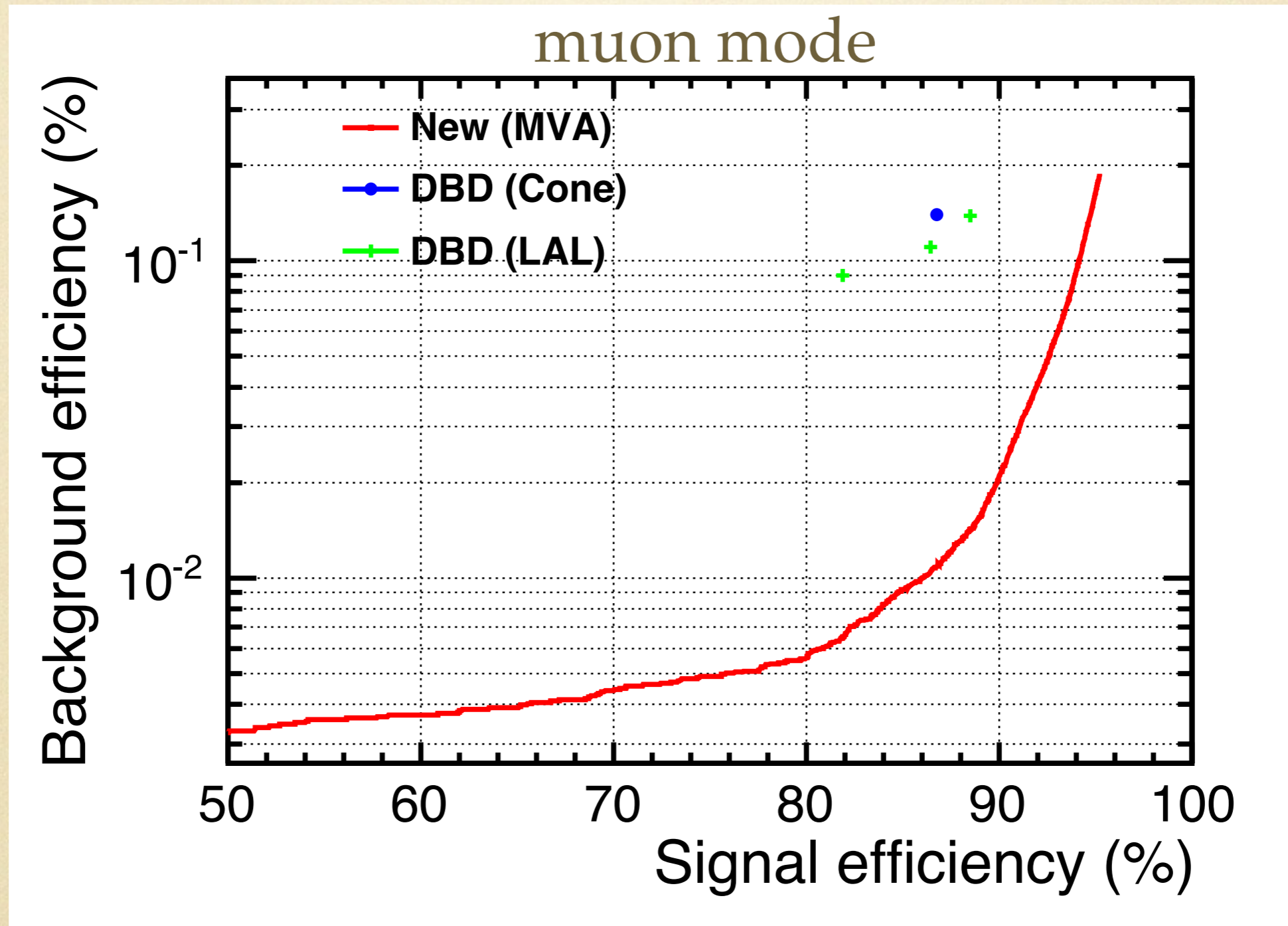
- see “README” for how to use it
- see “example” for steering file
- several weights trained for different processes available
- so far only one processor for single isolated leptons, will be added processor for pair of leptons (Z finder)

today's update:

performance of this new tagging and comparison with DBD ones

Signal (yyxylv) Eff. versus Background (yycyyc) Eff.

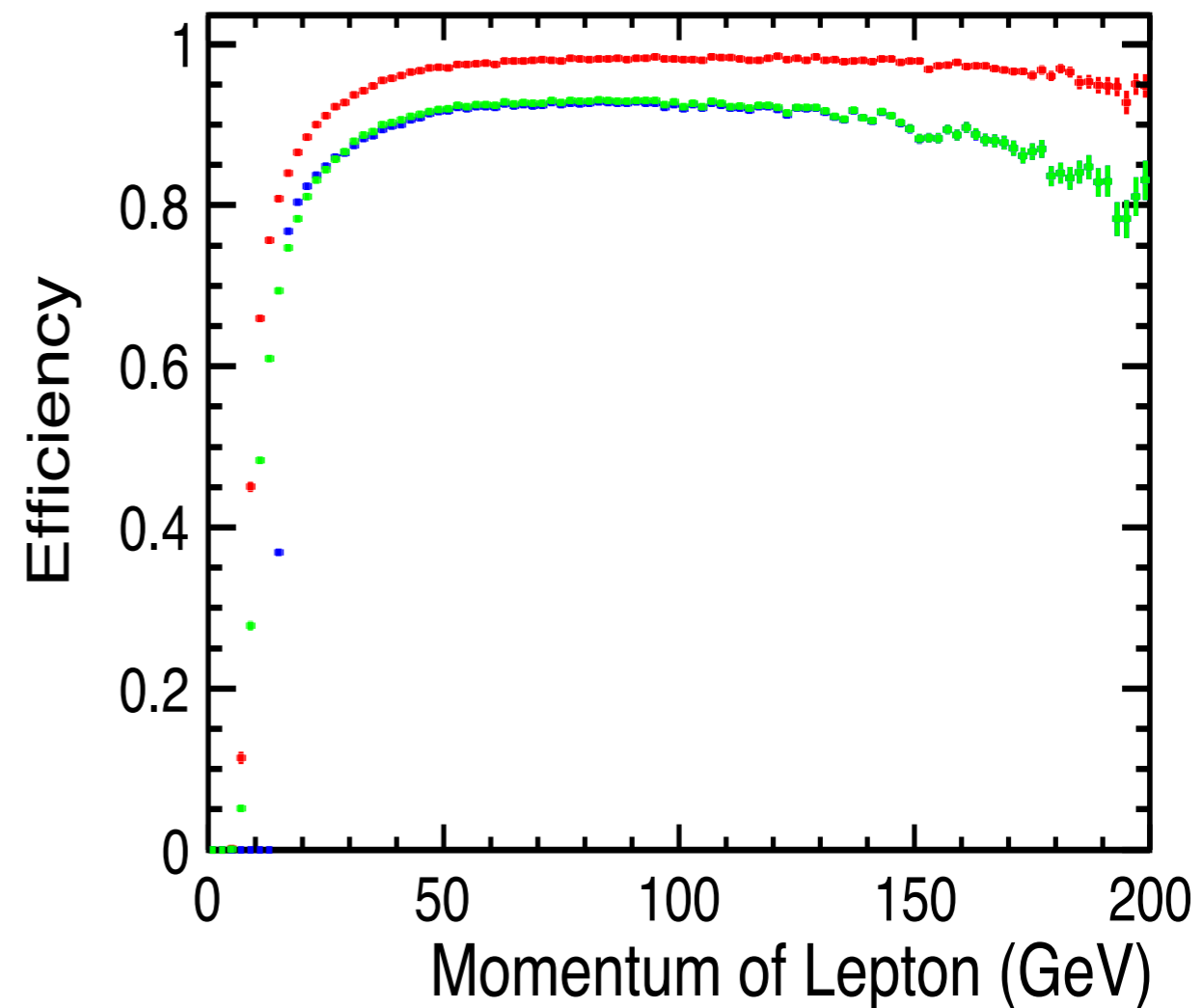
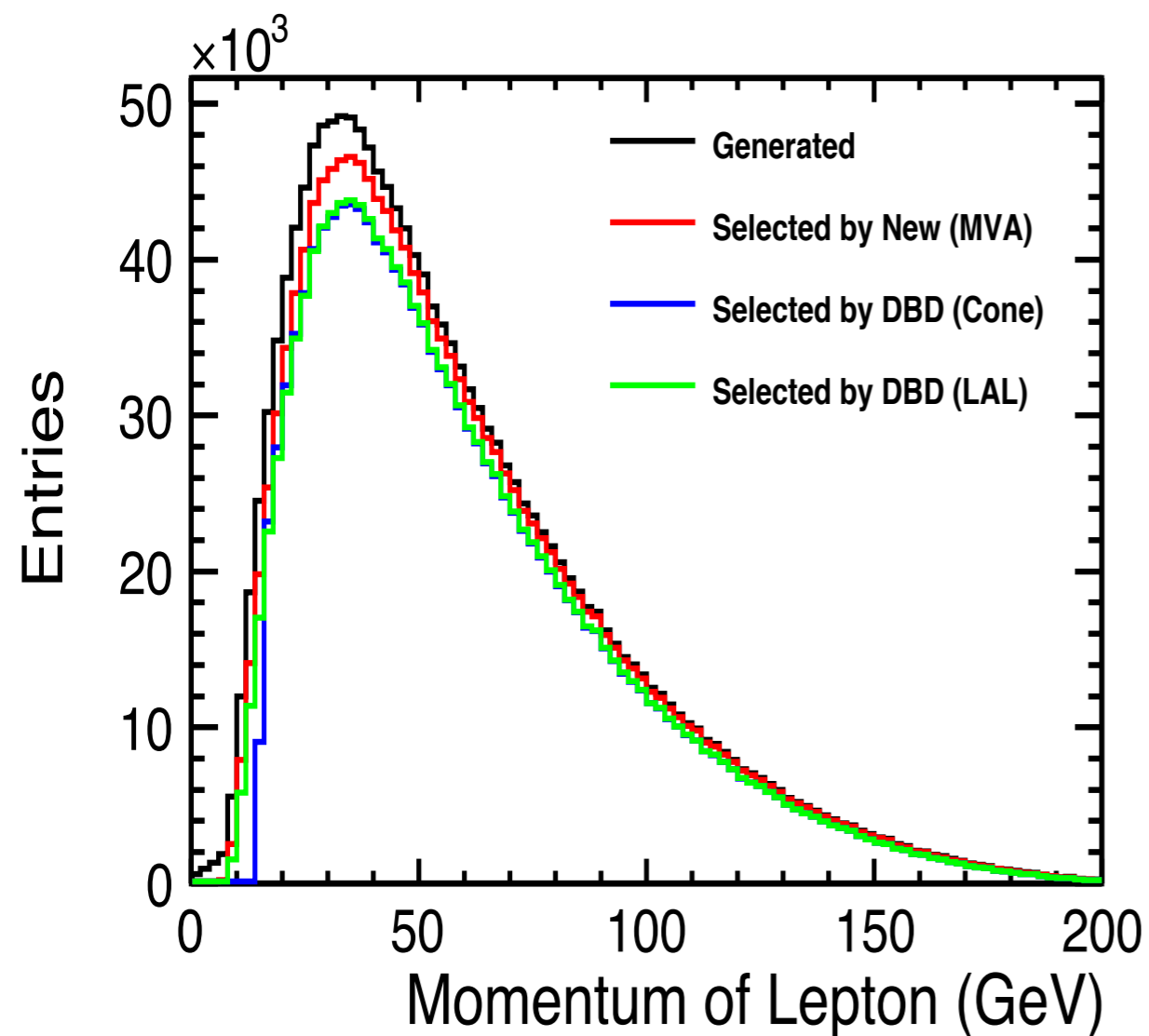
(thank Tino for the high statistics 6f samples)



disclaimer: I used default setting for DBD ones, might be optimised;
for LAL algorithm, I tested using 4/5/6 jets as input

signal efficiency dependence of momentum

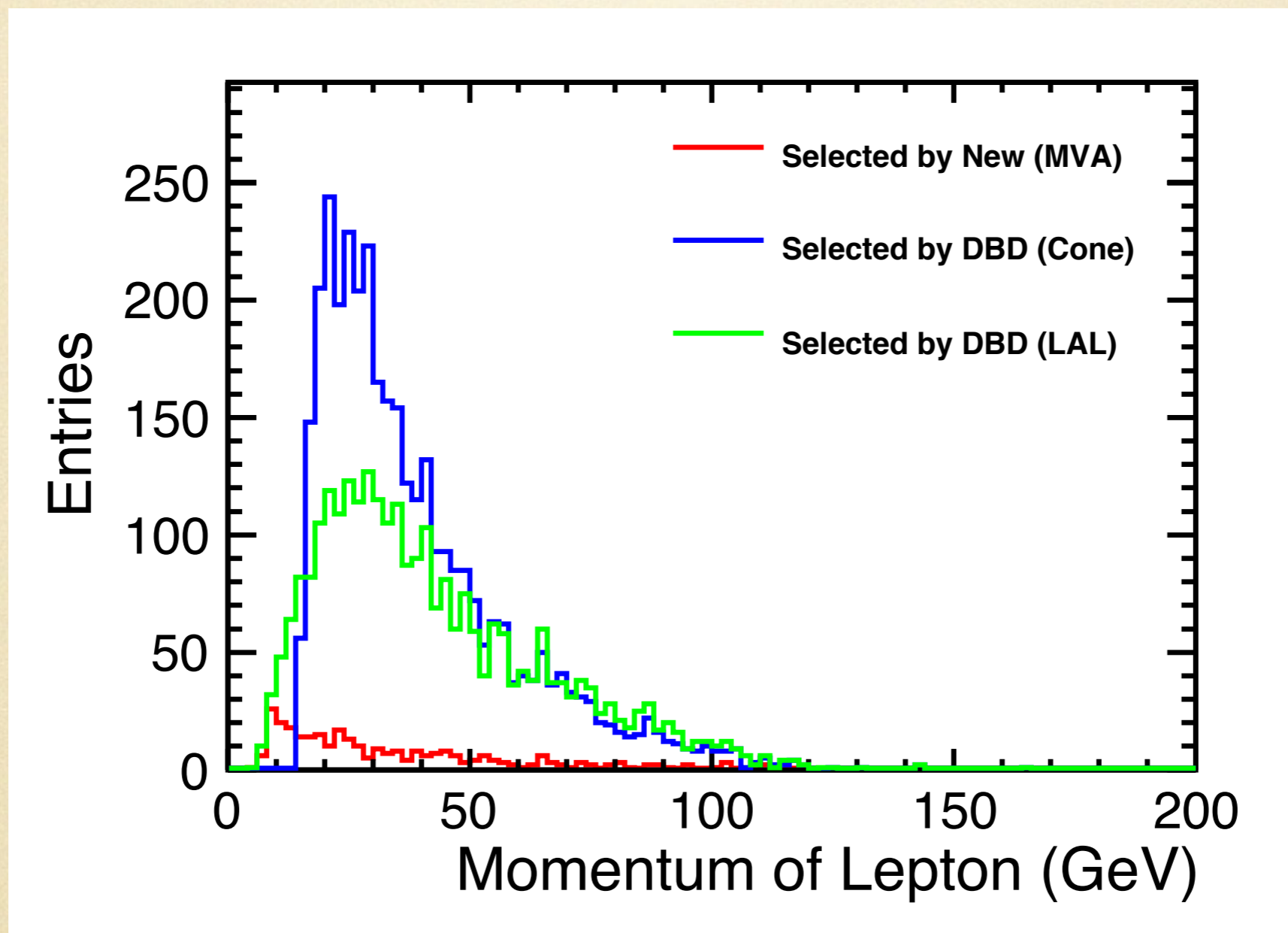
(keep same background eff.)



Eff.(%) / Algorithm :	New(MVA)	DBD(Cone)	DBD(LAL)
Sig (yyxy $\mu\nu$) :	94.65	86.75	88.51
Bkg (yy $\nu\mu\nu$) :	0.134	0.1396	0.1387

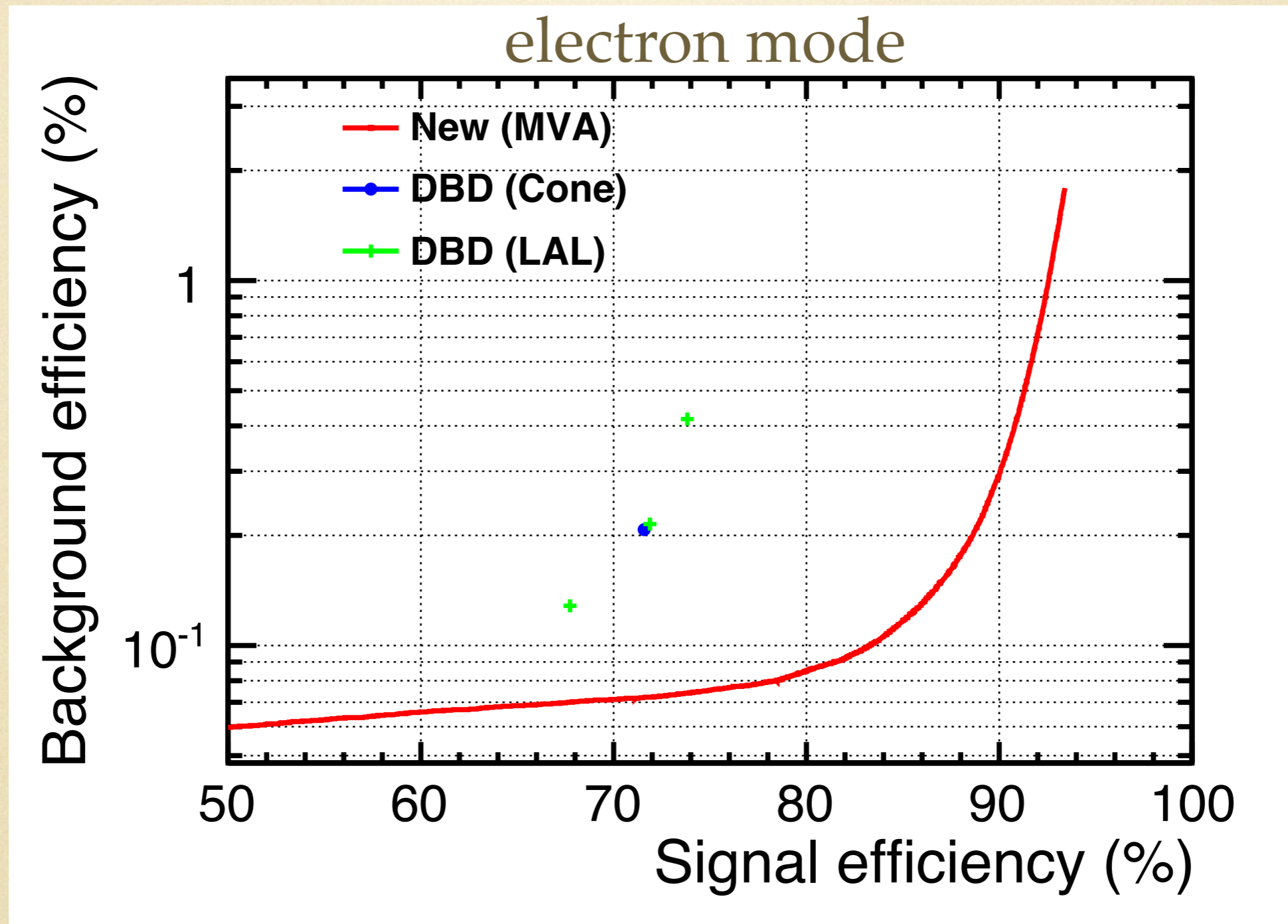
in terms of background suppression

(keep same signal eff.)



Eff.(%) / Algorithm :	New(MVA)	DBD(Cone)	DBD(LAL)
Sig (yyxy $\mu\nu$) :	87.26	86.75	86.44
Bkg (yy ν yc) :	0.01172	0.1396	0.1104

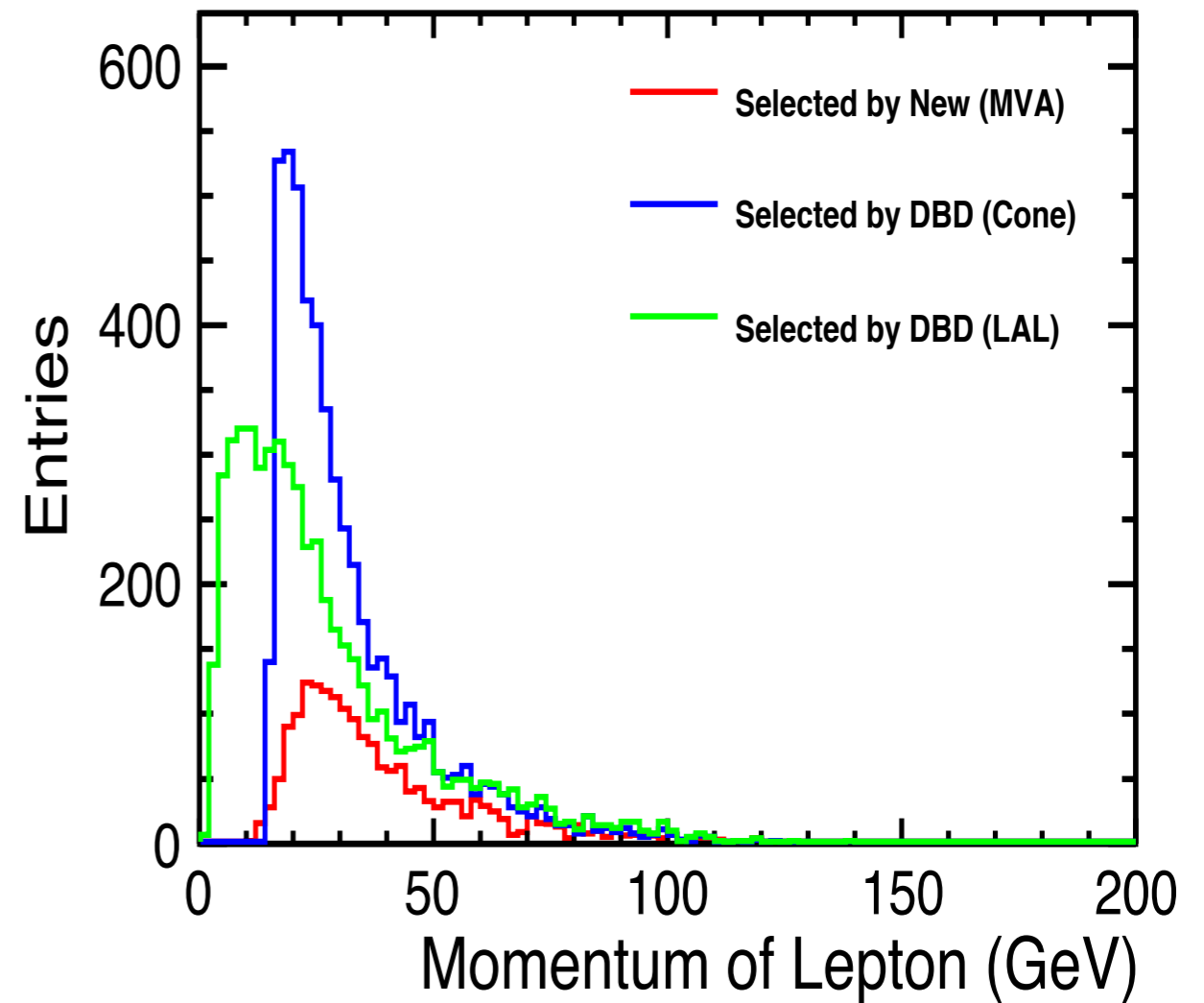
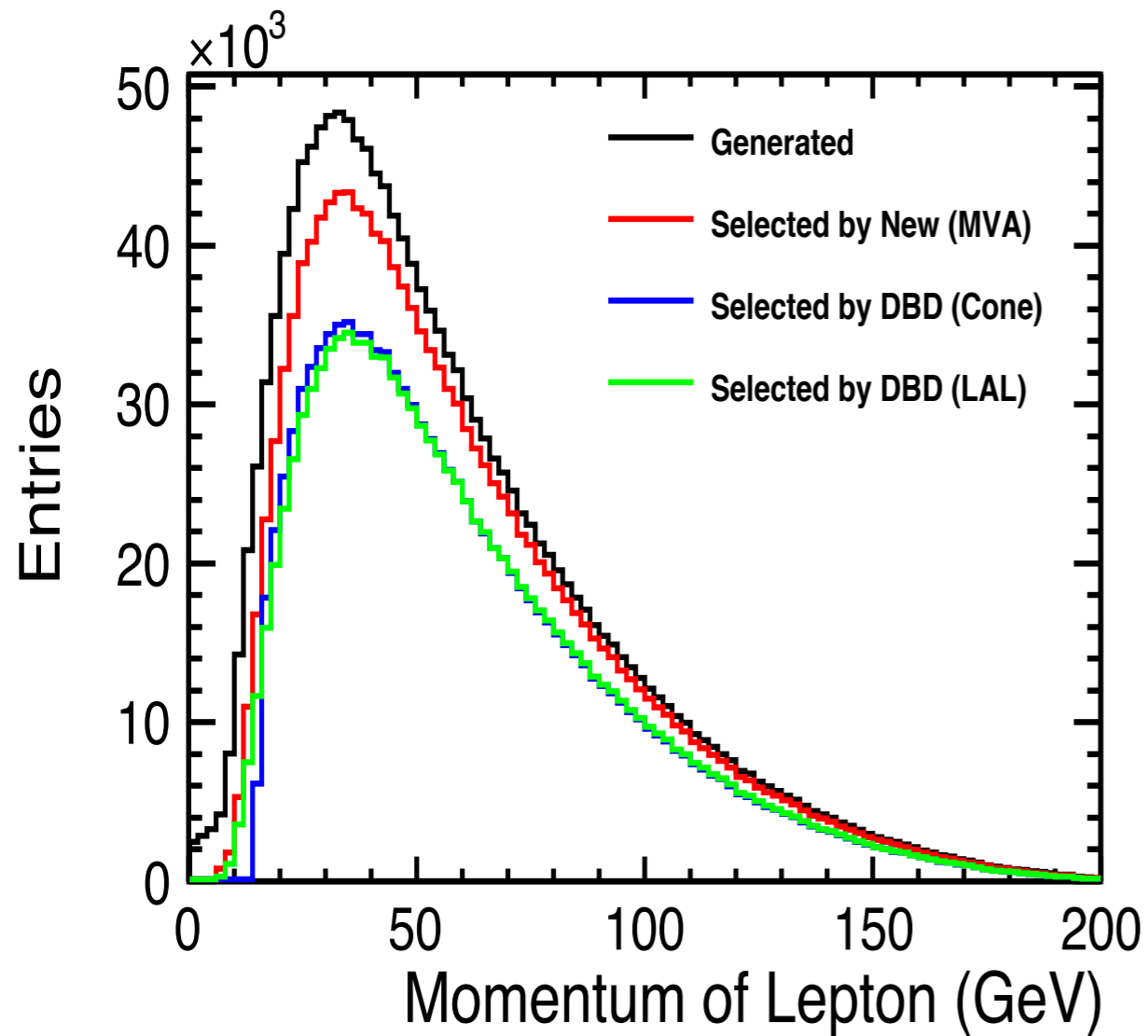
Signal (yyxylv) Eff. versus Background (yycyyc) Eff.



(it seems signal eff. of DBD ones are a bit too low, indicate significant tuning may be possible)

sig/bkg efficiency dependence of momentum

(keep same bkg/sig eff.)

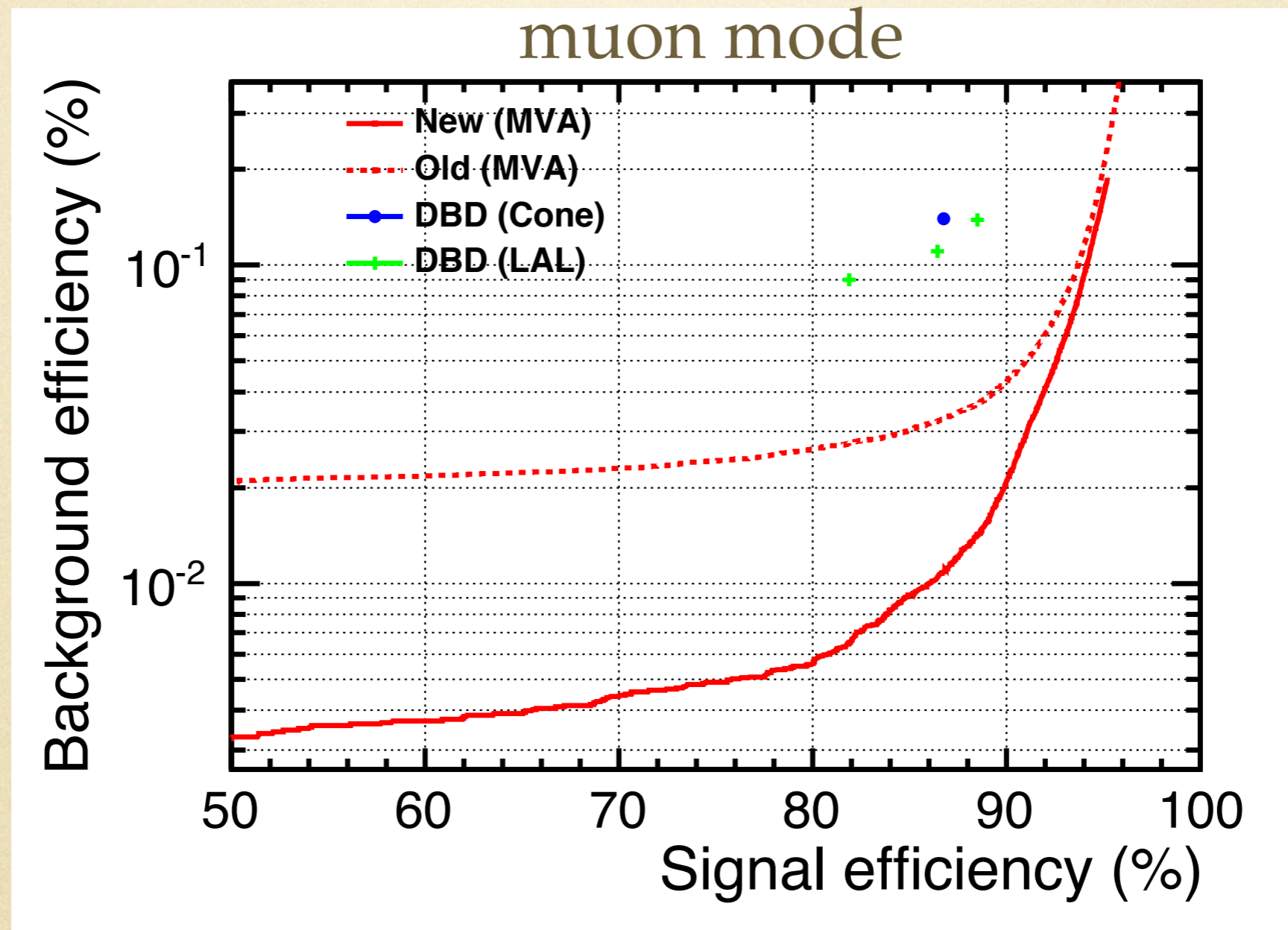


Eff.(%) / Algorithm :	New(MVA)	DBD(Cone)	DBD(LAL)
Sig (yyxyev) :	88.64	71.6	71.89
Bkg (yycyyc) :	0.2015	0.2079	0.2152

New(MVA)	DBD(Cone)	DBD(LAL)
72.17	71.6	71.89
0.07247	0.2079	0.2152

impact of training

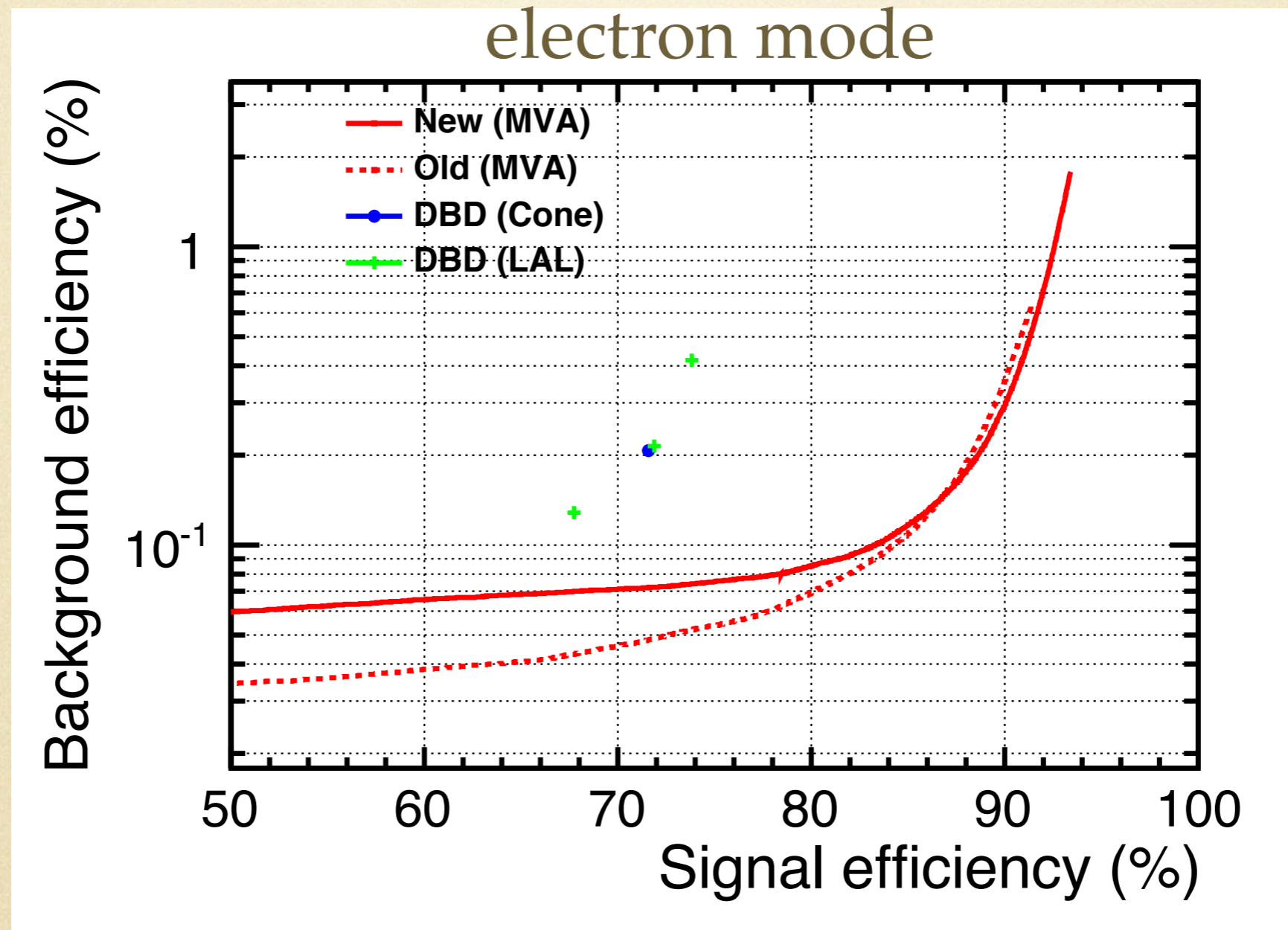
new training: yyxylv / yycyyc with overlay;
old training : yyxylv / bbbb no overlay



~a factor of 2 difference between trainings, in terms of bkg suppression
—> similar performance already confirmed in $\mu\mu HH$ versus $yyxy\mu\nu$

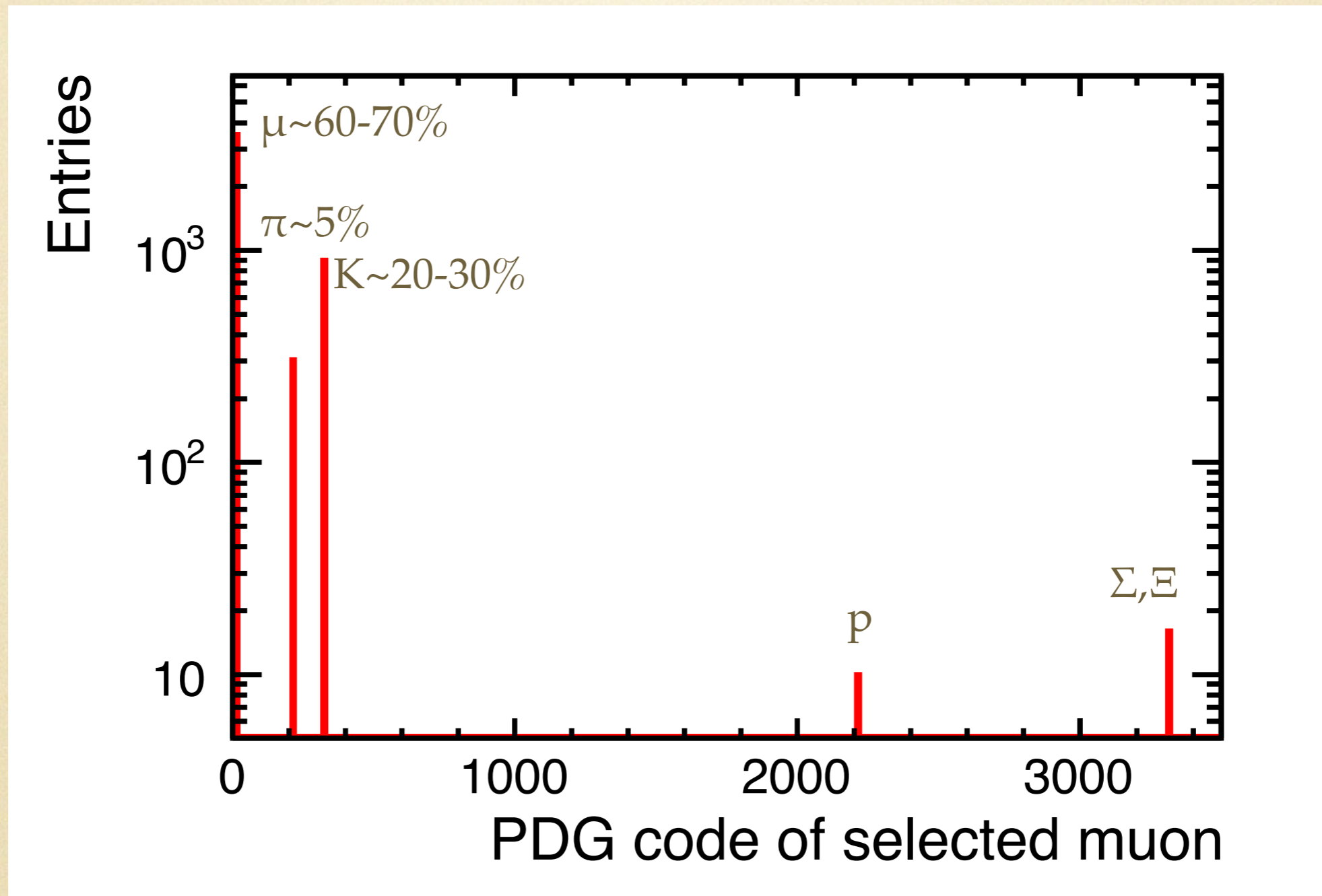
impact of training → puzzle?

new training: yyxylv / yycyyc with overlay;
old training : yyxylv / bbbb no overlay



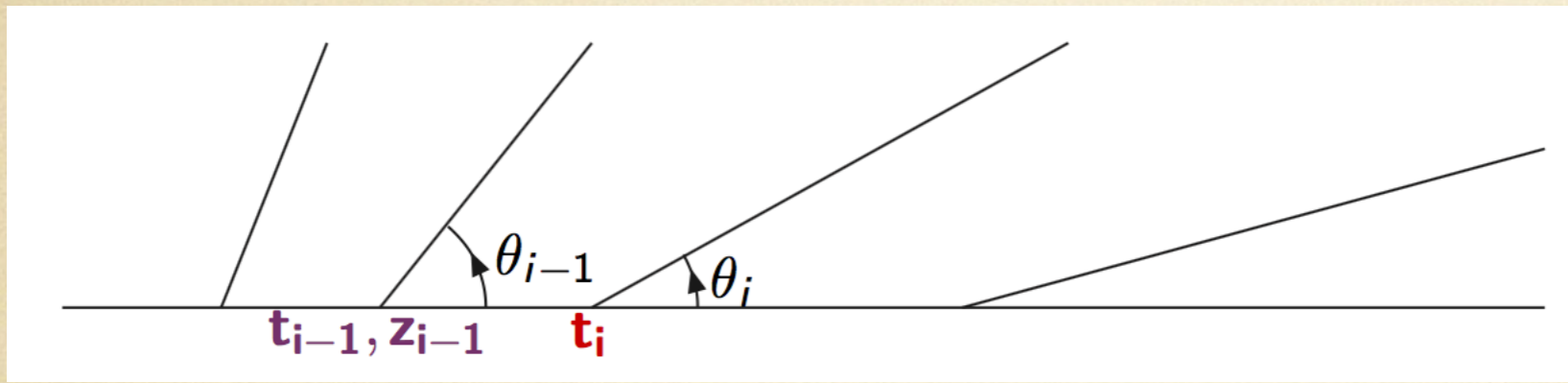
old training gives even better bkg suppression when sig eff is low, strange?

for next step: what kind of particles are still mis-tagged?



K is tagged mostly because $K \rightarrow \mu \nu$ in Calorimeter

proposal of new clustering: reconstruct the parton shower history



$$d\mathcal{P}_a = \sum_{b,c} \frac{\alpha_{abc}}{2\pi} P_{a \rightarrow bc}(z) dt dz .$$

$$P_{q \rightarrow qg}(z) = C_F \frac{1+z^2}{1-z} ,$$

$$P_{g \rightarrow gg}(z) = N_C \frac{(1-z(1-z))^2}{z(1-z)} ,$$

$$P_{g \rightarrow q\bar{q}}(z) = T_R (z^2 + (1-z)^2) ,$$

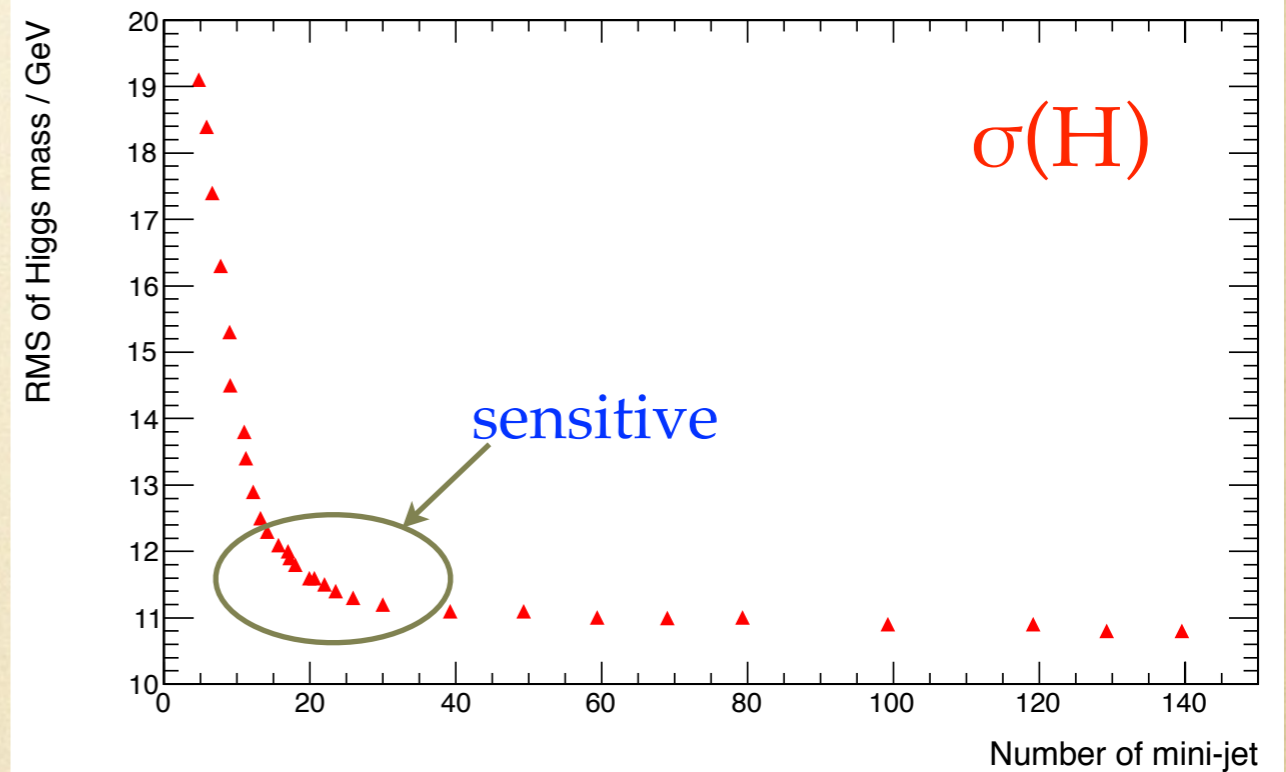
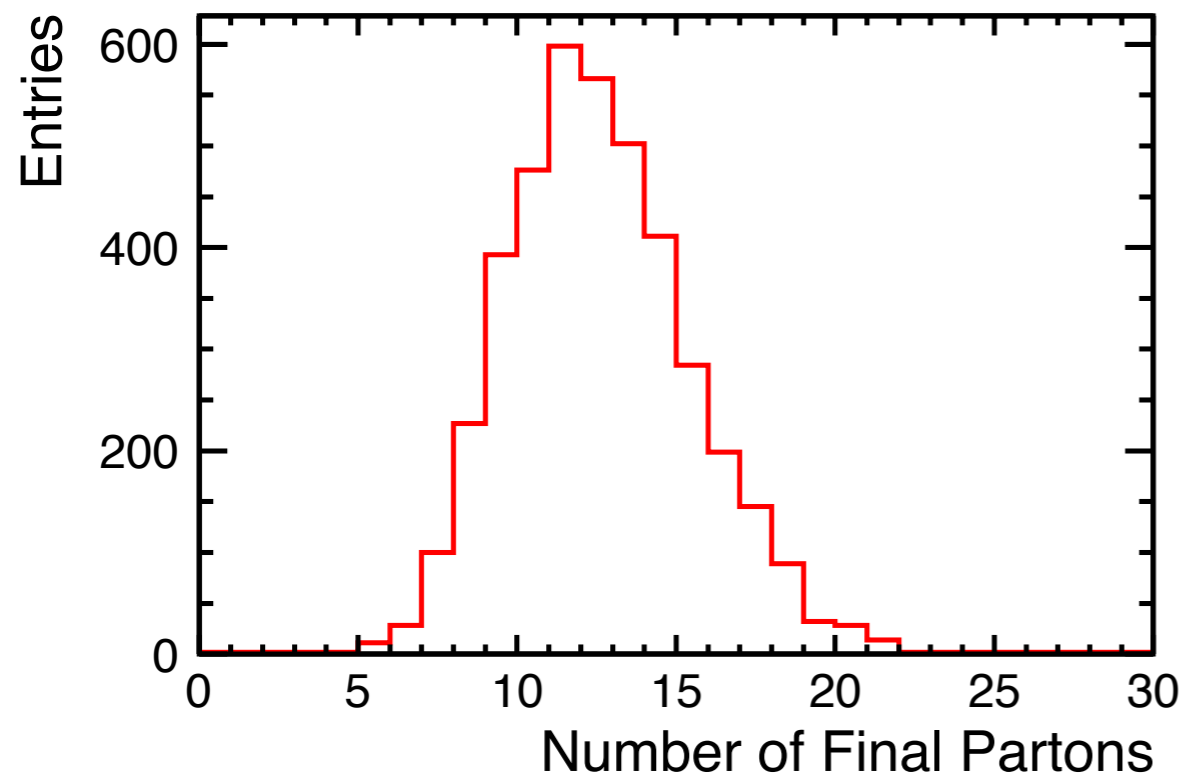
$$P_{q \rightarrow q\gamma}(z) = e_q^2 \frac{1+z^2}{1-z} ,$$

$$P_{l \rightarrow l\gamma}(z) = e_l^2 \frac{1+z^2}{1-z} ,$$

- do mini-jet clustering until #mini-jet ~ 20 , during which Durham algorithm works well \rightarrow which also helps retain infrared safe
- new algorithm comes in to combine the mini-jets: if these mini-jets are actually the relic of parton shower, we should aim for the reconstruction of parton shower history as the maximum information
- generic feature of parton shower can be used to help: angular ordering, $\theta_{i-1} < \theta_i < \theta_{i+1}$, $t_{i-1} < t_i < t_{i+1}$.
- assign each branching with a probability $P_{q \rightarrow qg}$, etc.
- above is intraJet parton shower; we can use some feature of color correlation interJet, such as rapidity gap, etc. (not really sure).

a brief update: relation between mini-jet and parton?

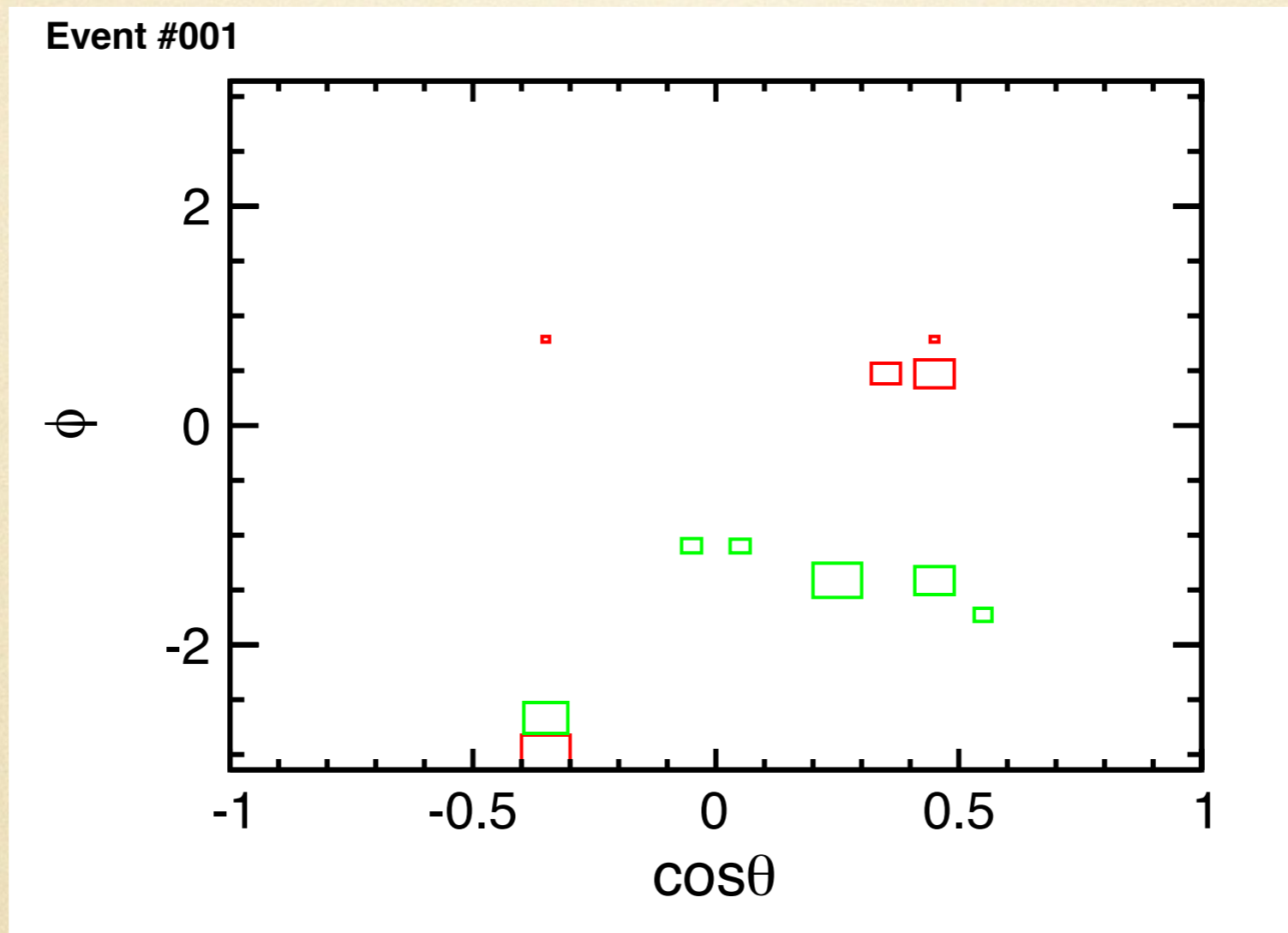
sample: $e^+ e^- \rightarrow \nu\nu HH \rightarrow \nu\nu bbb$ @ 500 GeV



we are on the right track, or just coincidence?

the kinematics of final partons

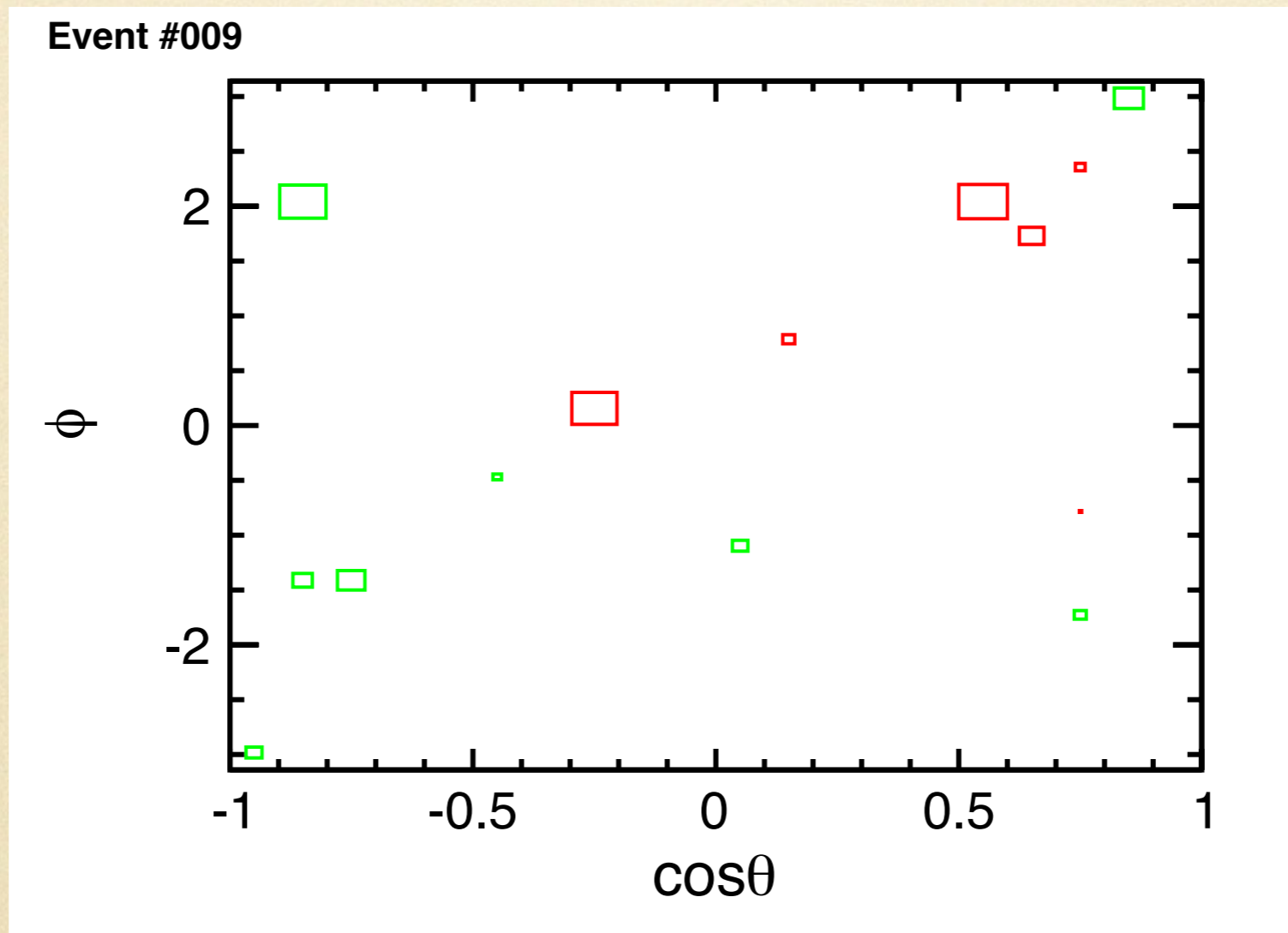
(energy distribution in 2D of ϕ .vs. $\cos\theta$)



red: from H1
green: from H2

the kinematics of final partons

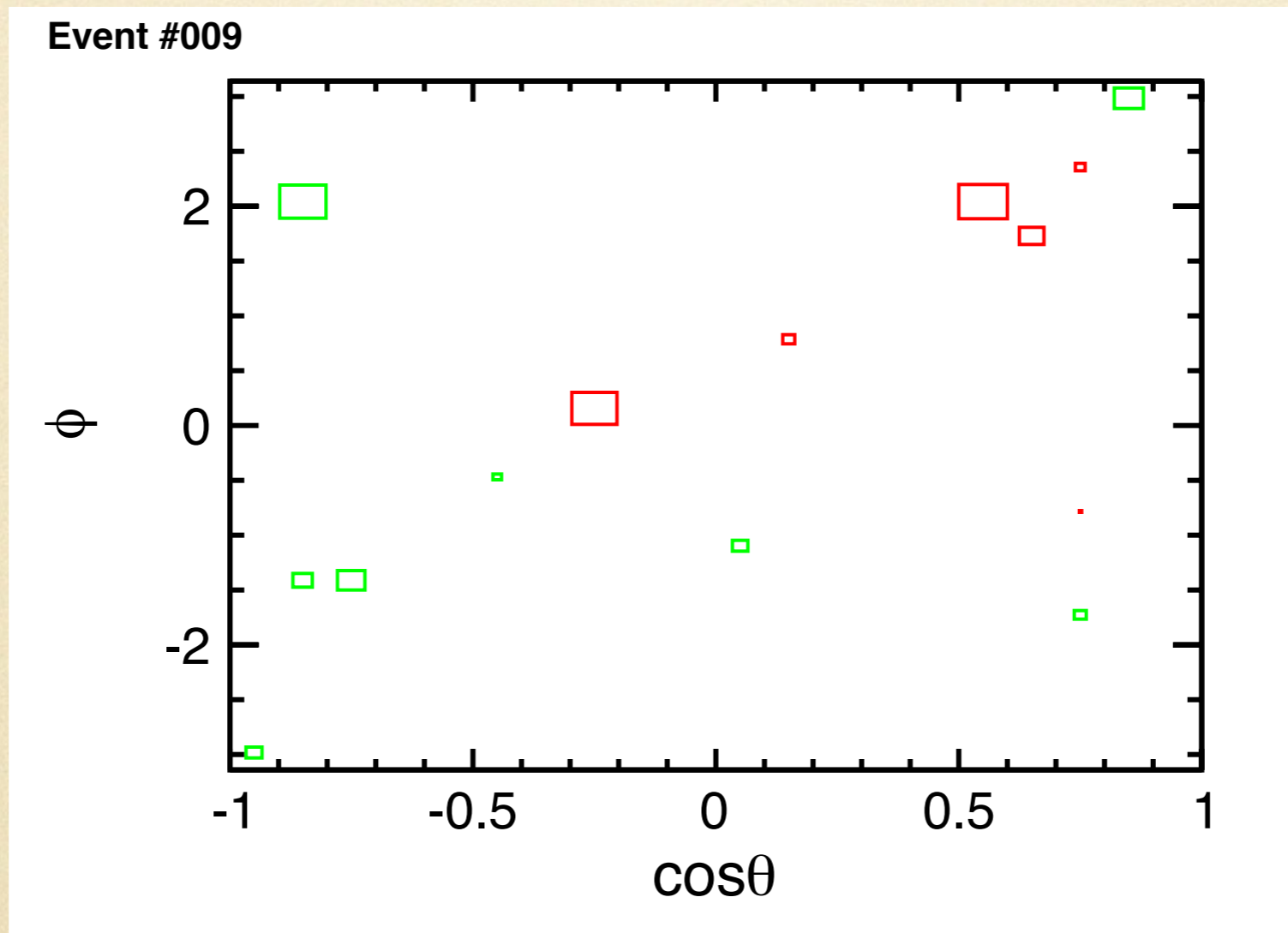
(energy distribution in 2D of ϕ .vs. $\cos\theta$)



red: from H1
green: from H2

the kinematics of final partons

(energy distribution in 2D of ϕ .vs. $\cos\theta$)

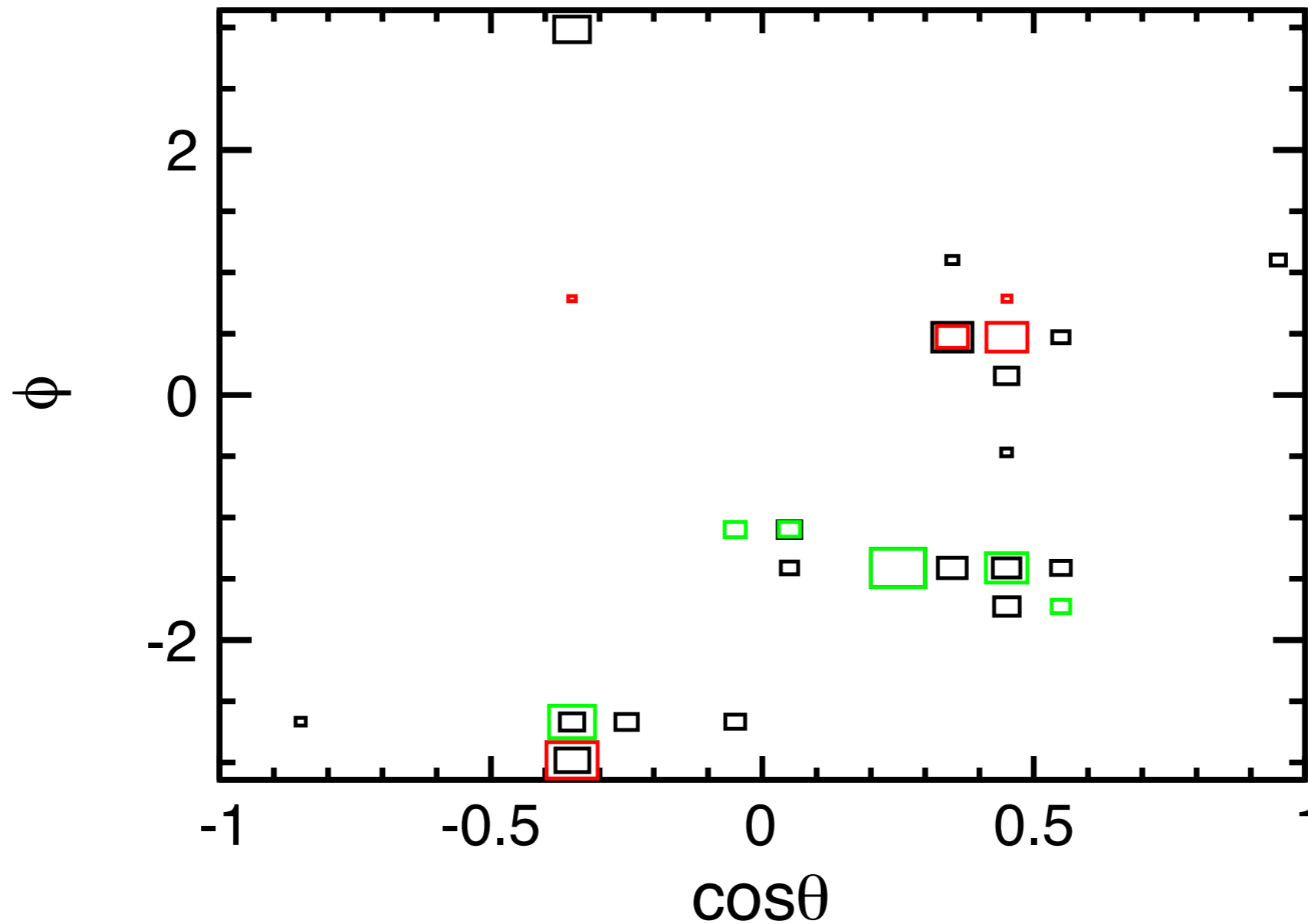


red: from H1
green: from H2

the kinematics of mini-jets versus final partons

(energy distribution in 2D of ϕ .vs. $\cos\theta$)

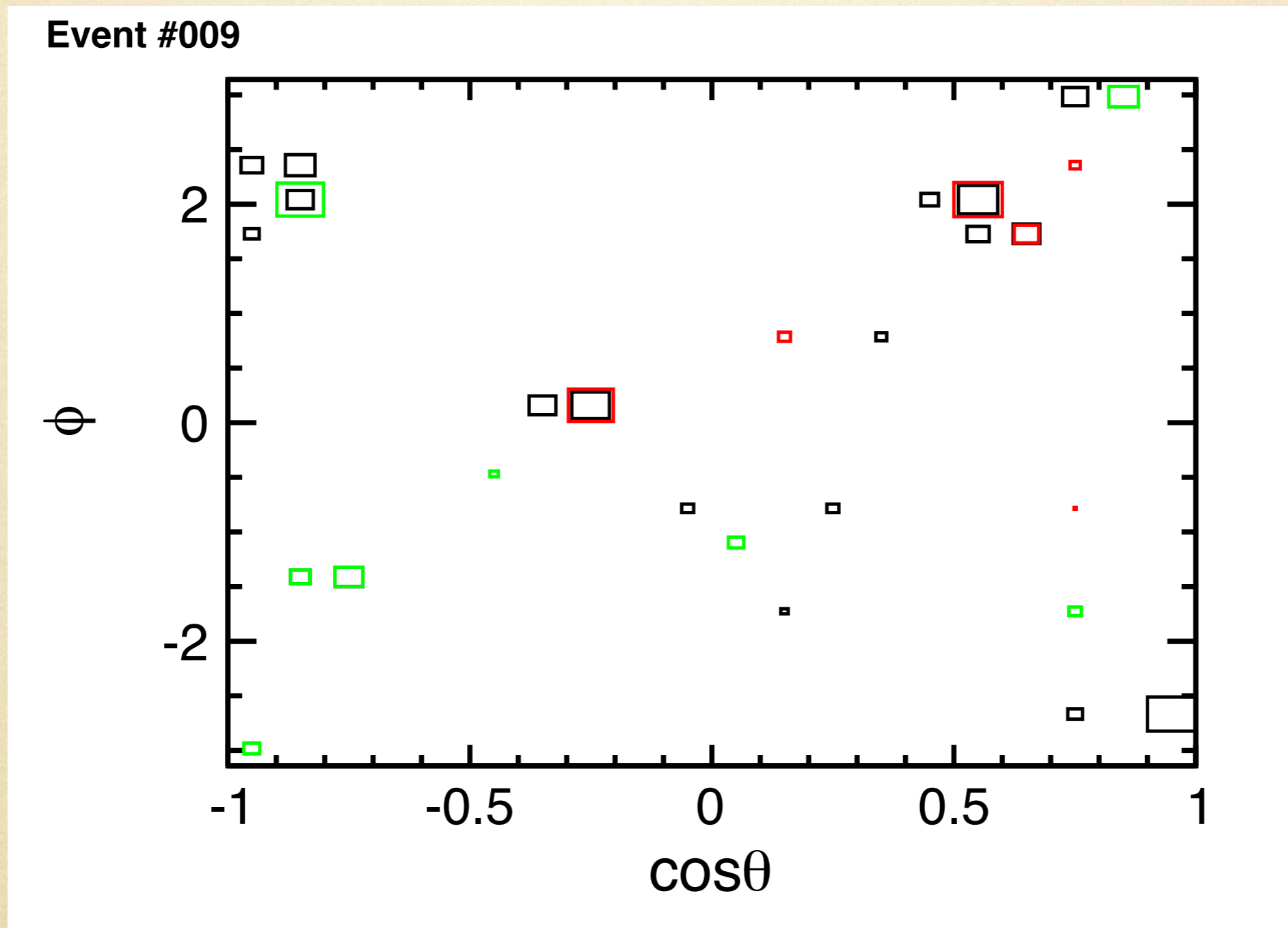
Event #001



black: mini-jets; red: from H1; green: from H2

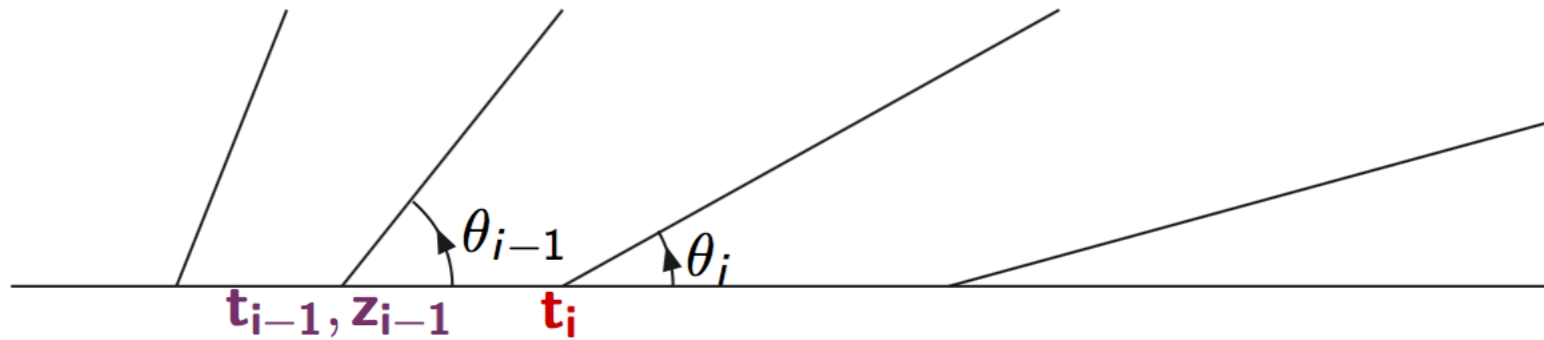
the kinematics of mini-jets versus final partons

(energy distribution in 2D of ϕ .vs. $\cos\theta$)



mini-jets idea looks promising, at least theoretically!

next step: try to implement the likelihood



$$d\mathcal{P}_a = \sum_{b,c} \frac{\alpha_{abc}}{2\pi} P_{a \rightarrow bc}(z) dt dz .$$

$$P_{q \rightarrow qg}(z) = C_F \frac{1+z^2}{1-z} ,$$

$$P_{g \rightarrow gg}(z) = N_C \frac{(1-z(1-z))^2}{z(1-z)} ,$$

$$P_{g \rightarrow q\bar{q}}(z) = T_R (z^2 + (1-z)^2) ,$$

$$P_{q \rightarrow q\gamma}(z) = e_q^2 \frac{1+z^2}{1-z} ,$$

$$P_{\ell \rightarrow \ell\gamma}(z) = e_\ell^2 \frac{1+z^2}{1-z} ,$$

theoretical paper having similar idea- [arXiv:1102.3480](https://arxiv.org/abs/1102.3480)