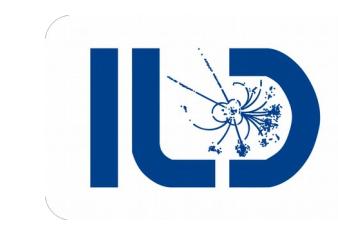


# CPIX: A Comprehensive Particle Identification Framework for Future $e^+e^-$ Colliders

Uli Einhaus  
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21.03.2023 Dresden  
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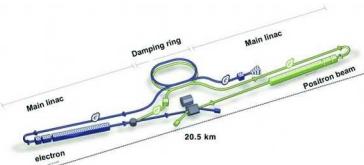
**HELMHOLTZ**

**CLUSTER OF EXCELLENCE**  
QUANTUM UNIVERSE



# The Landscape of Proposed Next-Gen Colliders / Future Higgs Factories

**ILC**



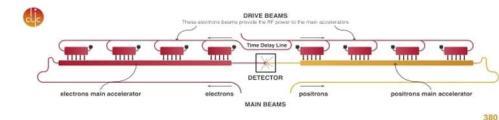
**CEPC**



**FCC-ee**

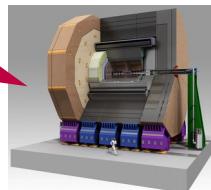


**CLIC**

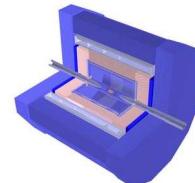


This talk uses ILD

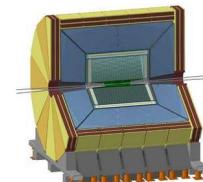
**ILD**



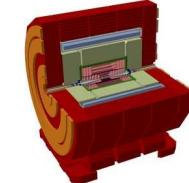
**CEPC Baseline**



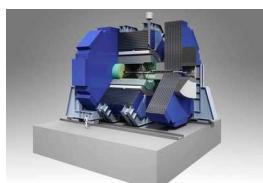
**IDEA**



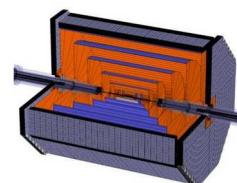
**CLICdp**



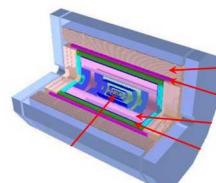
**SiD**



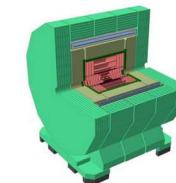
**FST**



**CEPC 4th concept**



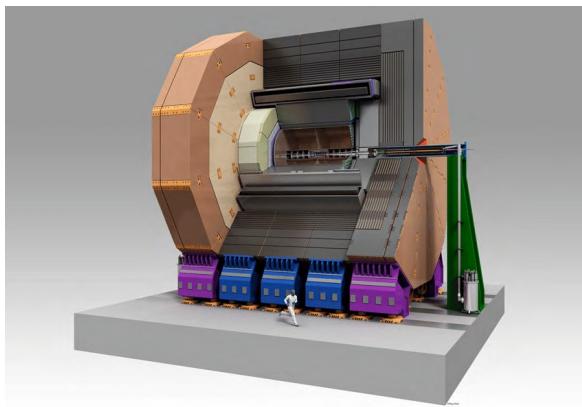
**CLD**



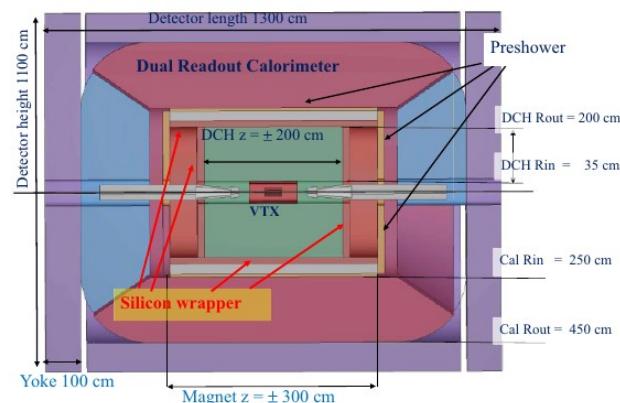
- Many proposals under consideration – common tools desired, in particular software!  
→ key4HEP / EDM4HEP

# PID at Future Higgs Factories

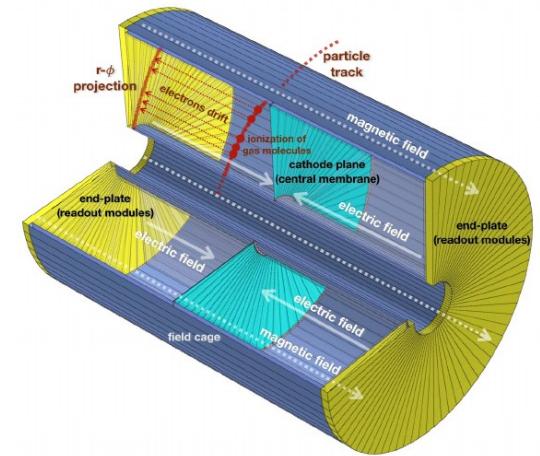
- Increasing understanding that particle identification (PID), in particular charged hadron PID, is a very valuable observable at a Future Higgs Factories
- Recent studies focus on 90-250 GeV and precision flavour physics instead of direct (BSM) detection at TeV range → PID is more effective and more relevant there
- This work: new software framework for comprehensive PID



ILD



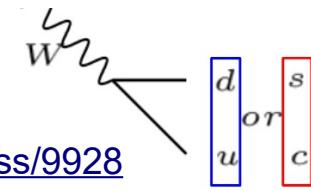
IDEA



CEPC-det

# Examples for PID Applications at ILD

- Z and W hadronic decay branching fractions via flavour tagging  
→ make connection between quark flavour and jet composition  
<https://ediss.sub.uni-hamburg.de/handle/ediss/9634> , <https://ediss.sub.uni-hamburg.de/handle/ediss/9928>



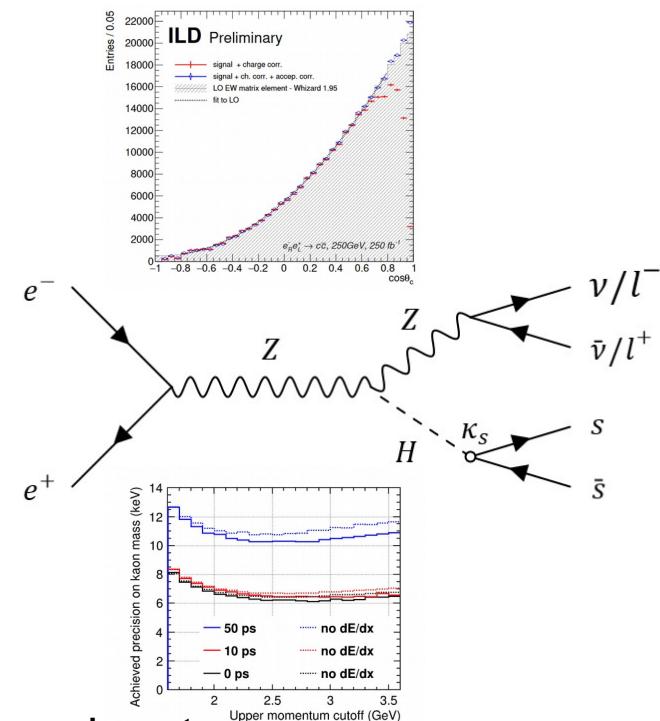
- Forward-backward asymmetry in  $e^+e^- \rightarrow q\bar{q}$   
→ study asymmetry in each flavour channel exclusively

overview: <https://tel.archives-ouvertes.fr/tel-01826535>  
 $e^+e^- \rightarrow tt, bb$ : <https://agenda.linearcollider.org/event/8147>  
 $e^+e^- \rightarrow bb/cc$ : <https://arxiv.org/abs/2002.05805> ,  
<https://agenda.linearcollider.org/event/9211/contributions/49358/>  
 $e^+e^- \rightarrow bb/cc, ss$ : <https://agenda.linearcollider.org/event/9440> ,  
<https://agenda.linearcollider.org/event/9285>

- $H \rightarrow s\bar{s}$  with s-tagging  
→ identify high-momentum kaons to tag  $s\bar{s}$  events  
<https://arxiv.org/abs/2203.07535>

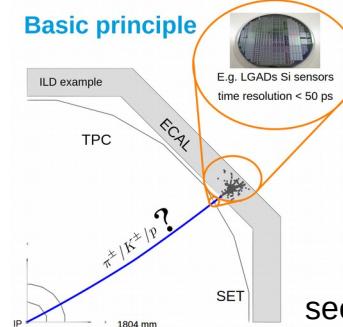
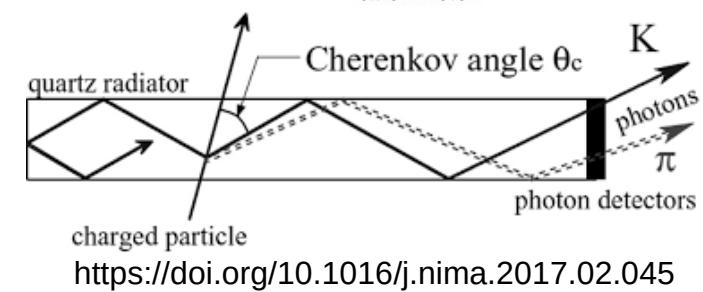
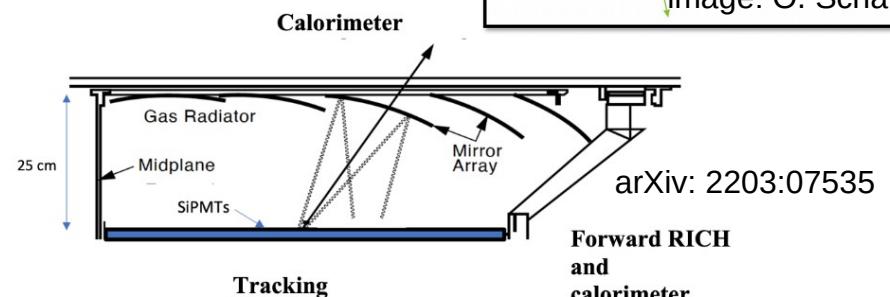
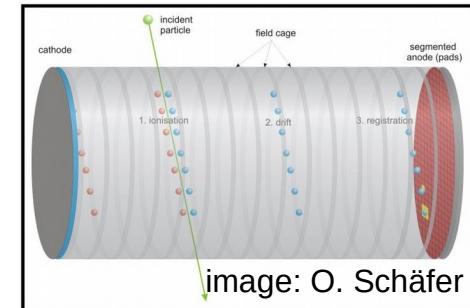
- Kaon mass with TOF  
<https://pos.sissa.it/380/115/>

- Track refit with correct particle mass for better momentum and vertex  
<https://agenda.linearcollider.org/event/8498/>



# PID Technology

- Gaseous trackers (Time Projection Chamber, Drift Chamber): specific energy loss  $dE/dx$ , via gas ionisation, up to 20 GeV
- Ring Imaging Cherenkov Detectors: Cherenkov angle, via imaging, 10 to 50 GeV
- Time of Propagation Counter: Cherenkov angle, via timing, up to 10 GeV
- Time of Flight: time, via Silicon timing, up to 5 GeV



see: T 48.5 B. Dudar

# Central Questions

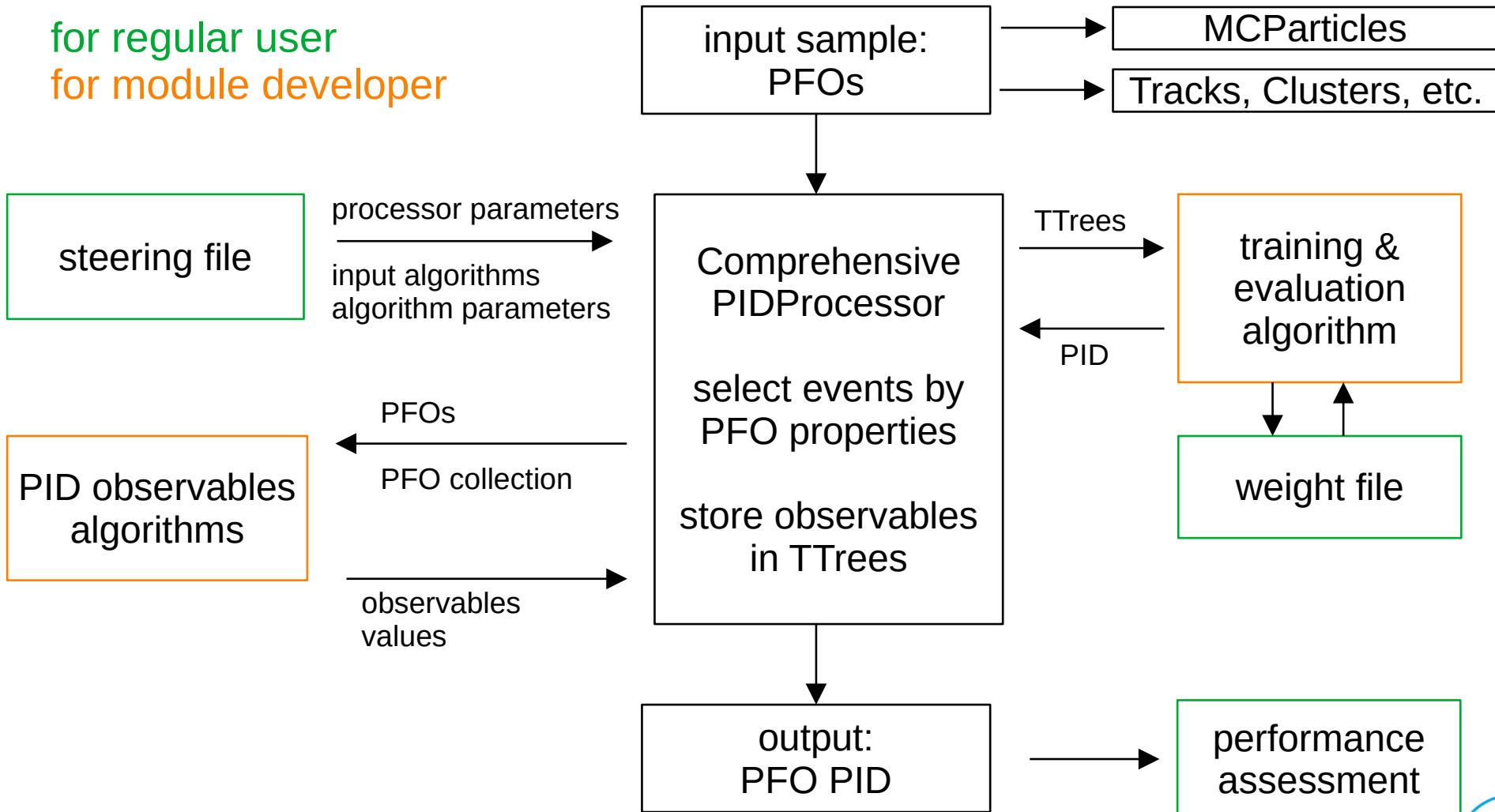
- How do we combine the different technologies best?
- How can we create a general assessment of PID performance valid for all of them?
- Can we use machine learning to extract the best performance from the PID observables?
- Optimise detectors and compare them
  - At what timing resolution starts TOF to be relevant for flavour tagging?
  - How does **my** physics result depend on the  $dE/dx$  resolution?
  - What if we add a RICH to SiD?

# New Framework: Comprehensive Particle Identification (CPID)

- Modularity as core philosophy:
  - observables algorithms
  - training methods / evaluation algorithms
- Core code takes care of book keeping
  - simple, well defined data structures for storage (TTree) and interfaces (std::vector)
- For now, being implemented in LCIO / Marlin in iLCSoft
  - immediately usable in Key4HEP via ‘Marlin wrapper’
  - target: implement in EDM4HEP, make available to whole future colliders community
- In ILD: goal to replace current, somewhat inflexible algorithm

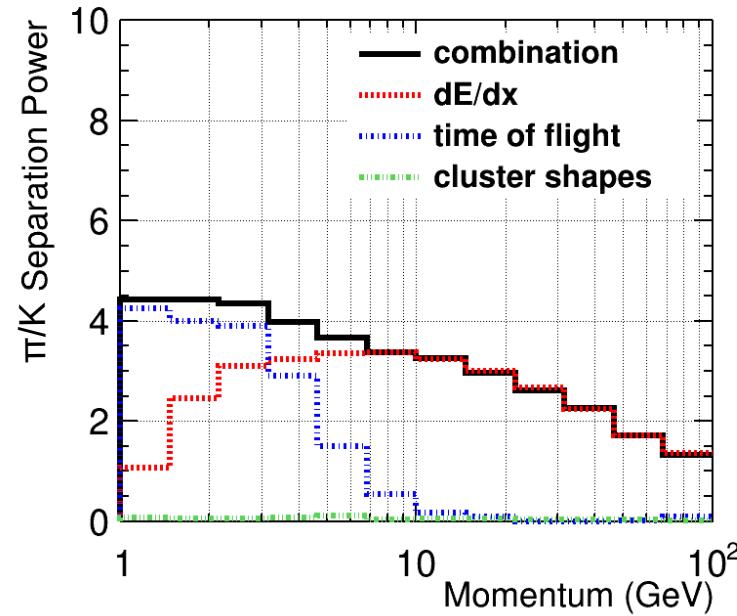
# General Structure

for regular user  
for module developer



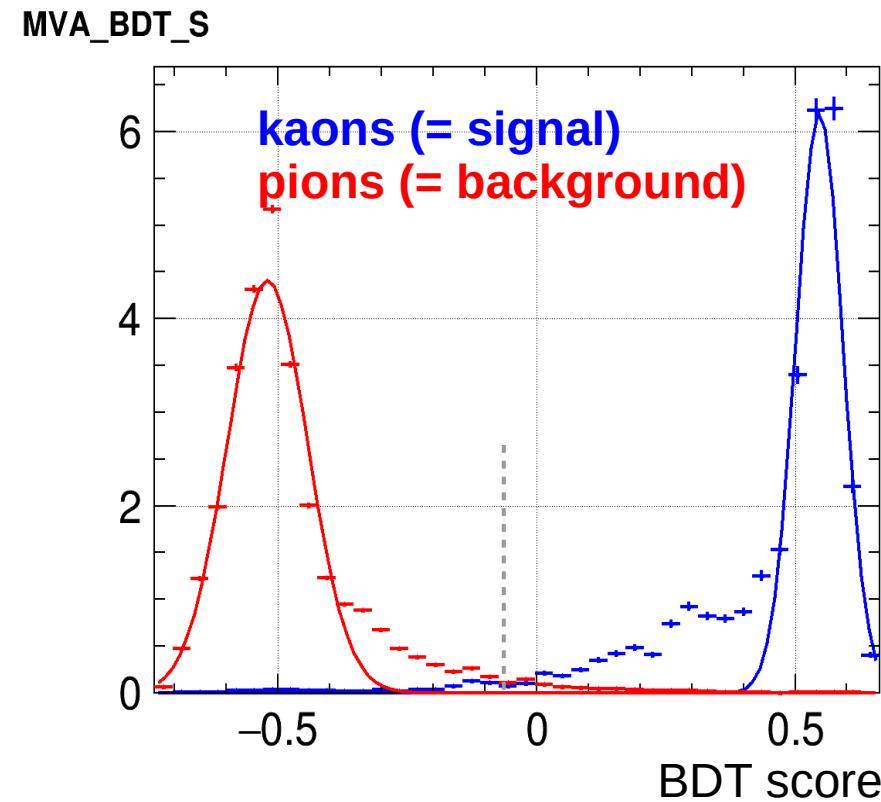
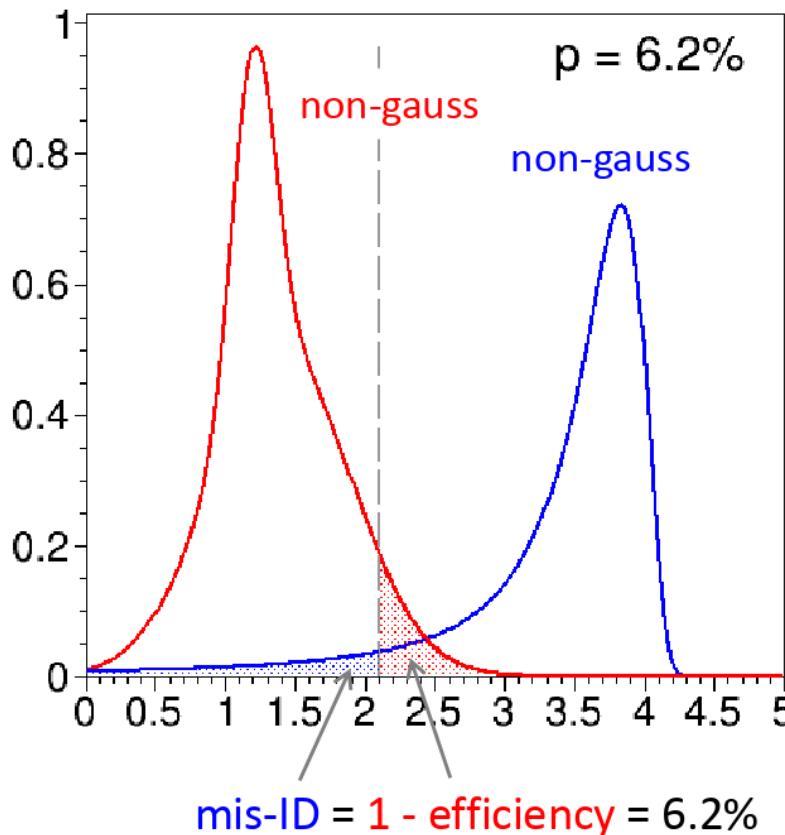
# Example 1: $\pi/K$ Separation with Combined Observables

- $dE/dx + \text{TOF}$
- Single particles ‘calibration’ events, flat in  $\log(p)$  and  $\cos(\theta)$
- BDT with sig = K, bkg =  $\pi$ ; train & eval per 12 mom bins and per used observable(s)  
→ How do we calculate a separation power from a BDT score?



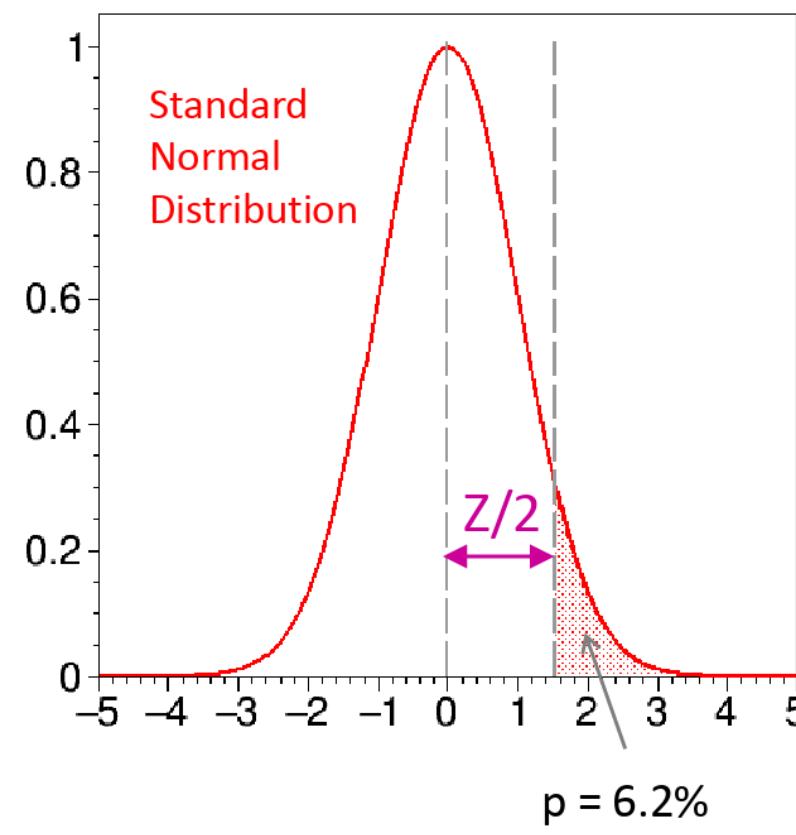
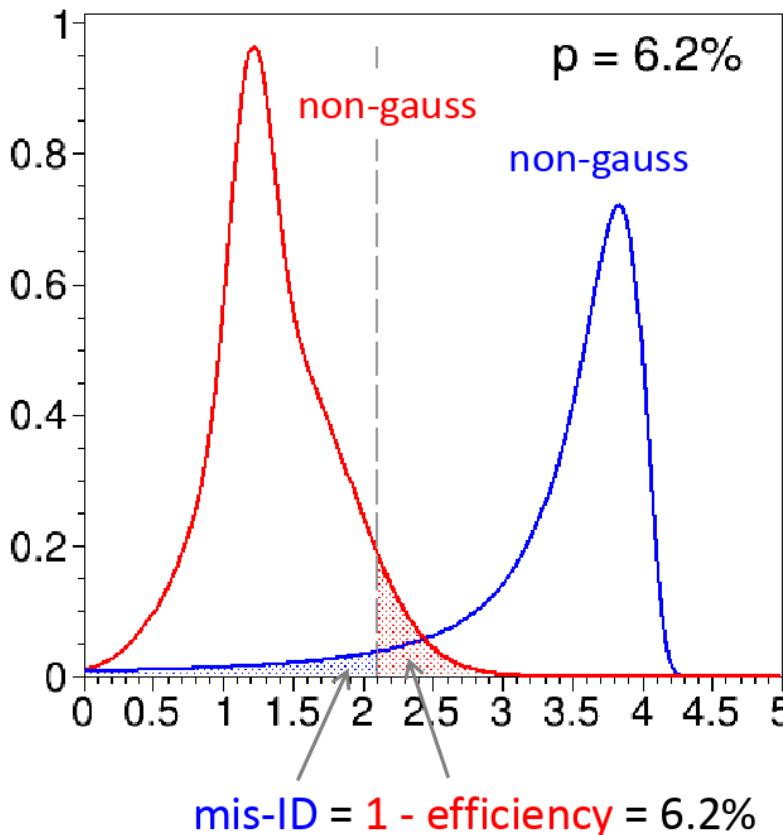
# p-value Assessment

- Find cut with  $\text{mis-ID} = 1 - \text{efficiency} = p\text{-value} \rightarrow \text{find Gaussian quantile}$   
→ compute  $Z = 2 \cdot \text{quantile of standard Gauss}$



# p-value Assessment

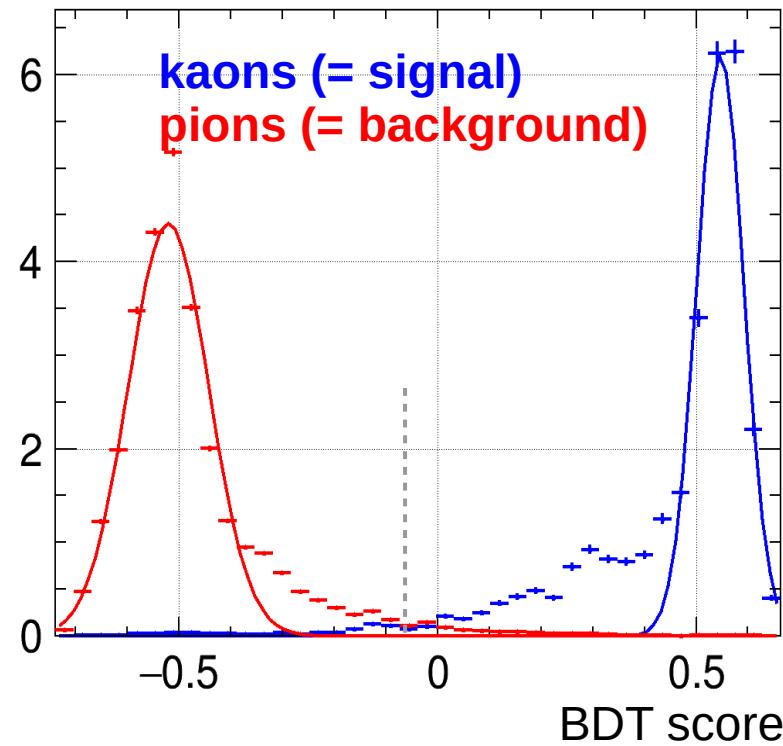
- Find cut with  $\text{mis-ID} = 1 - \text{efficiency} = p\text{-value} \rightarrow \text{find Gaussian quantile}$   
 $\rightarrow \text{compute } Z = 2 \cdot \text{quantile of standard Gauss}$



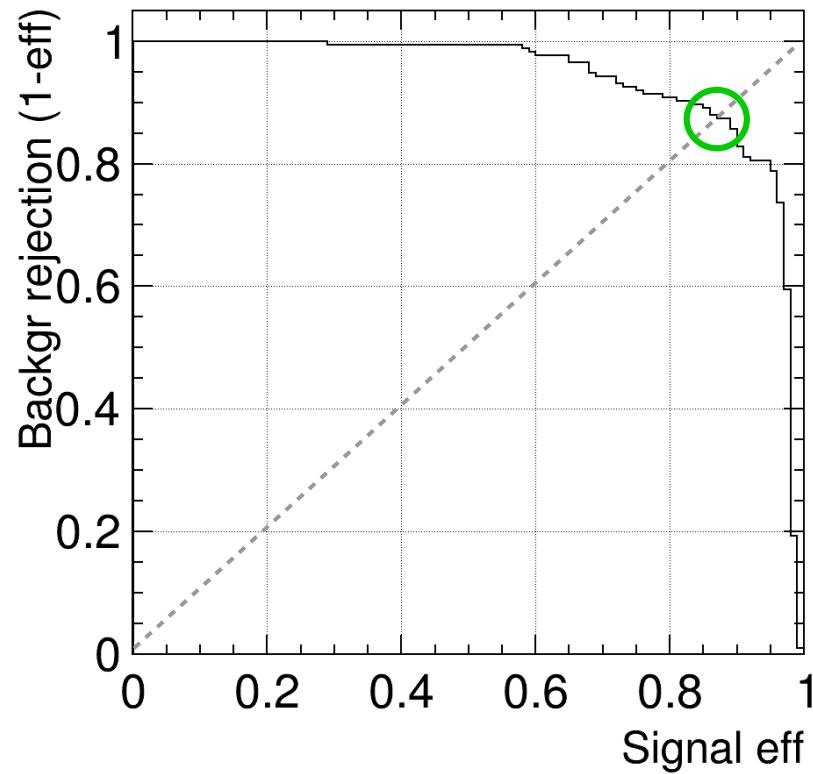
# p-value Assessment

- ‘Central tail split’ of BDT score is equivalent to crossing point of ROC curve with  $x=y$  line

MVA\_BDT\_S

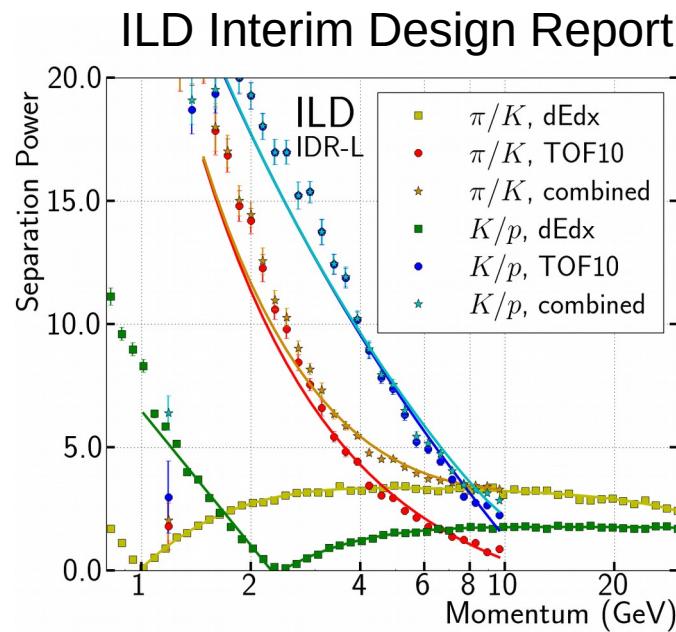


MVA\_BDT

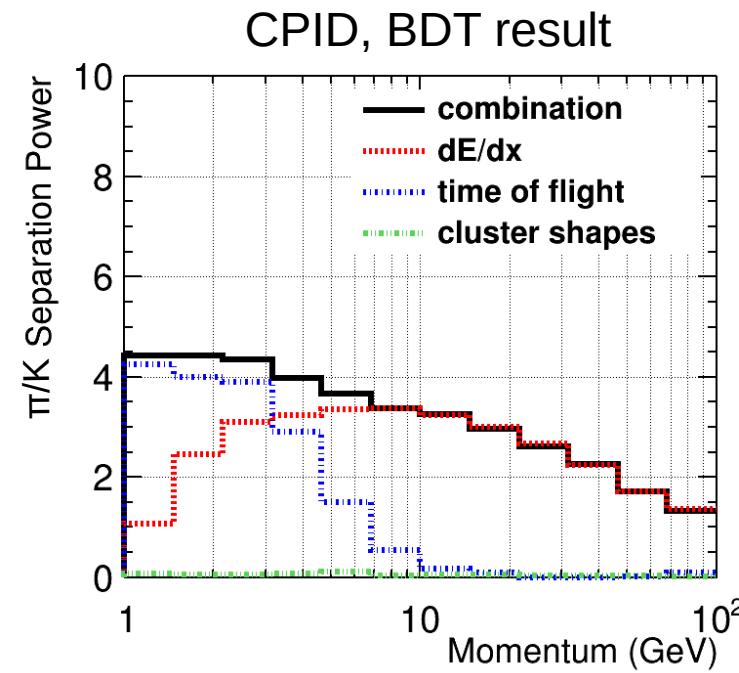


# Example 1: $\pi/K$ Separation with Combined Observables

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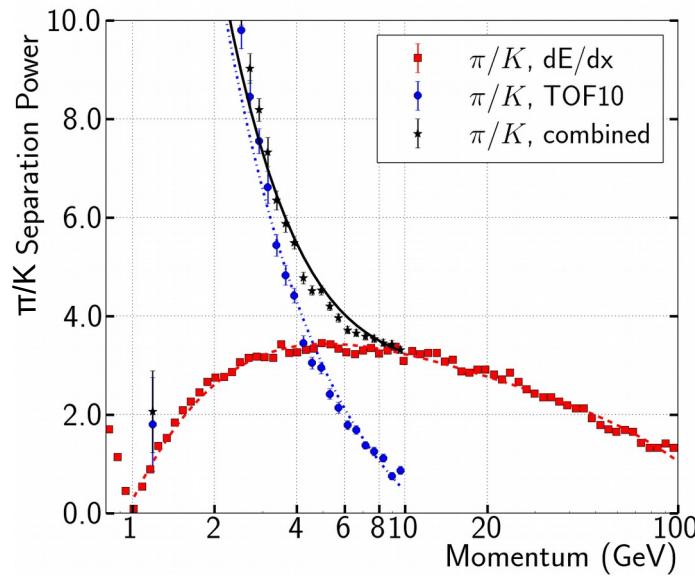
<https://arxiv.org/abs/2003.01116>



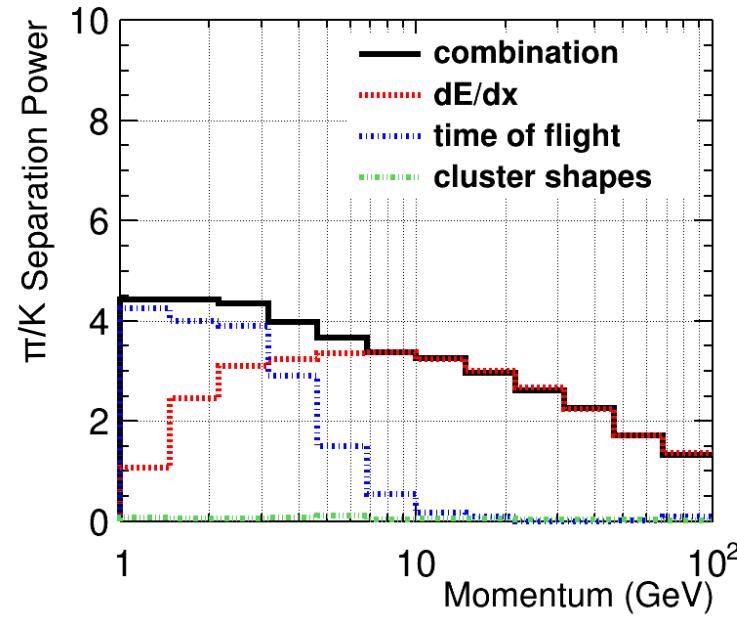
# Example 1: $\pi/K$ Separation with Combined Observables

- $dE/dx + \text{TOF}$
- Single particles ‘calibration’ events, flat in  $\log(p)$  and  $\cos(\theta)$
- BDT with sig = K, bkg =  $\pi$ ; train & eval per 12 mom bins and per used observable(s)
- $dE/dx$  very similar, TOF levels out due to badly reconstructed events ( $\rightarrow$  see next Talk!)

analogue to ILD Interim Design Report

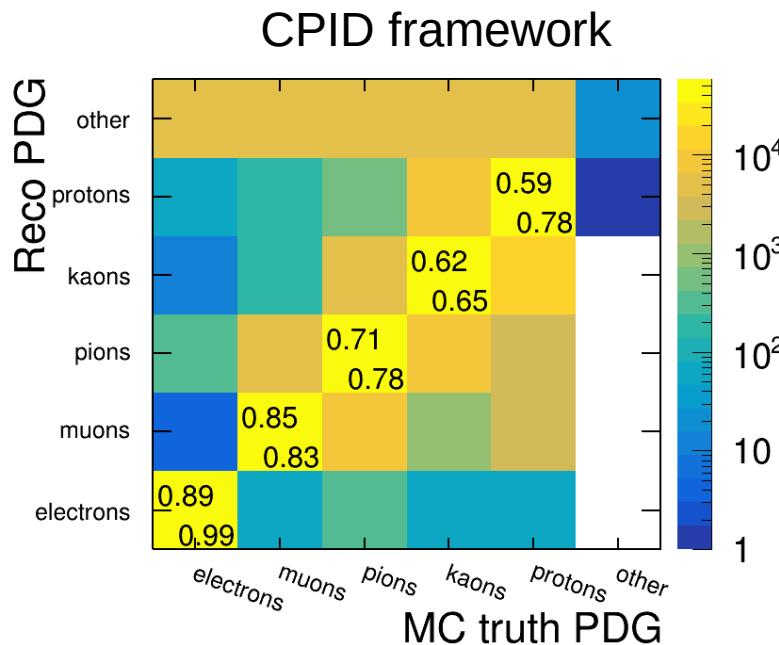
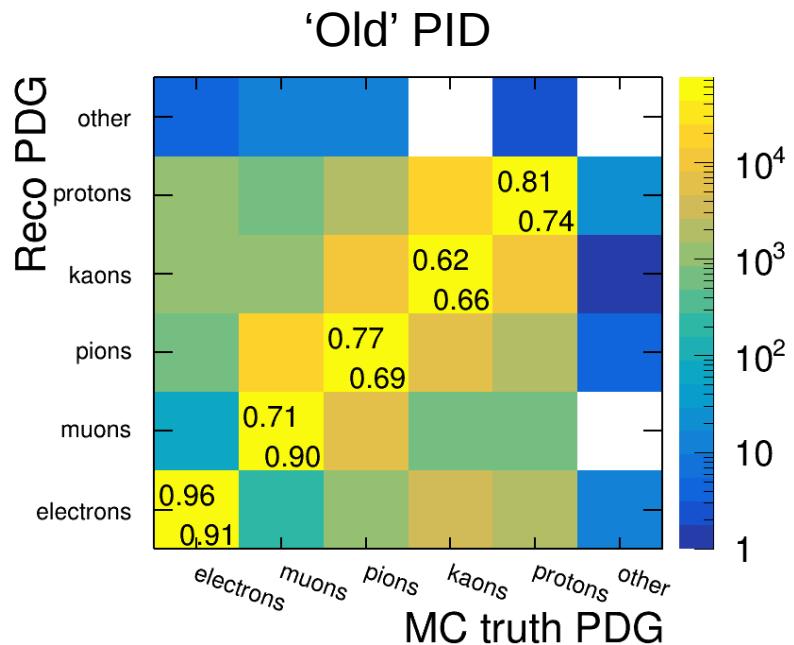


CPID, BDT result



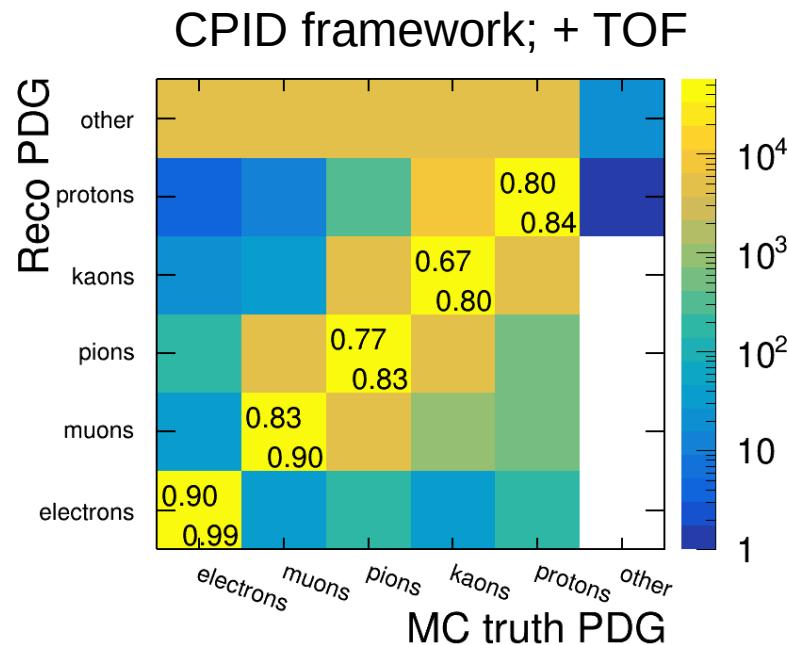
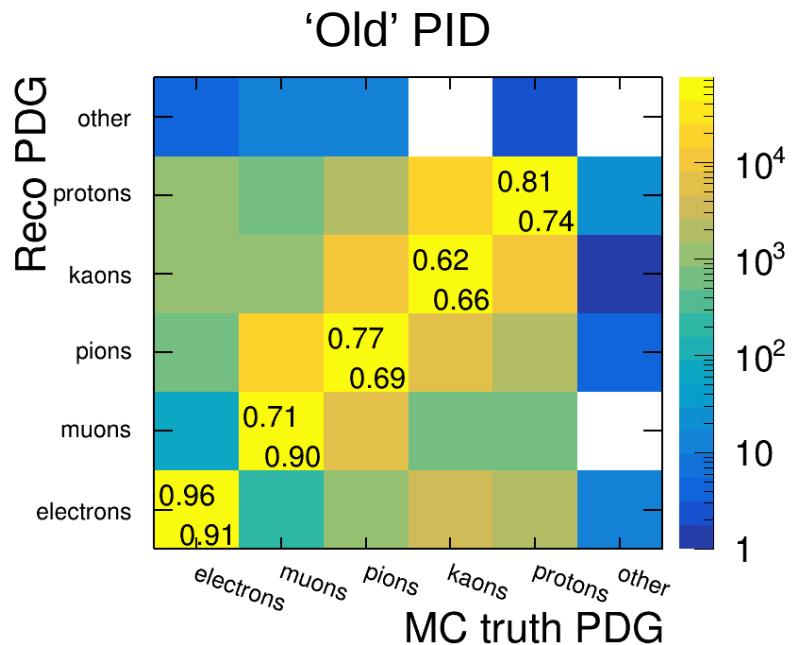
## Example 2: Multiclass Assignment Matrix

- $dE/dx +$  calorimeter cluster shapes
- Single particle ‘calibration’ events, flat in  $\log(p)$  and  $\cos(\theta)$
- $e, \mu, \pi, K, p$ ; multiclass BDT; assignment matrix with  $^{eff}/pur$  on diagonal
- Simple BDT has issue with misassignment, but already generates similar reco purity



## Example 2: Multiclass Assignment Matrix

- $dE/dx +$  calorimeter cluster shapes
- Single particle ‘calibration’ events, flat in  $\log(p)$  and  $\cos(\theta)$
- $e, \mu, \pi, K, p$ ; multiclass BDT; assignment matrix with  $^{eff}/pur$  on diagonal
- Addition of TOF gives immediately better result – previously hard, easy in CPID



# Summary

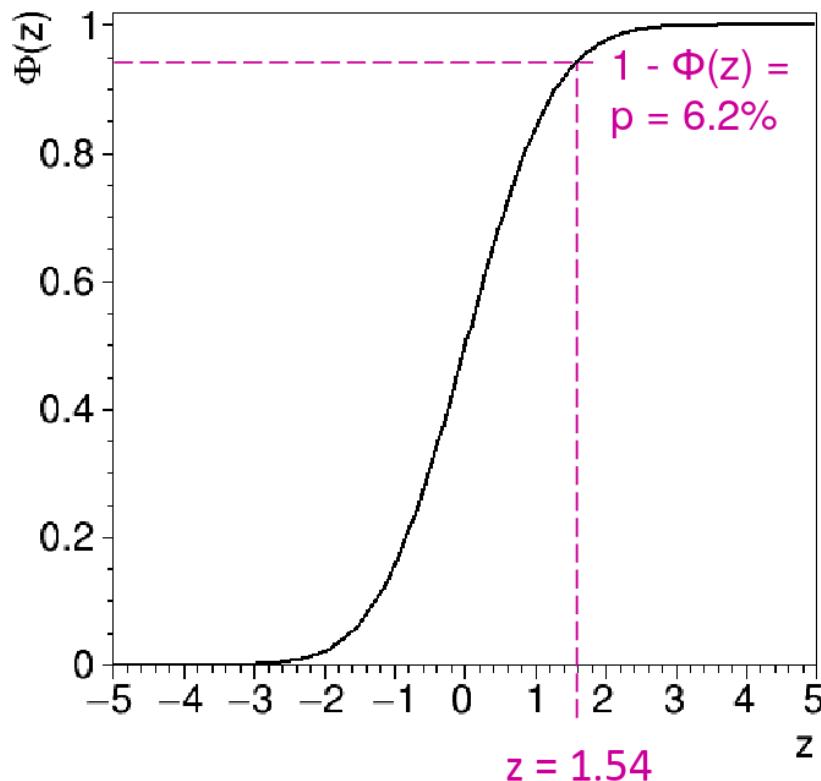
- New Comprehensive PID framework under development
- Aims to provide common platform for future e+e- Higgs factories
- Allow for
  - combining and comparing PID technologies
  - assessing on full detector level with robust performance quantities
  - easy-to-use retraining and flexible adaptation
- First performance indicators already comparable to state-of-the-art
- Much more to come!
- Your feedback and input are welcome!

# What is particle identification / PID

- Identification of the species of high energy particles
- E.g.  $e$ ,  $\mu$ ,  $\gamma$ ,  $\pi$ ,  $K$ ,  $p$ ,  $\Lambda$ ,  $n$ , [*whatever is detector-stable*]
- Origin: collider interaction point, fixed target, annihilating DM, active galactic nucleus
- Average over large statistics, or event-by-event identification of individual particles  
→ improve precision physics via event-by-event PID!

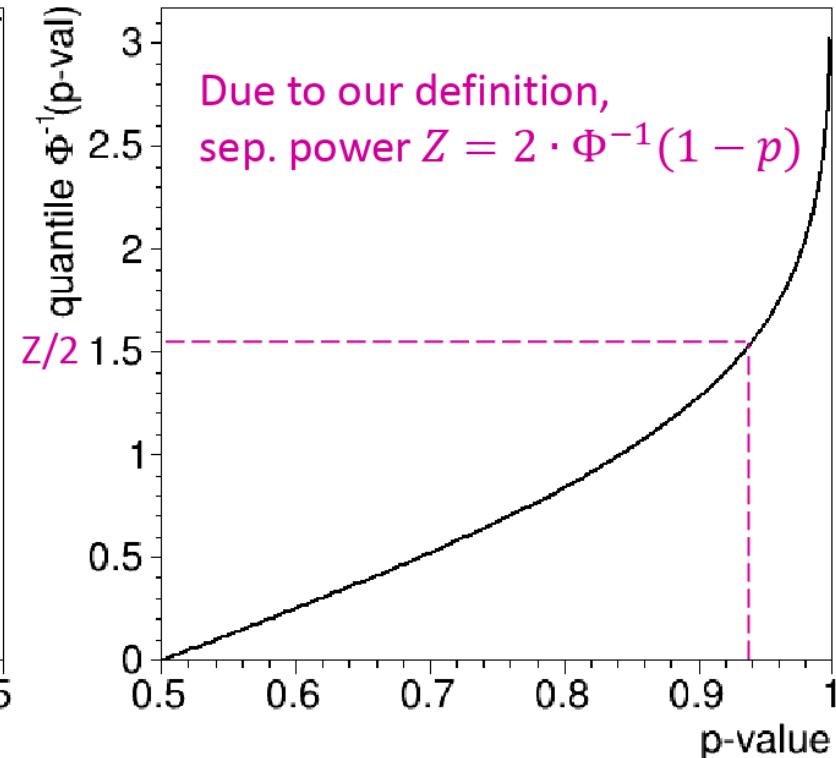
# p-value assessment

Gaussian quantile is inverse of distribution function  $\Phi(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-z^2/2} dz$



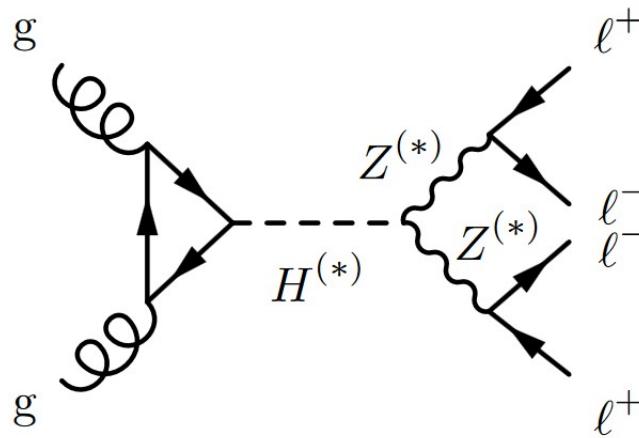
```
ROOT::Math::gaussian_cdf(z)
```

Talk by K. Götzen, PANDA, 2017

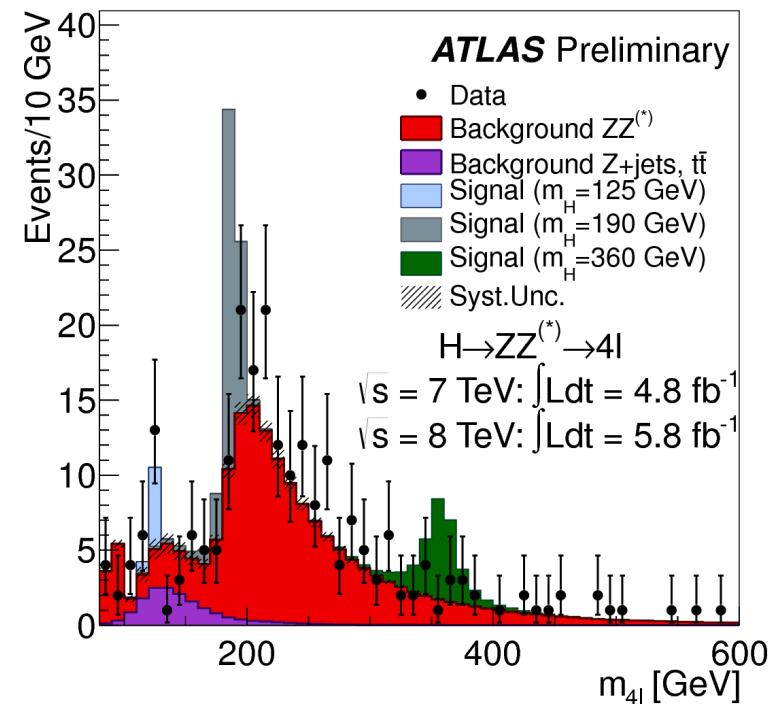


```
ROOT::Math::gaussian_quantile_c(p,1)
```

- $H \rightarrow ZZ^* \rightarrow 4l \text{ (e}/\mu)$

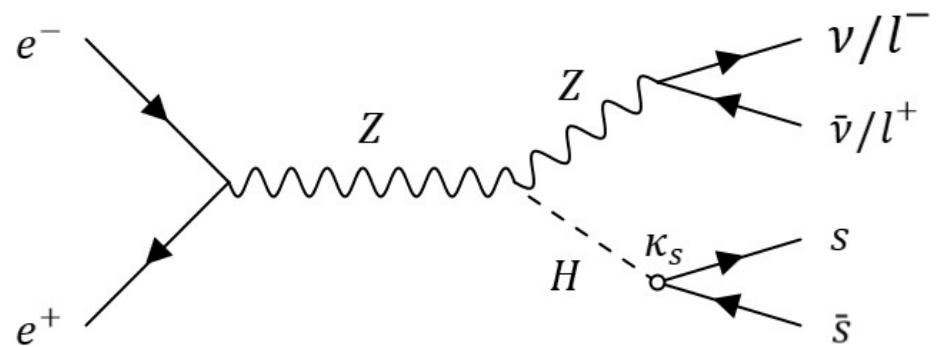


- easily identifiable via muon chambers and ECal
- more difficult: e,  $\mu$  at low momenta,  $\pi^+$  vs.  $K^+$  vs.  $p^+$ ,  $K^0_S$  and  $\Lambda^0$

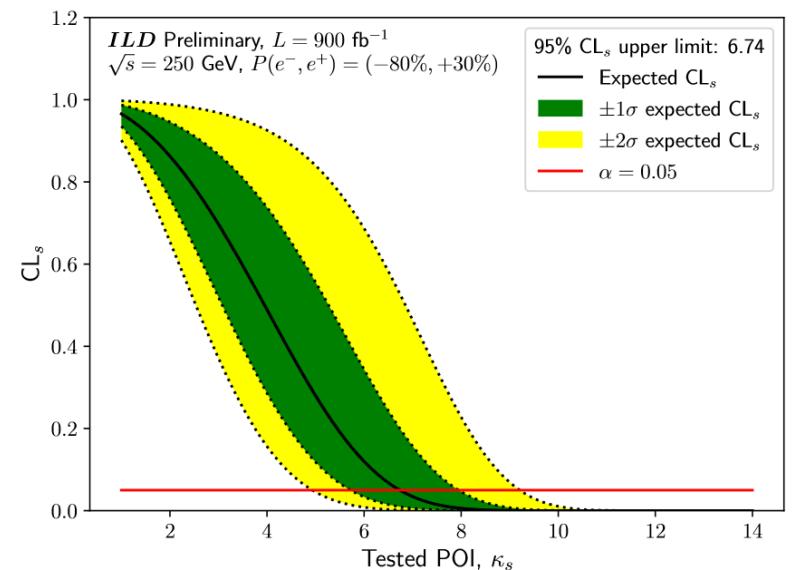
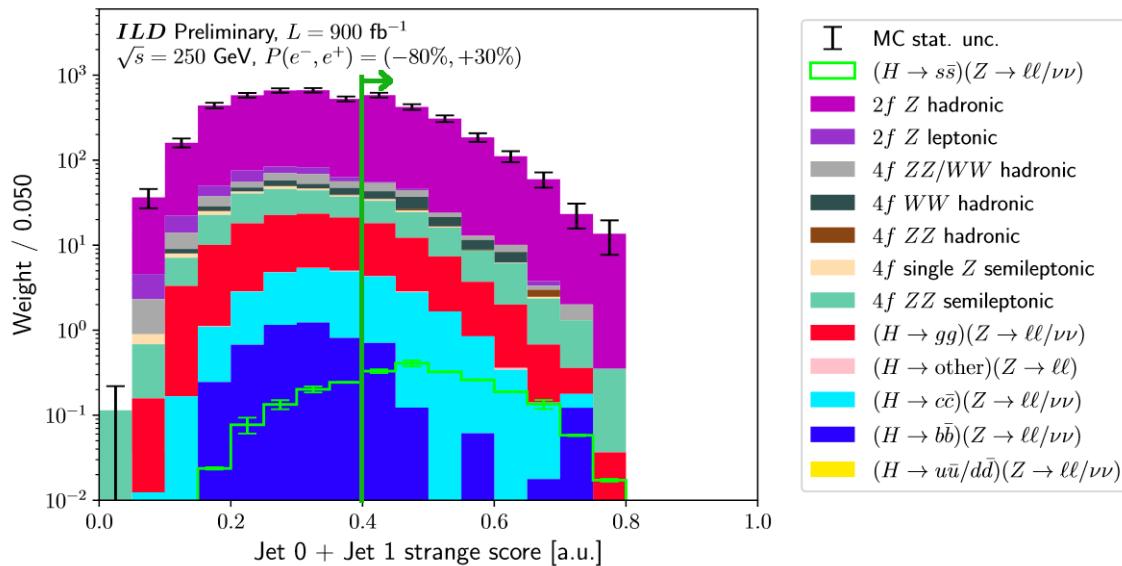


# Higgs to strange

- Study Higgs to strange coupling
- Cut-based analysis, final cut:  
developed strange tagger using  $K^\pm$ ,  $K^0_s$ ,  $\Lambda^0$   
→ allows to cut background by factor 3
- Results in upper limit on  $\kappa_s < 6.7$

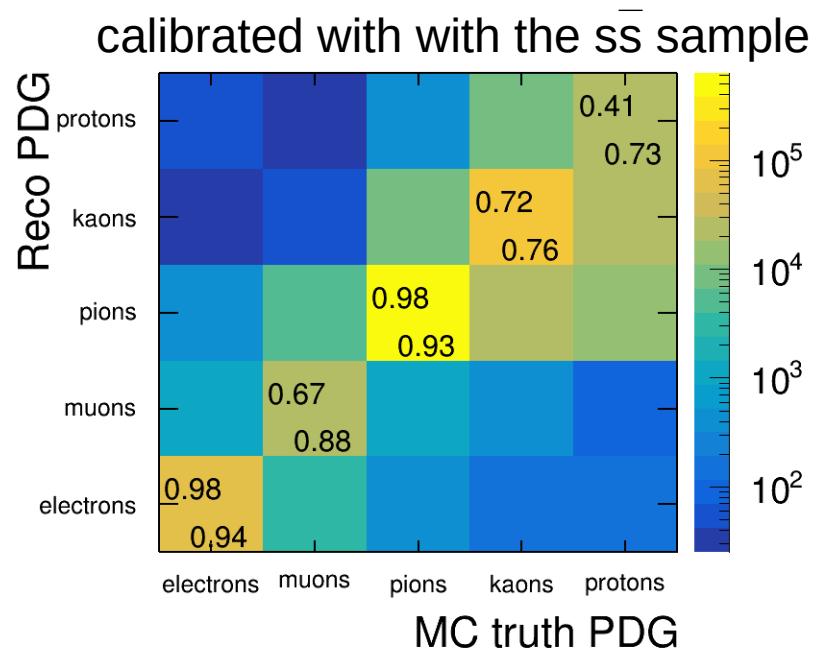
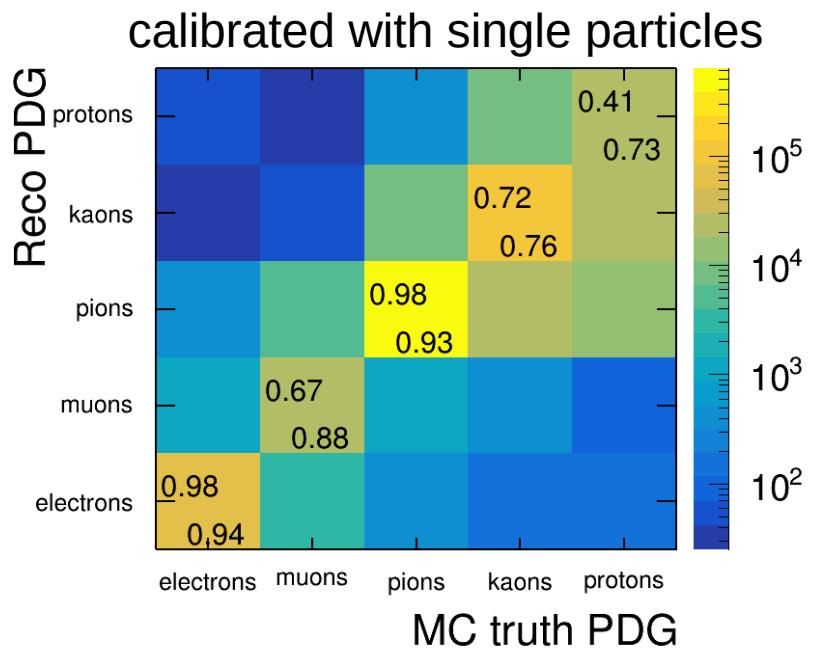
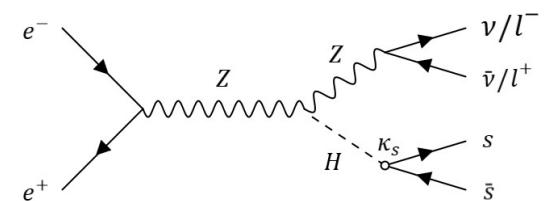


[<https://arxiv.org/abs/2203.07535>]



# Example 3

- $dE/dx$ , TOF, cluster shapes
- $e^+e^- \rightarrow ZH \rightarrow l\bar{l} s\bar{s}$
- Kaon ID vs. all others, assignment matrix
- Kaons well identifiable; calibration source not very relevant



# View of the steering file

- Steering file

- input sample
- observables algorithms
- signal categories PDGs
- evaluation algorithm
- weight file
- sample cuts etc.

```
<processor name="MyComprehensivePIDProcessor" type="ComprehensivePIDProcessor">

    <parameter name="PFOCollection" type="string" value="PandoraPF0s"/>

    <parameter name="inputAlgoSpecs" type="StringVec">
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        TOF:TOF0
        TOF:TOF10
        TOF:TOF50
        dEdx_RCD:dEdx_RCD
    </parameter>

    <parameter name="dEdx.F" type="FloatVec" value="1 2 3"/>
    <parameter name="dEdx.S" type="StringVec" value="a b c"/>

    <parameter name="TOF0.S" type="StringVec" value="TOFEstimators0ps" />
    <parameter name="TOF10.S" type="StringVec" value="TOFEstimators10ps" />
    <parameter name="TOF50.S" type="StringVec" value="TOFEstimators50ps" />

    <parameter name="dEdx_RCD.F" type="FloatVec">
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        6.49143971e-02  1.55775592e+03   9.31848047e+08   2.32201725e-01  2.50492066e-04
        6.54955215e-02  8.26239081e+04   1.92933904e+07   2.52743206e-01  2.26657525e-04
        7.52235689e-02  1.59710415e+04   1.79625604e+06   3.15315795e-01  2.30414997e-04
        7.92251260e-02  6.38129720e+04   3.82995071e+04   2.80793601e-01  7.14371743e-04
    </parameter>

</processor>
```