

Strange quark tagging with ILD to search for new physics in the Higgs sector

Higgs 2021 – October 18-22, 2021

YSF Plenary Track – October 21, 2021 – [Indico](#)

Presented by **Matthew Basso** (University of Toronto),
on behalf of everyone on the **Snowmass 2021 LoI**
and the **ILD Collaboration**



UNIVERSITY OF
TORONTO



SnowMass2021



Overview

- Submitted a **Letter of Interest** as part of **Snowmass 2021**
 - Basic goal: assess sensitivity of Higgs to strange couplings with ILD@ILC and set constraints on detector design
 - In line with ILC Snowmass 2021 study questions ([2007.03650](#))
 - Interplay with the instrumentation: strange tagging capabilities strongly depend on the detector (e.g., PID)

Strange quark as a probe for new physics in the Higgs Sector

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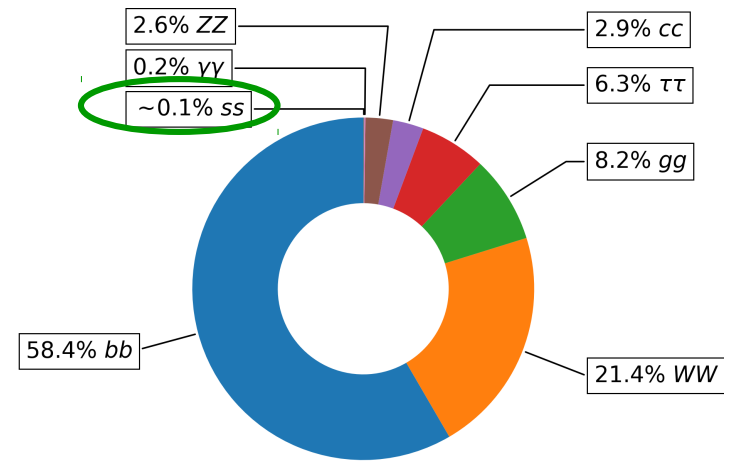
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(b) SLAC National Accelerator Laboratory, Stanford CA – USA

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(d) University of Oregon, Eugene OR – USA

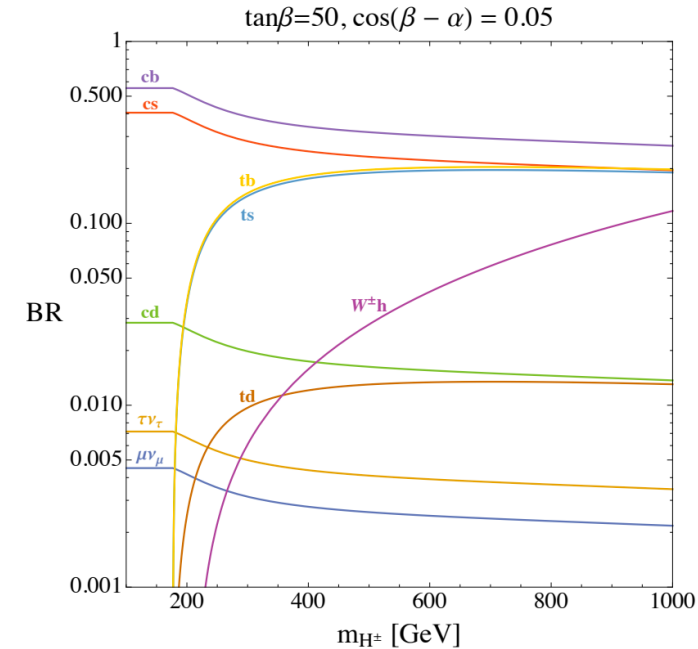
(e) High Energy Accelerator Research Organization, Tsukuba – Japan



$\sqrt{s} = 13 \text{ TeV}, m_H = 125 \text{ GeV}$

$H \rightarrow ss$ and $H \rightarrow cs$

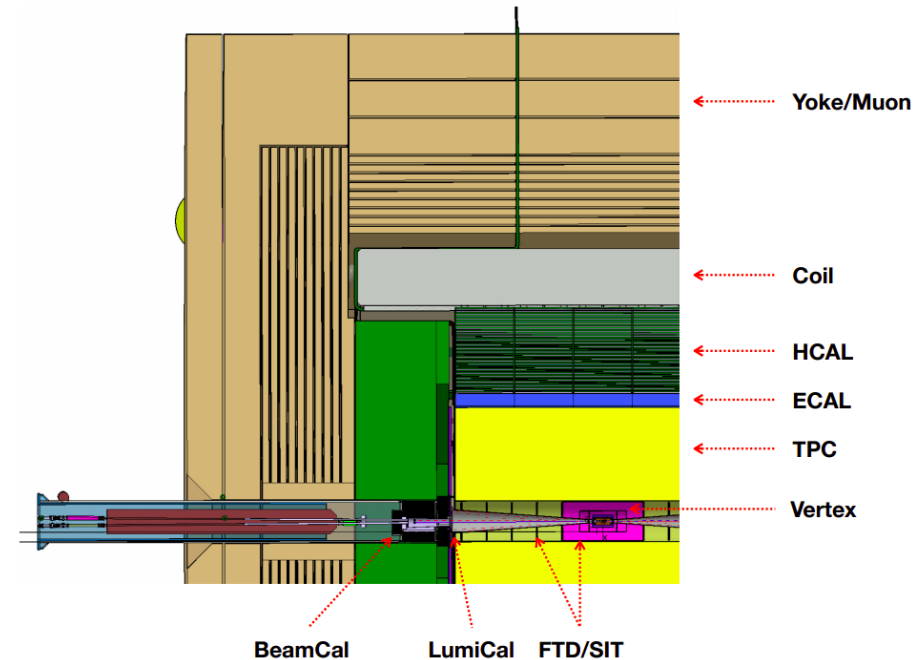
- $H \rightarrow ss$: extremely challenging unless enhanced relative to SM expectations
- $H \rightarrow cs$: some BSM models allow for the 1st and 2nd generation fermion masses to be an additional source of EW symmetry breaking
 - Result in “SM” and “heavy” Higgs doublets
 - Predicts an **enhancement** to Higgs cross section
 - Charged heavy Higgs can undergo flavour violating decays (e.g., cs) – s/c -tagging can help



Charged heavy Higgs branching ratios. Taken from Fig. 6 of [1610.02398](#).

The International Large Detector

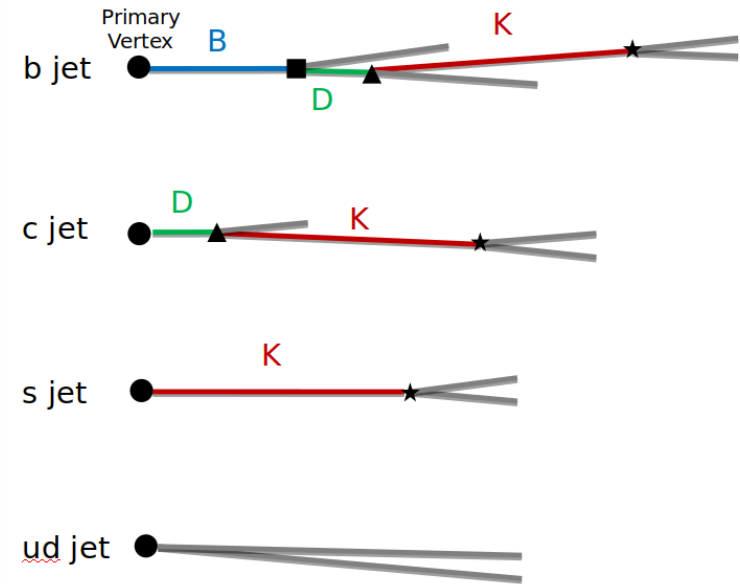
- ILC: $\sqrt{s} = 250 \text{ GeV}$, 2000 fb^{-1} ([1809.09504](#))
- The ILD detector
 - 3 double-layer pixel detectors for vertexing
 - Time projection chamber (TPC) for tracking with inner/outer Si layers
 - Low material assists in low- p tracking
 - High granularity sampling calorimeters for particle flow reconstruction
 - Challenge is reconstructing neutral hadrons
 - Precise EM/hadronic design still under study
 - Tracking/calorimetry contained in 3.5 T field



ILD detector quadrant. Taken from Fig. 1 of [1912.04601](#).

Jet flavour tagging classification

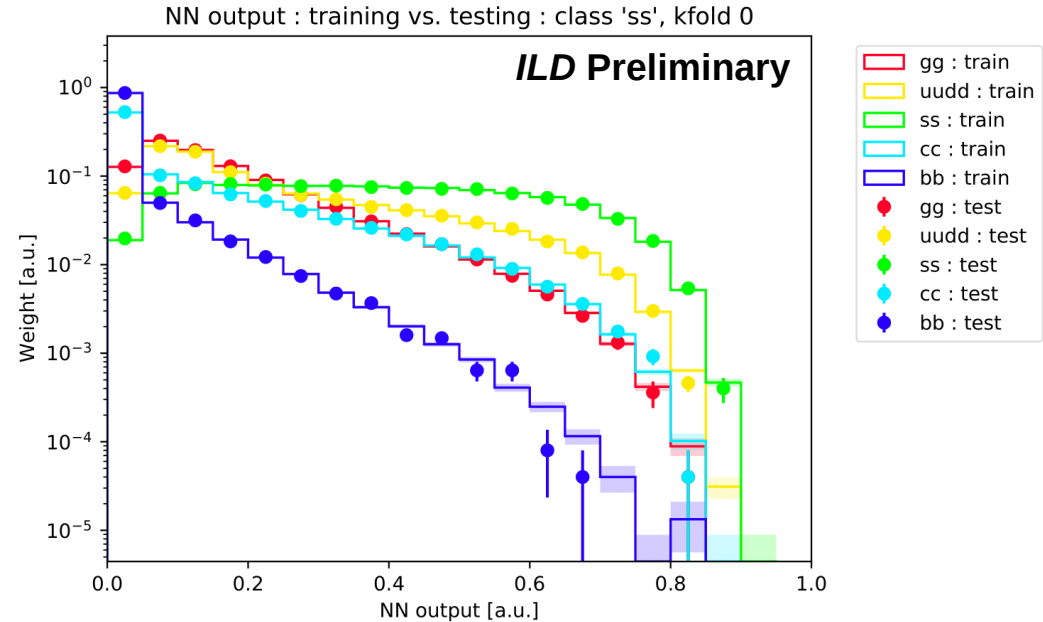
- Use a *neural network*-based tagger (architecture in [Backup](#)) for classifying jets by flavour
- Train on ILD-reconstructed $(Z \rightarrow \text{inv})(H \rightarrow qq/gg)$ samples (100% polarized)
- Use per-jet level inputs as well as variables on the 10 leading particles in each jet:
 - Jets: momentum p , pseudorapidity η , polar angle ϕ , mass m , *b/c/o*-tagger scores, category, $N_{\text{particles}}$
 - Particles: p , η , ϕ , m , charge, **truth** electron/muon/pion/kaon/proton likelihoods (0 or 1, using PDG ID – “kaons” include K_S^0 , $K^{+/-}$, and Λ)



Different jet types. Picture borrowed from T. Tanabe's slides, see [Backup](#).

Tagger performance

- Shown is the strange score output – **good** train-test agreement
- Good discrimination of s jets from u/d jets – likely comes from using truth likelihoods
 - Also good discrimination of s jets from g jets – here, $N_{\text{particles}}$ is powerful



Train-test agreement for all output nodes in Backup (ROC curves too).

$H \rightarrow ss$ analysis

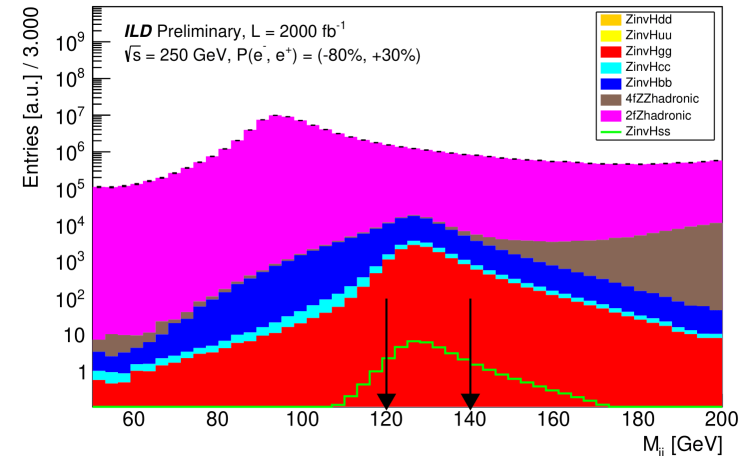
- Performed on same $H \rightarrow qq/gg$ samples (500K events per flavour) as well as $Z \rightarrow qq$ and $ZZ \rightarrow qqqq$ samples (~1M events each)
 - Scale $\text{BR}[H \rightarrow cc]$ by ratio of s/c quark mass ratio squared: $\text{BR}[H \rightarrow ss] \sim 2\text{E-}4$
- Kinematic selection:
 - Jet quantities: leading/subleading jet momenta, p_j ; dijet mass, M_{jj} ; dijet energy, E_{jj}
 - Missing 4-vector quantities: mass, M_{miss} ; angular separation, $\Delta R_{jj,\text{miss}} = \sqrt{(\Delta\phi_{jj,\text{miss}})^2 + (\Delta\eta_{jj,\text{miss}})^2}$
 - Leading/subleading $b/c/o$ -tagger ([1506.08371](#)) scores and jet category
 - Number of Particle Flow Objects (PFOs): per event, $N_{\text{PFOs}}/\text{event}$; per jet, $N_{\text{PFOs}}/\text{jet}$

Cutflow

ILD Preliminary, $\mathcal{L} = 2000 \text{ fb}^{-1}$, $\sqrt{s} = 250 \text{ GeV}$, $P(e^-, e^+) = (-80\%, +30\%)$

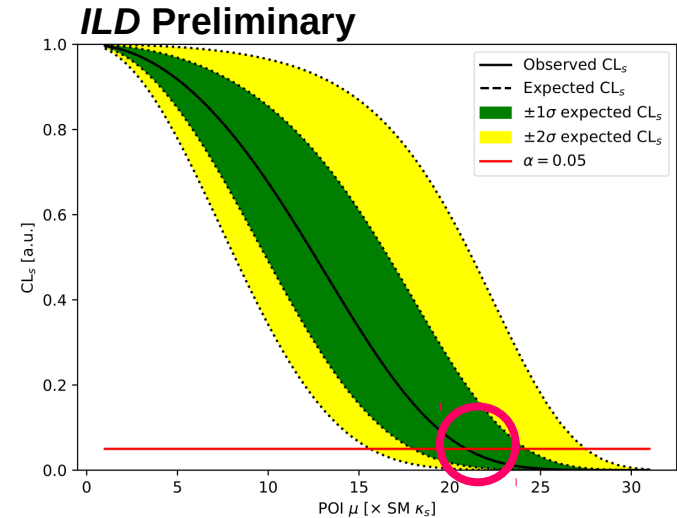
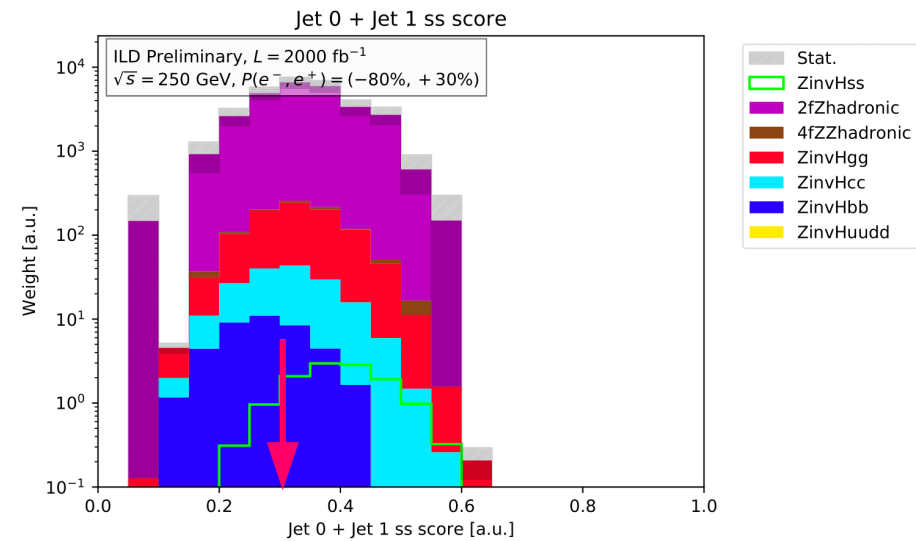
	$(H \rightarrow s\bar{s})(Z \rightarrow \nu\nu)$	$(H \rightarrow gg)(Z \rightarrow \nu\nu)$	$(H \rightarrow u\bar{u}/d\bar{d})(Z \rightarrow \nu\nu)$	$(H \rightarrow c\bar{c})(Z \rightarrow \nu\nu)$	$(H \rightarrow b\bar{b})(Z \rightarrow \nu\nu)$	$Z \rightarrow q\bar{q}$	$ZZ \rightarrow q\bar{q}q\bar{q}$	Sig. eff.	Bkg. eff.
No cut	42.65 ± 0.06	17254.17 ± 24.41	0.59 ± 0.0	5858.77 ± 8.29	116168.67 ± 164.29	$176876516.6 \pm 161411.64$	1342206.08 ± 1338.33	$1.00\text{e}+00$	$1.00\text{e}+00$
No leptons	42.55 ± 0.06	17225.89 ± 24.39	0.59 ± 0.0	5846.08 ± 8.28	115535.31 ± 163.84	$175328405.19 \pm 160703.71$	1335436.33 ± 1334.95	$9.98\text{e}-01$	$9.91\text{e}-01$
≥ 2 jets	42.55 ± 0.06	17225.89 ± 24.39	0.59 ± 0.0	5846.08 ± 8.28	115535.31 ± 163.84	$175328405.19 \pm 160703.71$	1335436.33 ± 1334.95	$9.98\text{e}-01$	$9.91\text{e}-01$
$p_{j0}, p_{j1} > 30 \text{ GeV}$	39.46 ± 0.06	16424.08 ± 23.81	0.55 ± 0.0	5619.05 ± 8.12	109492.68 ± 159.5	$131310044.43 \pm 139074.89$	1331247.44 ± 1332.86	$9.25\text{e}-01$	$7.44\text{e}-01$
$M_{jj} \in [120, 140] \text{ GeV}$	29.75 ± 0.05	12459.56 ± 20.74	0.42 ± 0.0	3883.41 ± 6.75	63849.78 ± 121.8	7424895.55 ± 33070.82	8041.49 ± 103.59	$6.97\text{e}-01$	$4.21\text{e}-02$
$E_{jj} \in [125, 160] \text{ GeV}$	29.62 ± 0.05	12401.25 ± 20.69	0.42 ± 0.0	3862.38 ± 6.73	63407.65 ± 121.38	4027593.77 ± 24356.93	6111.86 ± 90.31	$6.94\text{e}-01$	$2.31\text{e}-02$
$M_{\text{miss}} \in [75, 120] \text{ GeV}$	27.56 ± 0.05	11614.11 ± 20.02	0.39 ± 0.0	3612.75 ± 6.51	59551.31 ± 117.63	867590.51 ± 11304.65	2105.79 ± 53.01	$6.46\text{e}-01$	$5.30\text{e}-03$
$\Delta R_{jj, \text{miss}} < 4$	23.82 ± 0.05	10039.07 ± 18.62	0.34 ± 0.0	3124.94 ± 6.05	51512.9 ± 109.4	151865.16 ± 4729.65	1537.31 ± 45.29	$5.58\text{e}-01$	$1.22\text{e}-03$
$\text{score}^b/\text{jet} < 0.2$	22.2 ± 0.04	8593.49 ± 17.22	0.32 ± 0.0	1917.39 ± 4.74	551.1 ± 11.32	88968.53 ± 3620.08	689.92 ± 30.34	$5.20\text{e}-01$	$5.65\text{e}-04$
$\text{score}^c/\text{jet} < 0.35$	20.72 ± 0.04	7745.04 ± 16.35	0.3 ± 0.0	302.77 ± 1.88	179.83 ± 6.46	73060.25 ± 3280.5	548.47 ± 27.05	$4.86\text{e}-01$	$4.59\text{e}-04$
$N_{\text{PFOs}}/\text{event} \in [30, 60]$	13.93 ± 0.03	854.7 ± 5.43	0.2 ± 0.0	146.28 ± 1.31	44.14 ± 3.2	33584.15 ± 2224.16	64.05 ± 9.25	$3.27\text{e}-01$	$1.95\text{e}-04$
$N_{\text{PFOs}}/\text{jet} \in [10, 40]$	12.53 ± 0.03	778.96 ± 5.19	0.18 ± 0.0	136.34 ± 1.26	39.96 ± 3.05	26955.7 ± 1992.62	56.05 ± 8.65	$2.94\text{e}-01$	$1.57\text{e}-04$

- Largest **decrease** in signal efficiency at M_{jj} cut
 - Provides one of the strongest handles on reducing $Z \rightarrow q\bar{q}$
 - $H \rightarrow b\bar{b}$ $s/b = 0.00065$ @ No cut – comparable to analysis from T. Ogawa's [thesis](#) (Section 6.2)
- **29% signal, 0.016% background efficiency**
 - All histograms in [Backup](#)



Limits on coupling strength modifier

- Cut on (0.5x) sum of strange scores for leading and subleading jets >0.3 , generated limits for modifier to SM BR
 - Asymptotic significance $\sim 0.1\sigma$ (see Backup)
- 95% upper confidence bounds at $\sim 21 \times \text{SM } \kappa_s$ (in the kappa framework, κ_s^2 is the modifier to $\text{BR}[H \rightarrow ss]$)
 - N.B.: observed is *identical* (by design) to expected in the plot
 - Sigma bands around expected decrease with smaller cut values (better $Z \rightarrow qq$ MC stats)



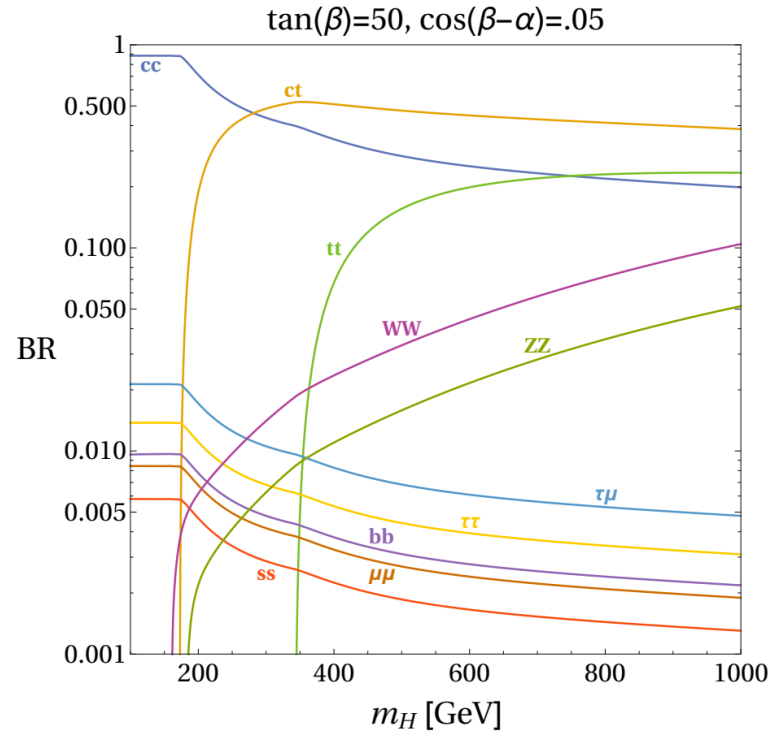
Discussion and outlook

- Discovery measurement seems *unlikely* – looking at best case tagger
 - For 30% signal efficiency, need **10,000x** better background rejection
 - Set limits on coupling strength modifier κ_s at **O(20) x SM prediction** using 2000 fb⁻¹ of data at $\sqrt{s} = 250$ GeV (combined limits for ILC and other future colliders in [Backup](#))
- Gains would come from **reducing** the $Z \rightarrow qq$ background
 - As a suggestion from the ILD community, quantities like $\Delta\phi_{jj}$ or p_{τ^j} should help
 - More statistics are available for the $Z \rightarrow qq$ and $ZZ \rightarrow qqqq$ backgrounds
- May try and provide prospects for BSM 2HDM $H \rightarrow cs$ or $H(125) \rightarrow bs$ decays
- Work will be documented in a paper as part of Snowmass 2021

Questions?

Backup

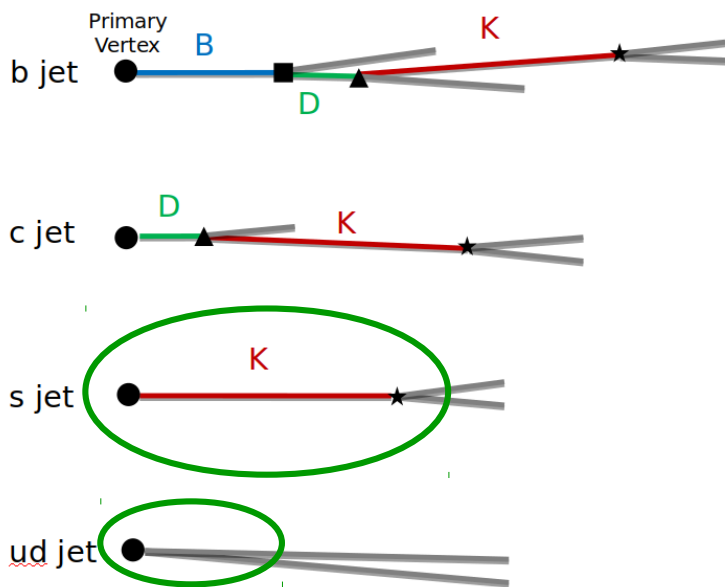
Neutral heavy Higgs BRs



Neutral heavy Higgs branching ratios. Taken from Fig. 3 of [1610.02398](#).

Different jet types, pictorially

Discriminants



Charged Kaon track

- Zero track impact parameter w.r.t. primary vertex
- Momentum fraction relative to the jet momentum carried by the leading Kaon
 - (Longitudinal vs transverse components?)

$\mathbf{V}^0(K_S^0, \Lambda^0)$

- Vertex momentum & displacement must point in the same direction
- Mean vertex distance smaller compared to b/c

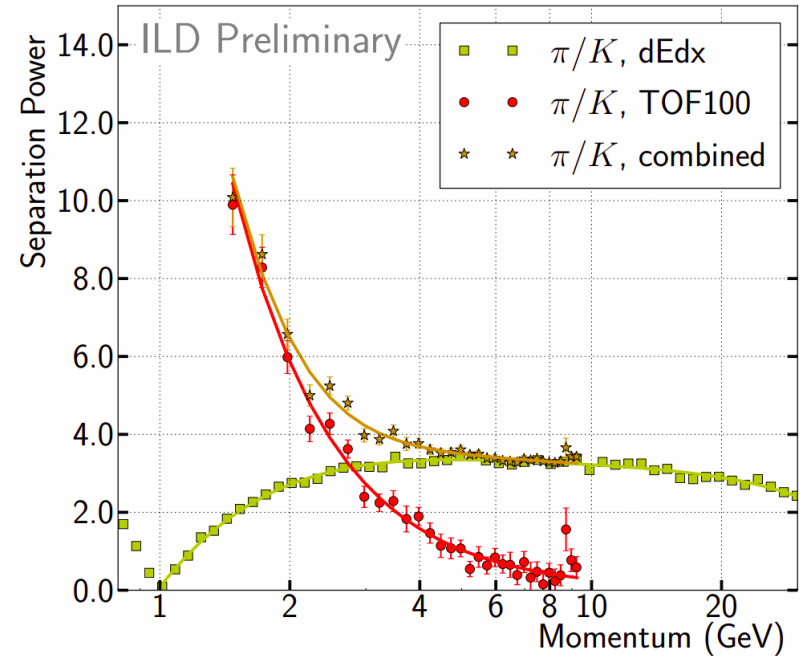
+ the usual b/c discriminants (vertex mass, impact parameter for all tracks, etc.)

Remember to normalize the discriminants to make them boost invariant (as much as possible)

Taken from Slide 5 of Tomohiko Tanabe's 2020/11/24 presentation.

Flavour tagging requirements

- Good impact parameter resolution, secondary vertexing – pertinent to b/c -tagging
- For strange versus up/down (“light”) quark tagging, there’s a need for **kaon tagging**
 - TPC provides dE/dx , Si detectors on either side of TPC provide time-of-flight (TOF) measurement
 - TOF works best at low p (< 10 GeV), expect dE/dx to work better for kaon tagging (where $p > 10$ GeV)
- ILD already provides BDT scores for b/c -taggers and an other (“o”) tagger per jet

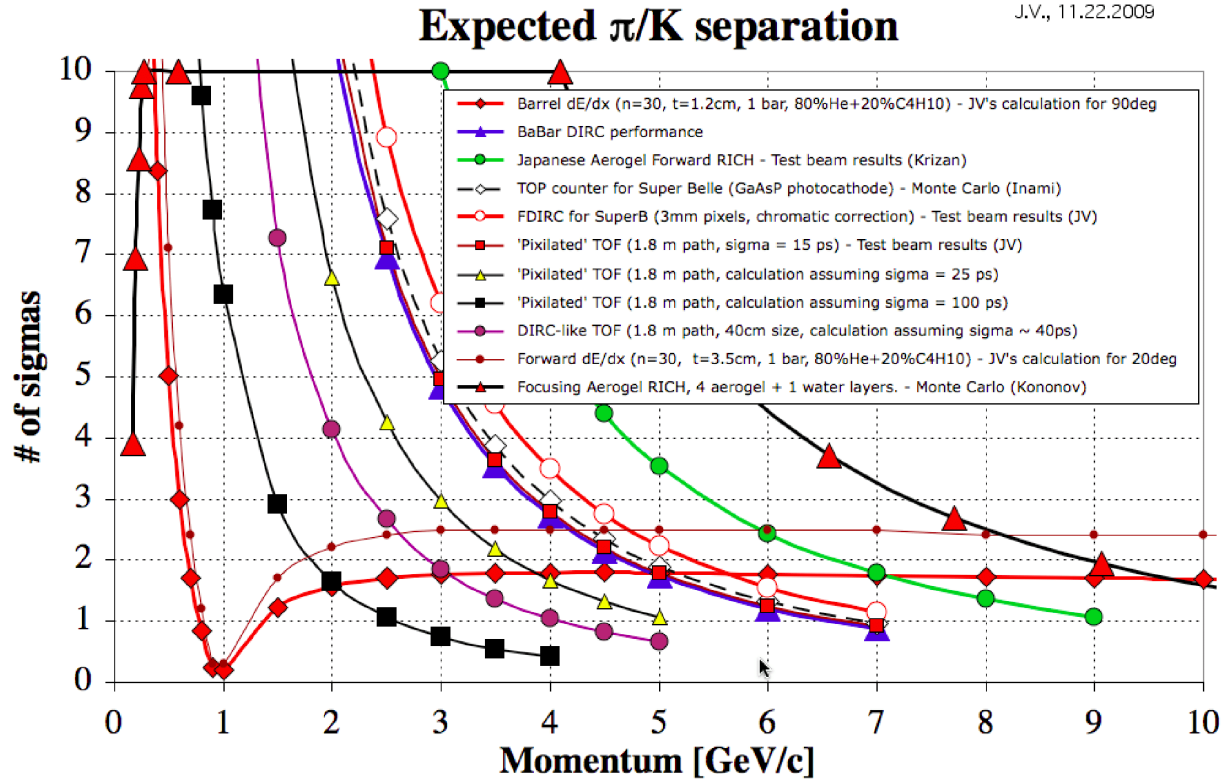


ILD separation power for pions and kaons using dE/dx and TOF (100 ps resolution). Taken from Fig. 3 of [1912.04601](#).

π/K separation for different detectors

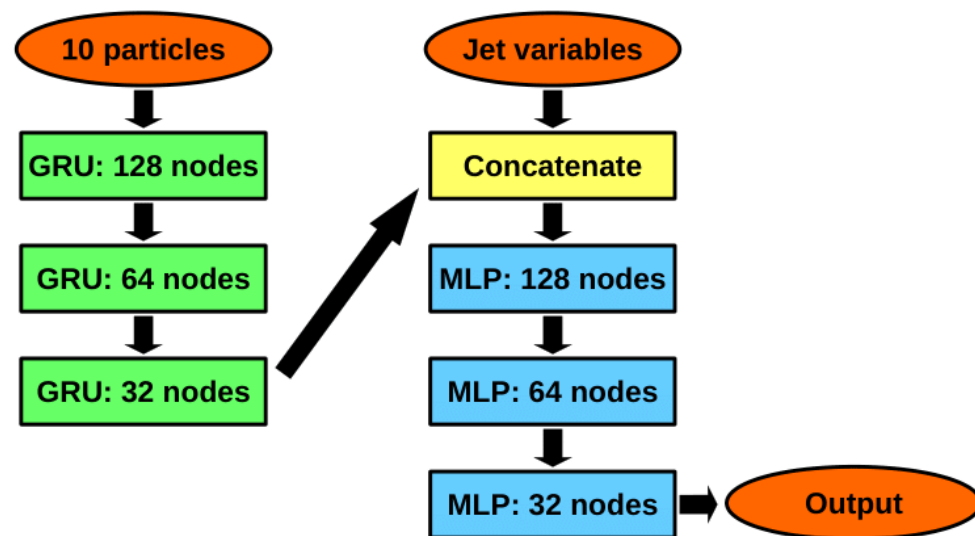
J.V., 11.22.2009

Plot from Jerry Va'vra (SLAC).



Tagger architecture: pictorially

- Neural network architecture:
 - *Multiclassifer* (5 output classes: gluon, light, strange, charm, or bottom)
 - 3 layer recurrent neural network (using Gated Recurrent Units or GRUs) for particle-level inputs
 - Concatenated with jet-level inputs and fed into a 3 layer MultiLayer Perceptron (MLP)
 - Applied to strange tagging performance at **hadron** colliders
 - “Maximum performance of strange-jet tagging at hadron colliders” ([2011.10736](#))



e⁺e⁻ cross sections

Table 2, taken from page 62 of Tomohisa Ogawa's thesis.

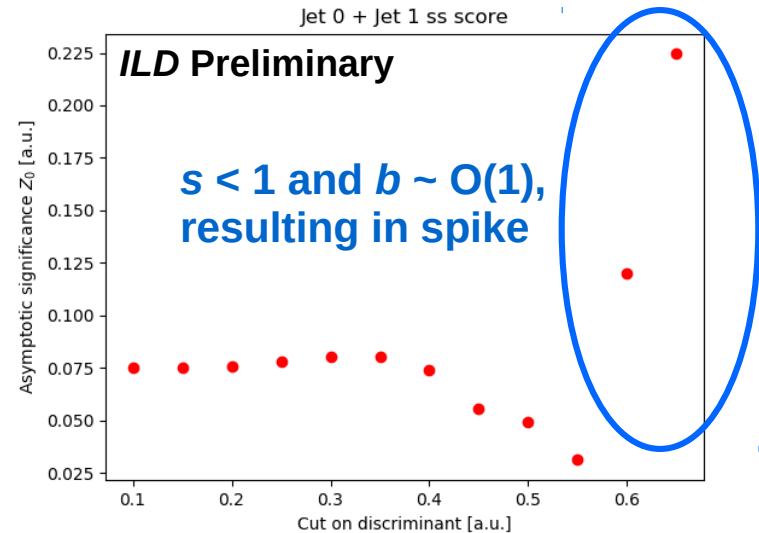
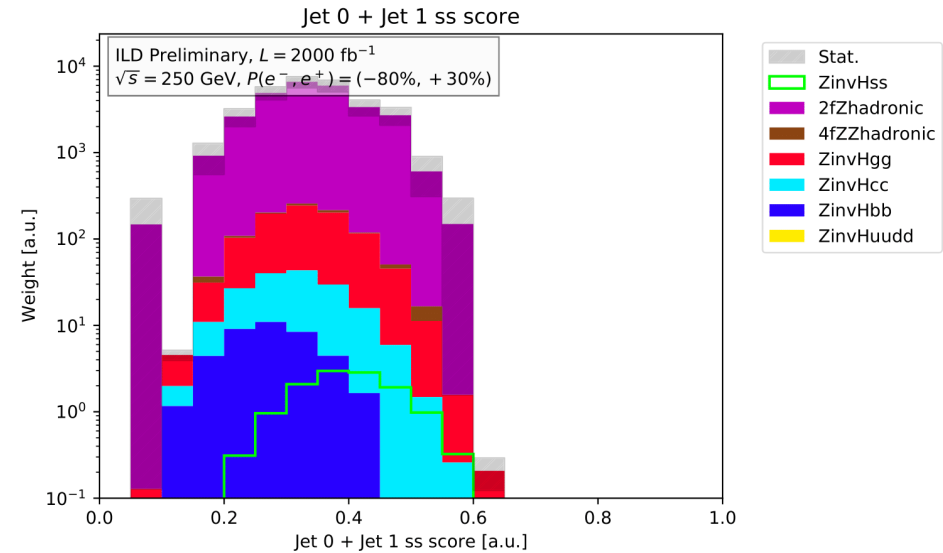
Table 2: Cross-sections and number of generated MC samples on the Higgs production processes and the major SM background processes for both $\sqrt{s} = 250$ and 500 GeV. The cross-sections given in the table are set to be each operation beam polarization states: $P(e^-, e^+) = (-80\%, +30\%)$ and $P(e^-, e^+) = (+80\%, -30\%)$, whereas the number of MC samples are given with fully beam polarization states: $P(e^-, e^+) = P_{e^-}^L P_{e^+}^R = (-100\%, +100\%)$. The $eeH(s)$ and $eeH(t)$ denote the s -channel ZH process and the t -channel ZZ -fusion processes. $2f_l$ and $2f_h$ in the table indicate that the final state has a lepton pair such as charged leptons or neutrinos, and a quark pair like $u\bar{u}$, $d\bar{d}$ except $t\bar{t}$. $4f_l$ and $4f_h$ are the same indication with $2f_l$ or $2f_h$, that means a final state has two lepton pairs or two quark pairs. $4f_sl$ shows that a final state has a lepton pair and a quark pair. At $\sqrt{s} = 500$ GeV $6f$ is included in the SM backgrounds, where possible diagrams of 6 fermions in a final state are considered such as $t\bar{t}$ and a fermion pair with two W bosons and two fermion pairs with the Z boson.

$\sqrt{s}=250$ GeV operation polarization	Cross-section (fb)		fully polarization			
	$P(e^-, e^+) = (-80\%, +30\%)$	$P(e^-, e^+) = (+80\%, -30\%)$	$P_{e^-}^L P_{e^+}^R$	$P_{e^-}^R P_{e^+}^L$	$P_{e^-}^L P_{e^+}^L$	$P_{e^-}^R P_{e^+}^R$
$eeH(s)$	10.7	7.14	$4.00 \cdot 10^4$	$1.00 \cdot 10^4$	0	0
$eeH(t)$	0.71	0.52	$1.00 \cdot 10^4$	$1.00 \cdot 10^4$	3992	3992
$\mu\mu H$	10.4	7.03	$4.00 \cdot 10^4$	$1.00 \cdot 10^4$	0	0
qqH	210.2	141.9	$5.45 \cdot 10^5$	$2.94 \cdot 10^5$	0	0
$\nu\nu H (s)$	61.6	41.6	$12.8 \cdot 10^4$	$6.50 \cdot 10^4$	0	0
$\nu\nu H (t)$	15.4	0.93	$12.8 \cdot 10^4$	$6.50 \cdot 10^4$	0	0
$2f_l$	$3.82 \cdot 10^4$	$3.49 \cdot 10^4$	$2.63 \cdot 10^6$	$2.13 \cdot 10^6$	$5.03 \cdot 10^5$	$5.03 \cdot 10^5$
$2f_h$	$7.80 \cdot 10^4$	$4.62 \cdot 10^4$	$1.75 \cdot 10^6$	$1.43 \cdot 10^6$	0	0
$4f_l$	$6.03 \cdot 10^3$	$1.47 \cdot 10^3$	$2.25 \cdot 10^6$	$9.80 \cdot 10^4$	$2.73 \cdot 10^5$	$2.73 \cdot 10^5$
$4f_sl$	$1.84 \cdot 10^4$	$2.06 \cdot 10^3$	$4.04 \cdot 10^6$	$3.56 \cdot 10^5$	$9.78 \cdot 10^4$	$9.78 \cdot 10^4$
$4f_h$	$1.68 \cdot 10^4$	$1.57 \cdot 10^3$	$2.38 \cdot 10^6$	$2.42 \cdot 10^5$	0	0

Signal discriminant

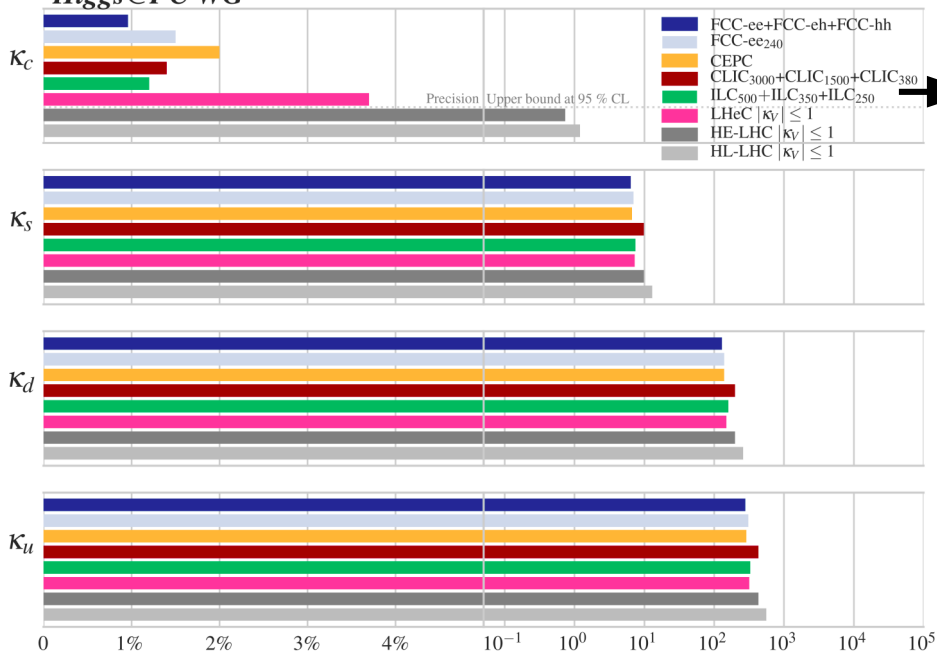
- Use (0.5x) sum of strange jet scores for the leading and subleading jets as the discriminant
 - Tested different cuts, calculating the asymptotic significance (neglecting MC stats) as:

$$Z_0 = \sqrt{2 * ((s + b) * \ln(1 + s / b) - s)}$$
- Best cut seems to be around **>0.3**
 - Corresponds to $Z_0 \sim 0.1\sigma$



Expected limits from future colliders

Higgs@FC WG

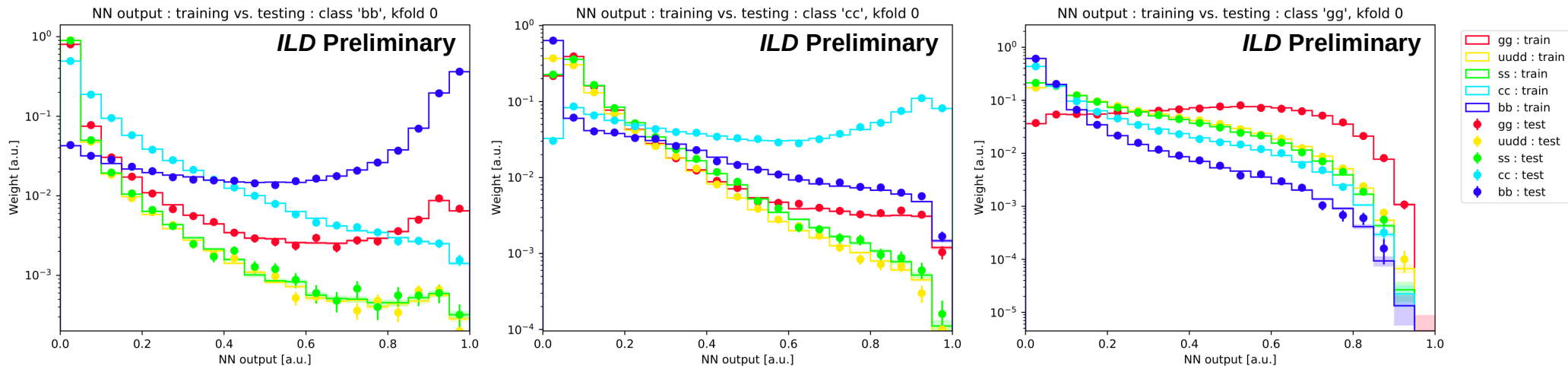


	T ₀	+5	+10	+15	+20	...	+26
ILC	0.5/ab 250 GeV		1.5/ab 250 GeV	1.0/ab 500 GeV	0.2/ab 2m _{top}	3/ab 500 GeV	
CEPC	5.6/ab 240 GeV		16/ab M _Z	2.6/ab 2M _W	SppC =>		
CLIC	1.0/ab 380 GeV			2.5/ab 1.5 TeV		5.0/ab => until +28 3.0 TeV	
FCC	150/ab ee, M _Z	10/ab ee, 2M _W	5/ab ee, 240 GeV	1.7/ab ee, 2m _{top}		hh,eh =>	
LHeC	0.06/ab		0.2/ab	0.72/ab			
HE-LHC	10/ab per experiment in 20y						
FCC eh/hh	20/ab per experiment in 25y						

Limits on couplings of Higgs to charm, strange, and light quarks and timeline for different future collider running scenarios. Taken from Figs. 1 and 12 of [1905.03764](https://arxiv.org/abs/1905.03764).

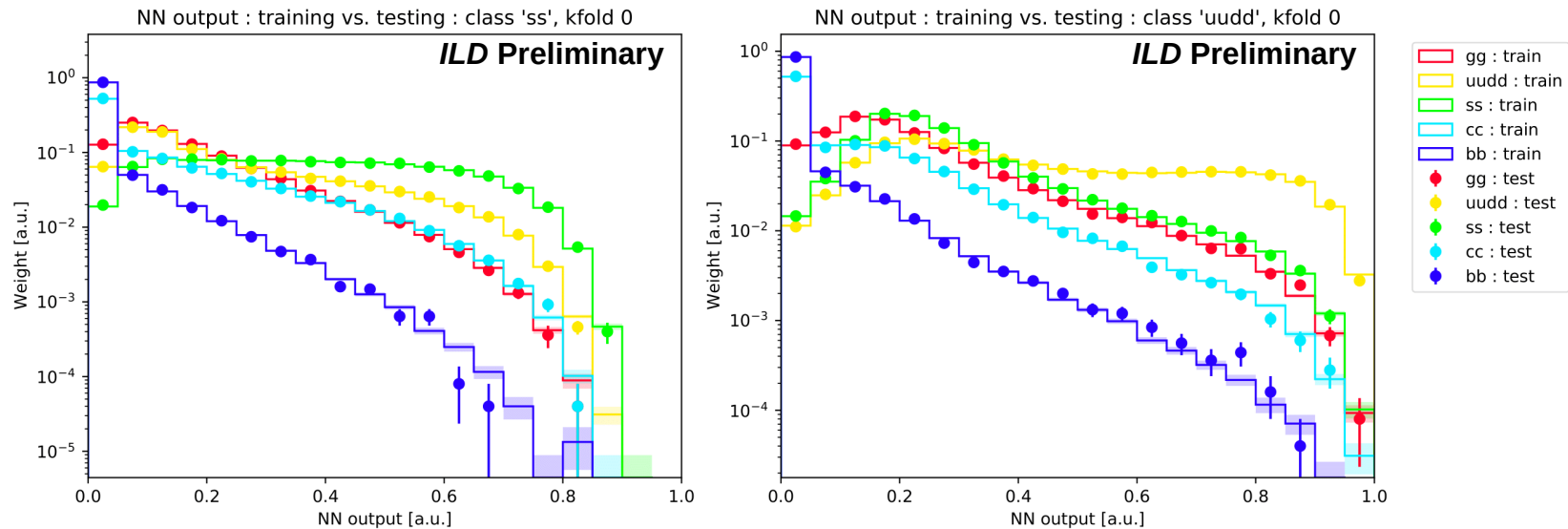
Train-test agreement in neural network output nodes

Performance: b , c , and g jets



- Network likely returning b/c -tagger scores – should do just as well or better than input BDT scores
- Good discrimination of gluon jets

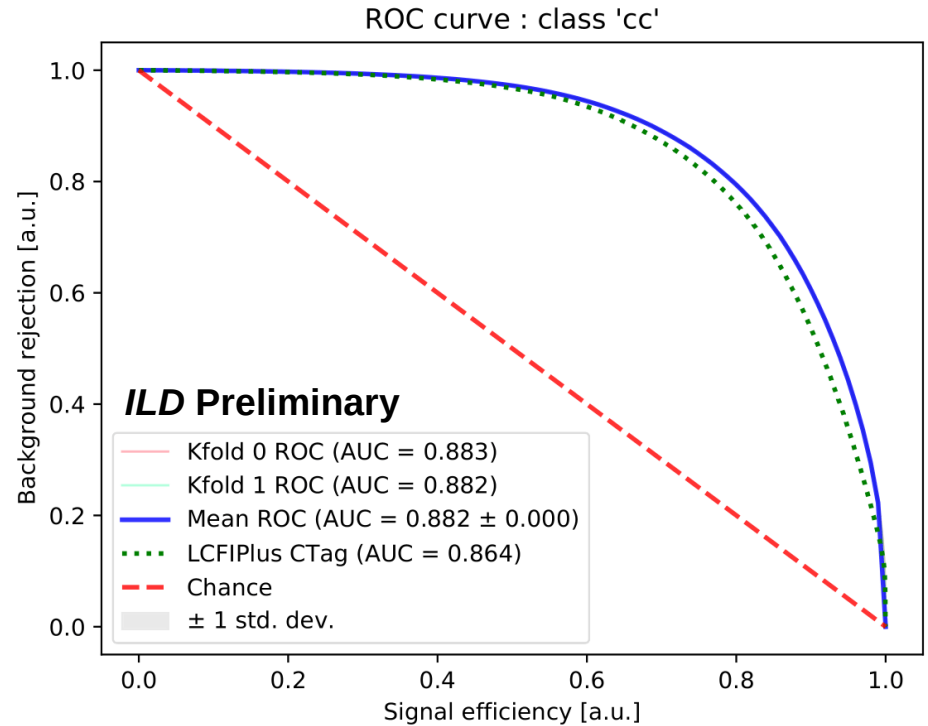
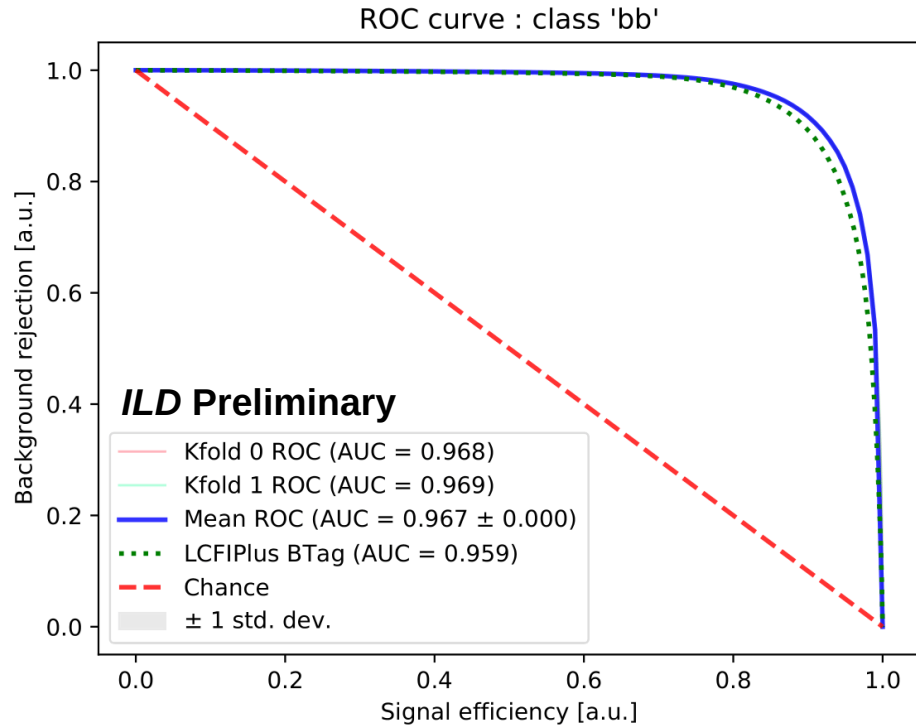
Performance: s and u/d jets



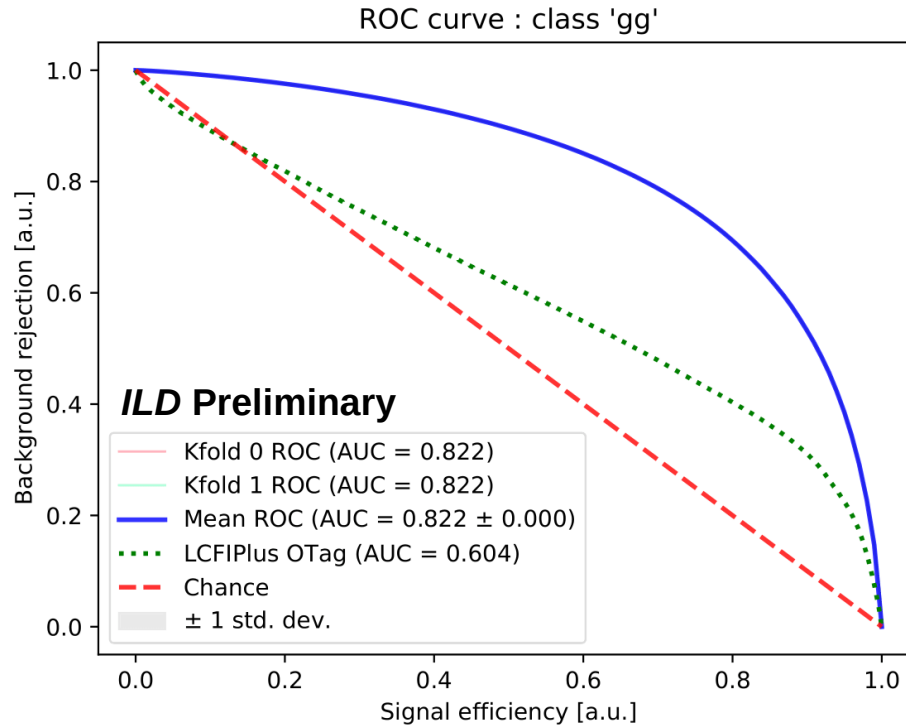
- Separation of s and u/d is **possible** with using truth likelihoods
- At 50% strange tagging efficiency, we have **90%** background rejection over **70%** for LCFIPlus Otag (see [ROC curves](#))

ROC curves for neural network output nodes

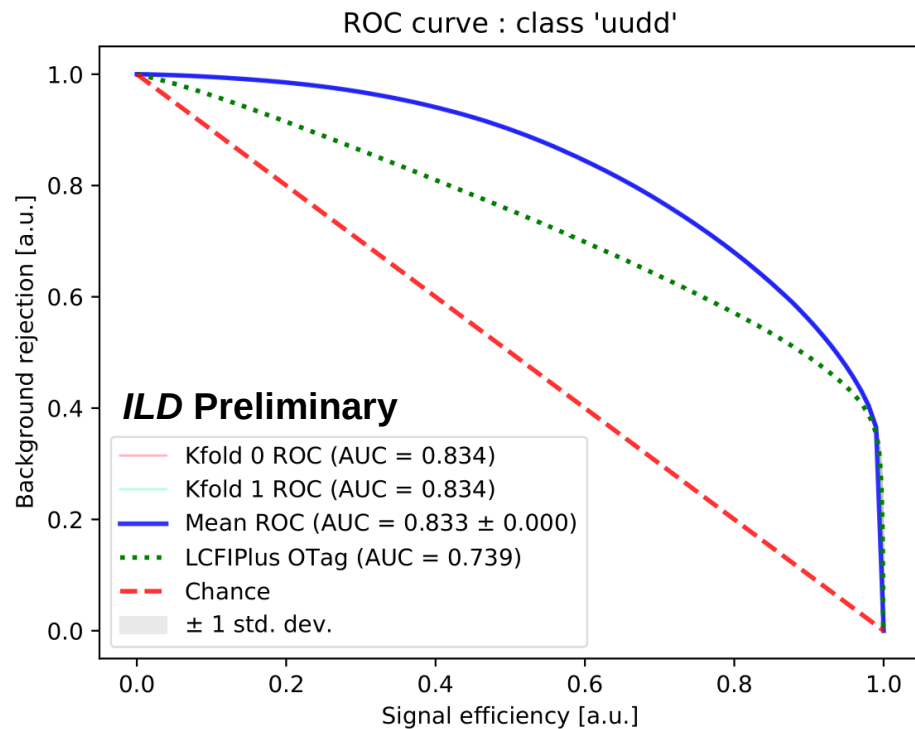
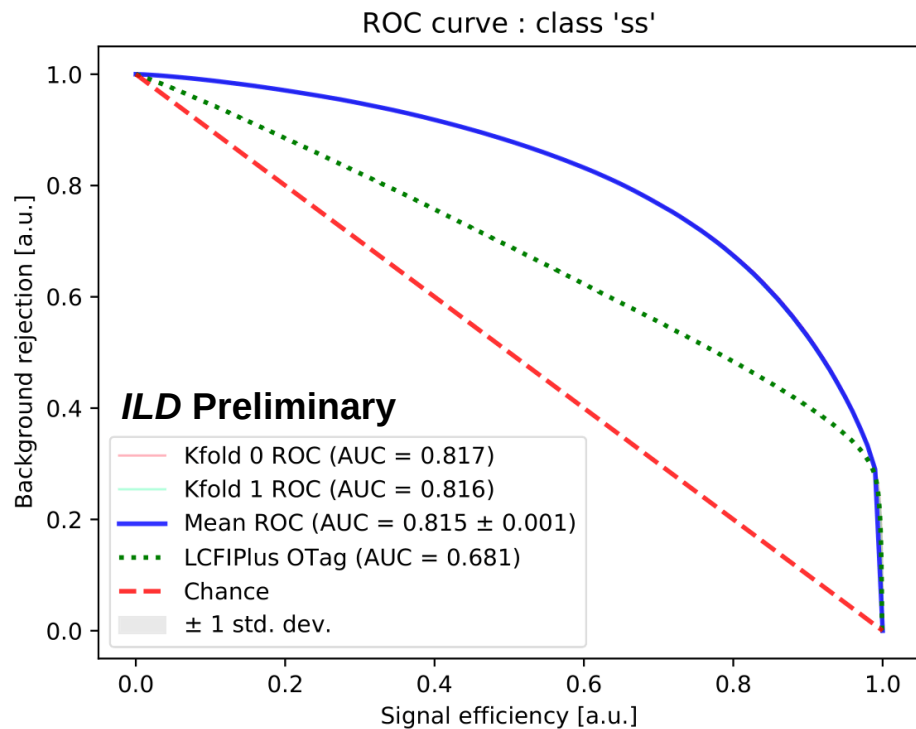
ROC curves: *b* and *c* jets



ROC curves: g jets

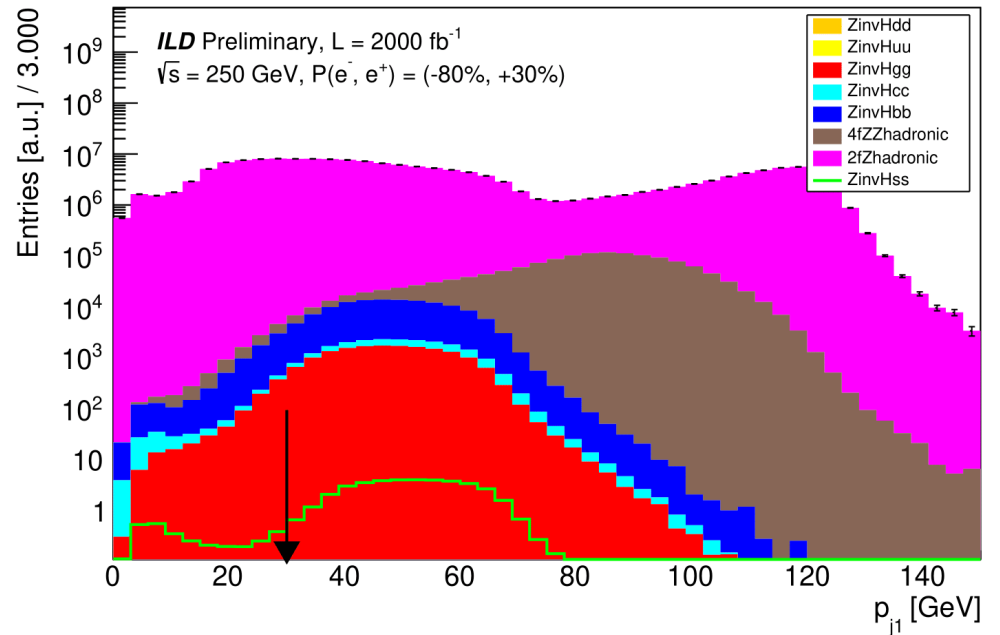
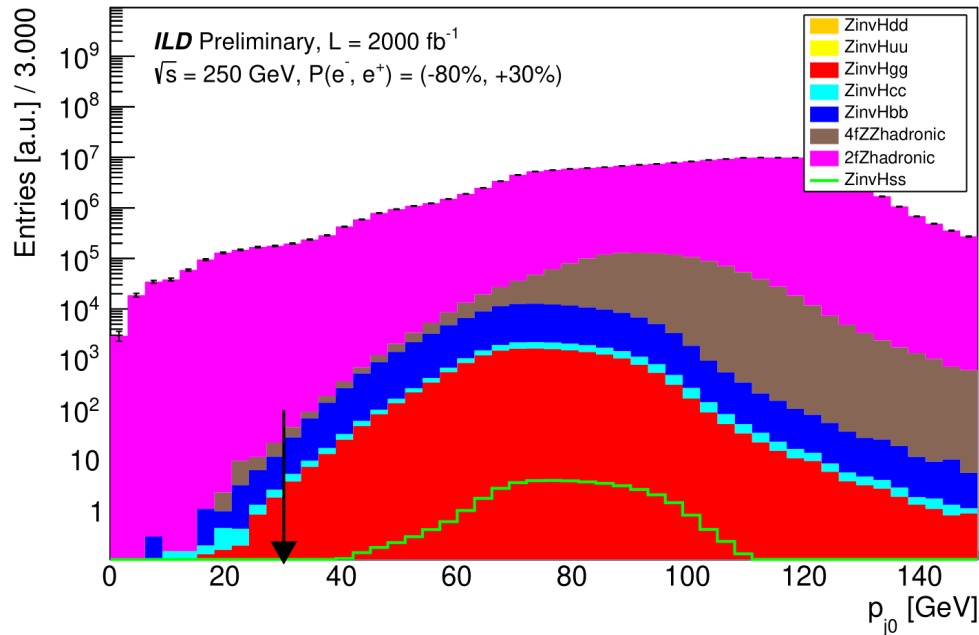


ROC curves: s and $u\bar{d}$ jets



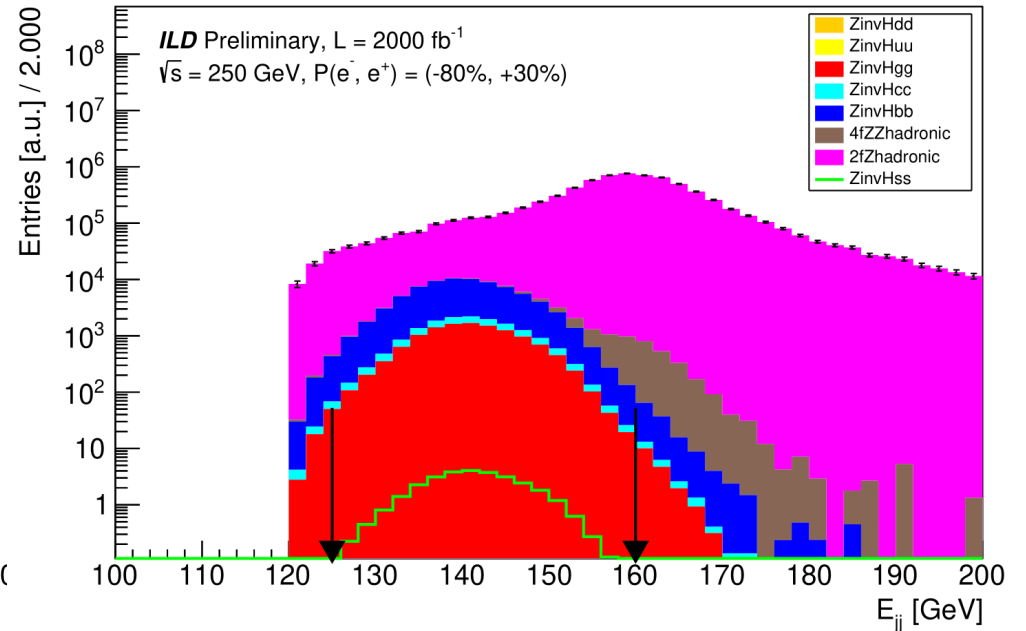
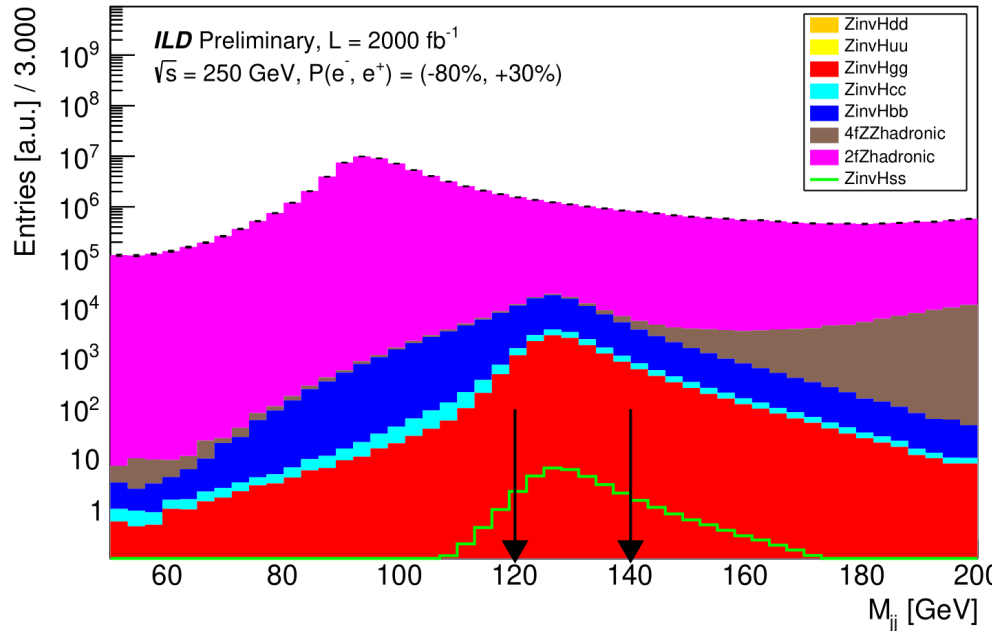
Histograms at each cut in $H \rightarrow ss$ analysis

Histograms: p_{j0} and p_{j1}



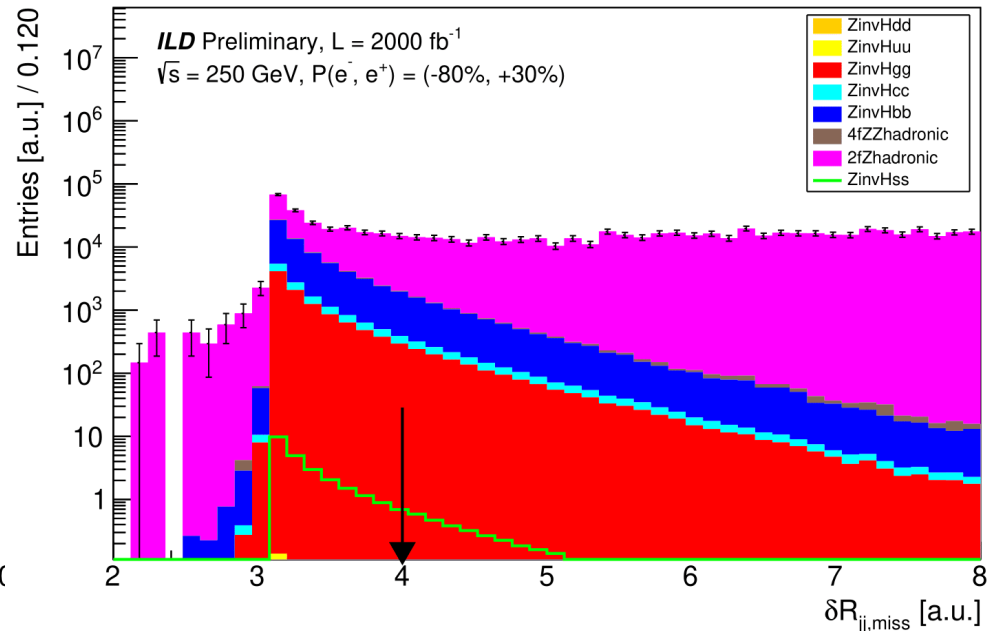
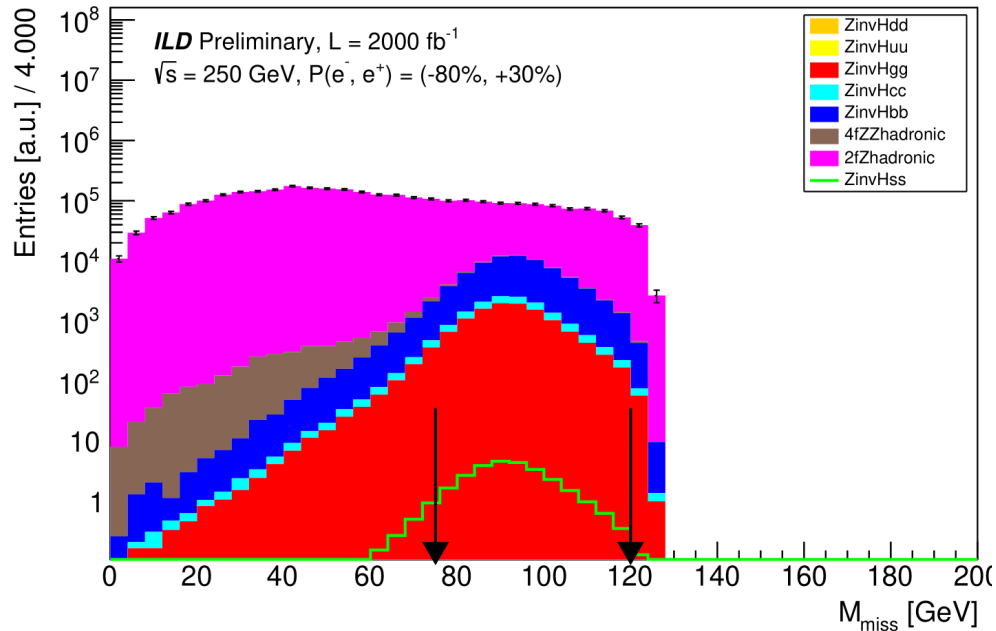
****Unstacked green line is signal****

Histograms: M_{jj} and E_{jj}



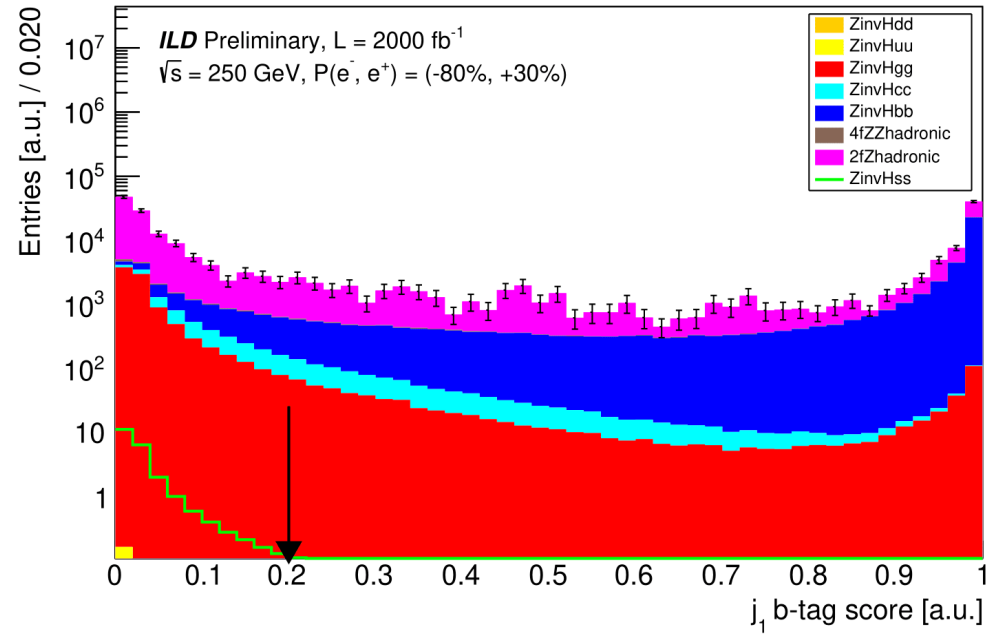
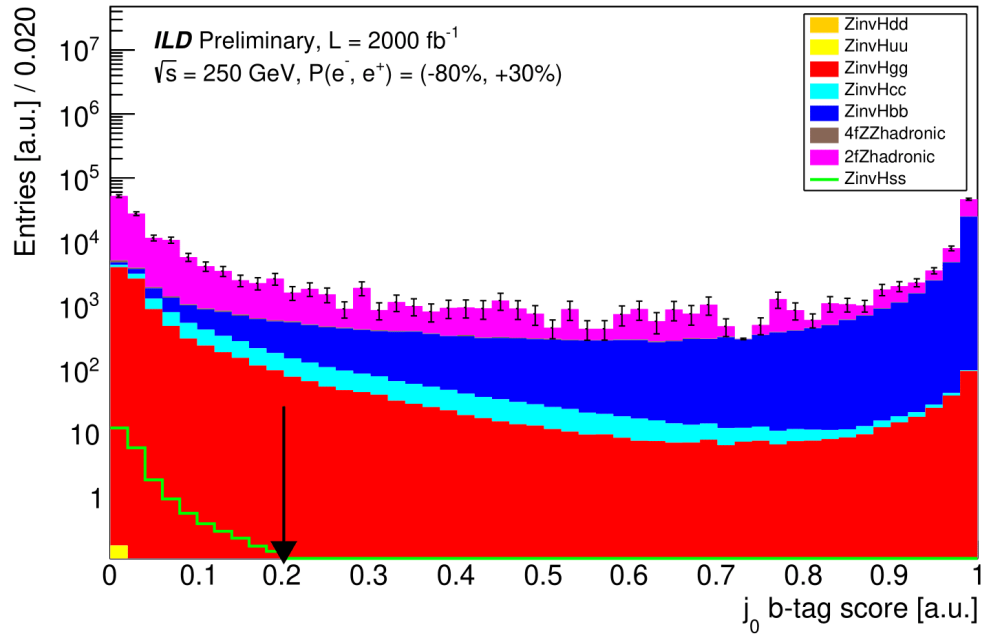
****Unstacked green line is signal****

Histograms: M_{miss} and $\Delta R_{jj,\text{miss}}$



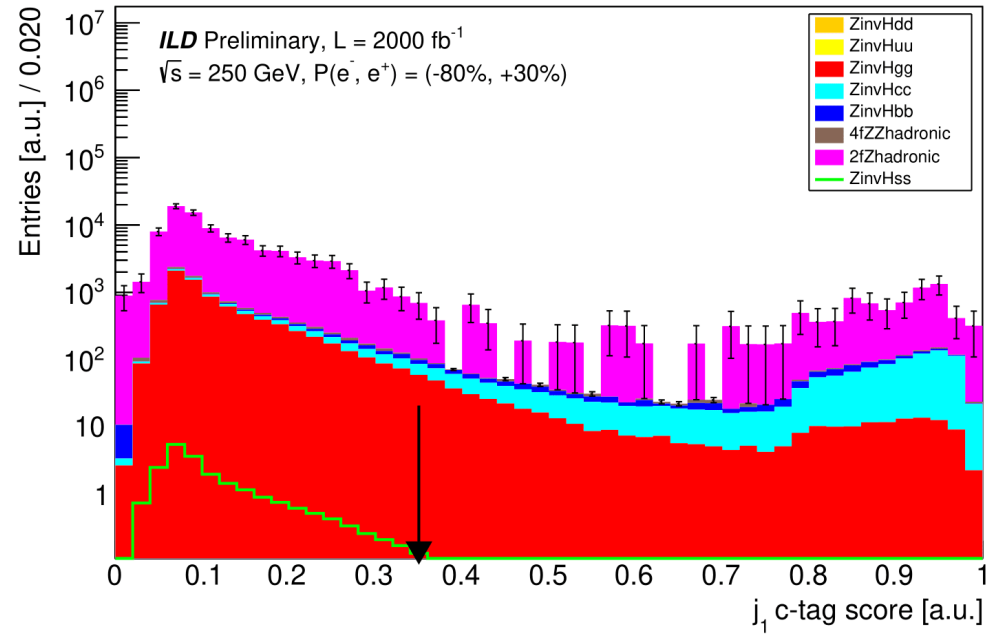
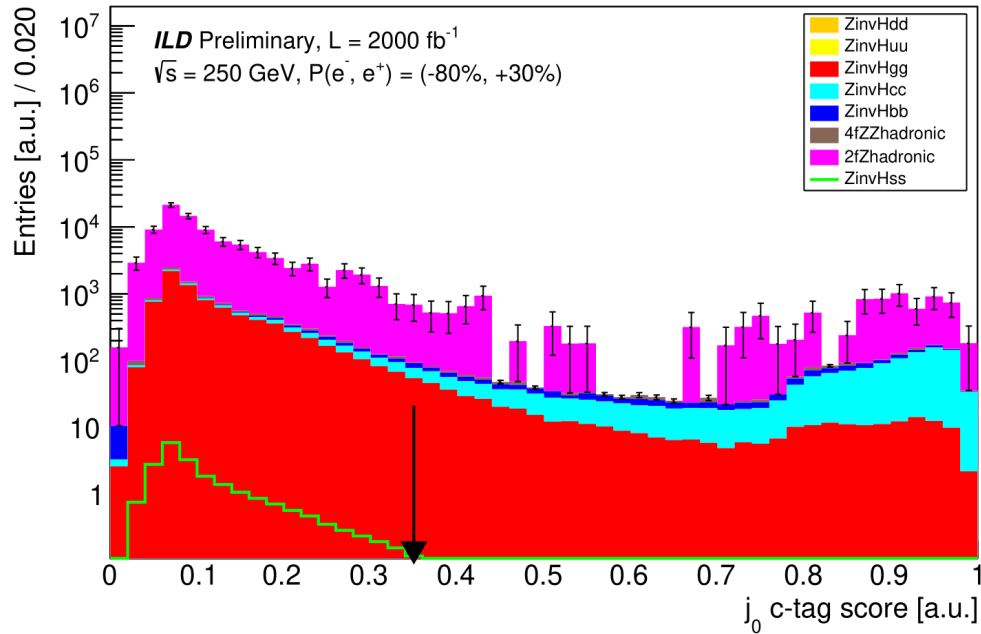
****Unstacked green line is signal****

Histograms: b -tagger scores



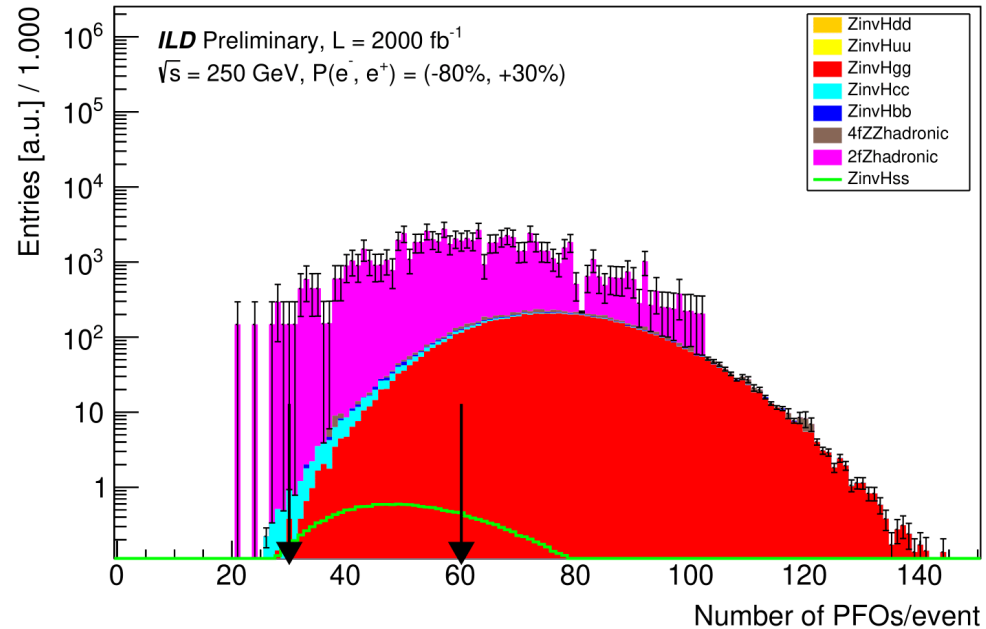
****Unstacked green line is signal****

Histograms: c-tagger scores



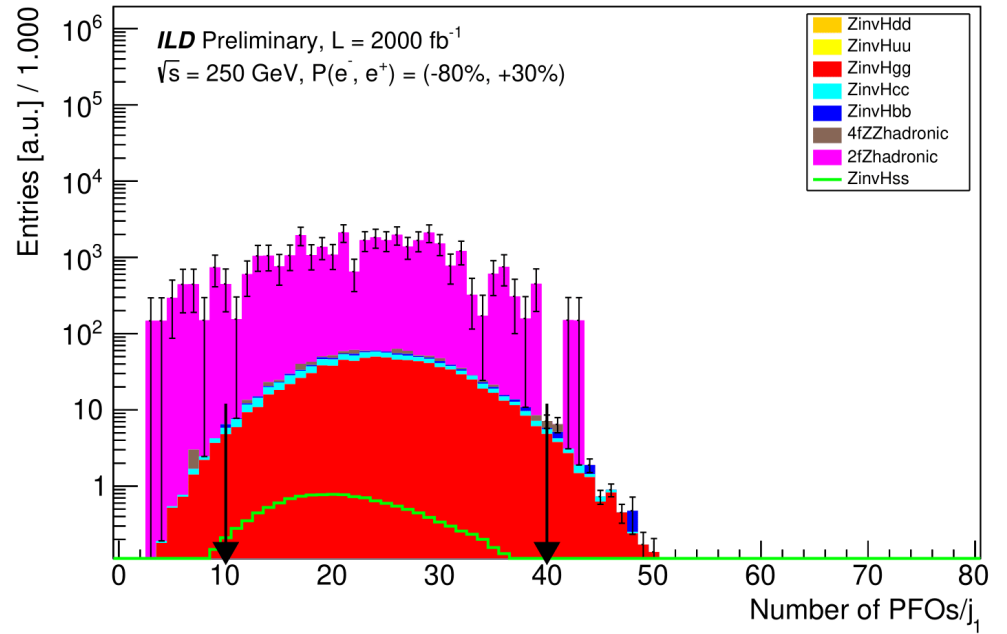
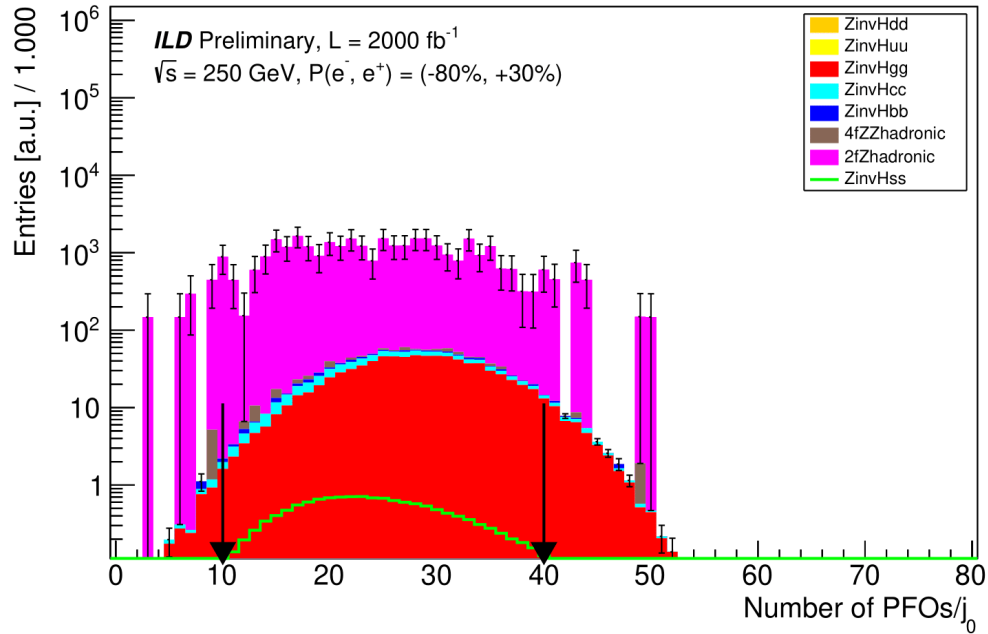
****Unstacked green line is signal****

Histograms: $N_{\text{PFOS}}/\text{event}$



****Unstacked green line is signal****

Histograms: $N_{\text{PFOs}}/\text{jet}$



****Unstacked green line is signal****