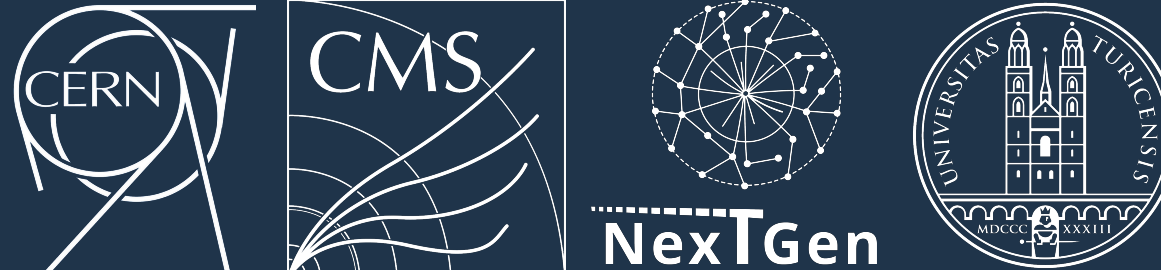


FPGA-Optimized ML for low-latency electron identification and p_T regression with the CMS Phase-2 L1 trigger

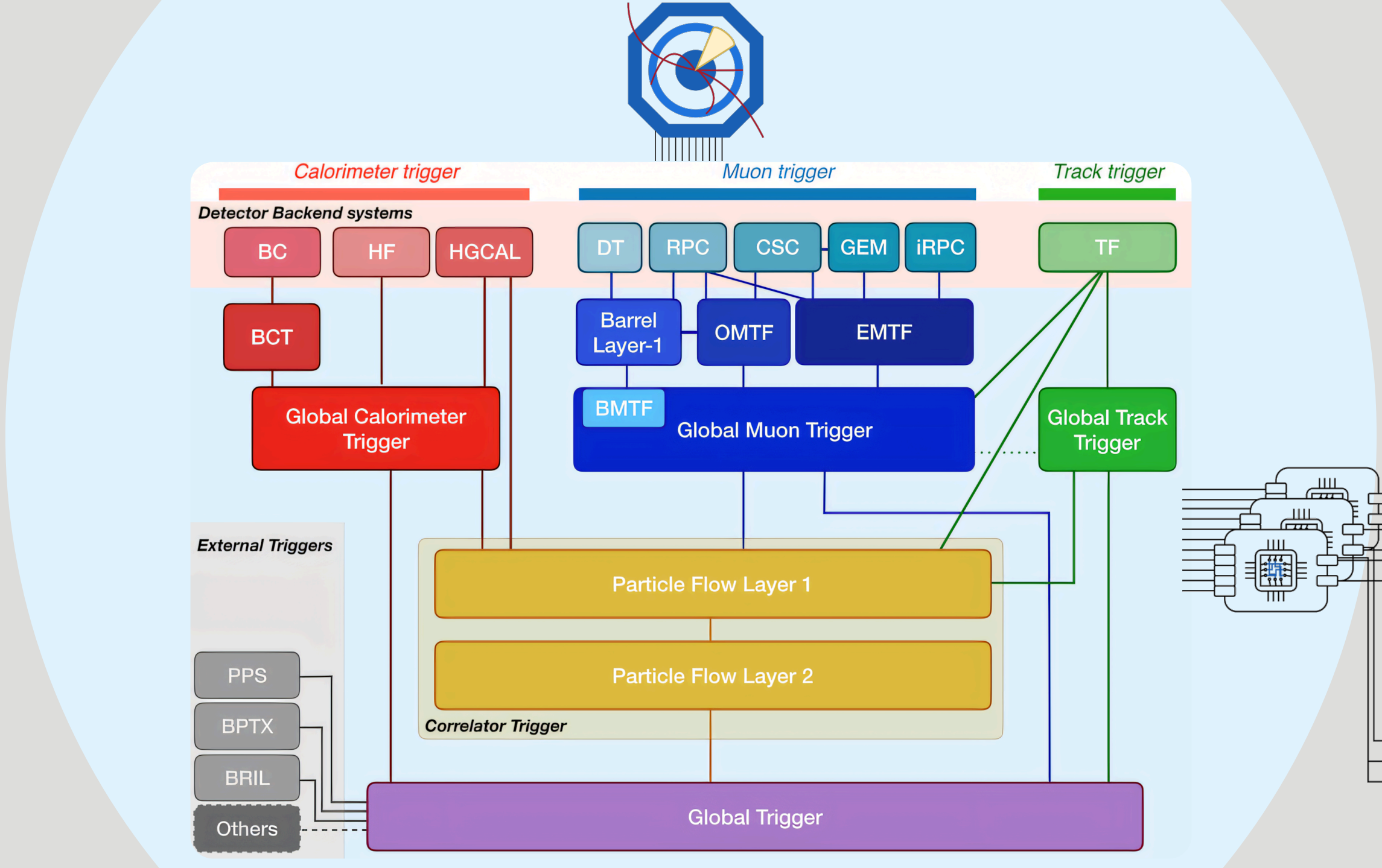
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¹CERN, ²University of Zürich
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1 CMS Phase-2 L1 trigger

The CMS Level-1 Trigger (L1T) is a real-time system that selects the most relevant collision events to reduce the data for further processing. For Phase-2 of the LHC, the trigger is being upgraded to handle higher luminosity.

- Access to both **calorimeter** and **tracking information**.
- Extended **latency** frame of **12.5 μ s**
- The **correlator trigger** (CT) creates **physics objects** candidates combining information from multiple subdetectors.



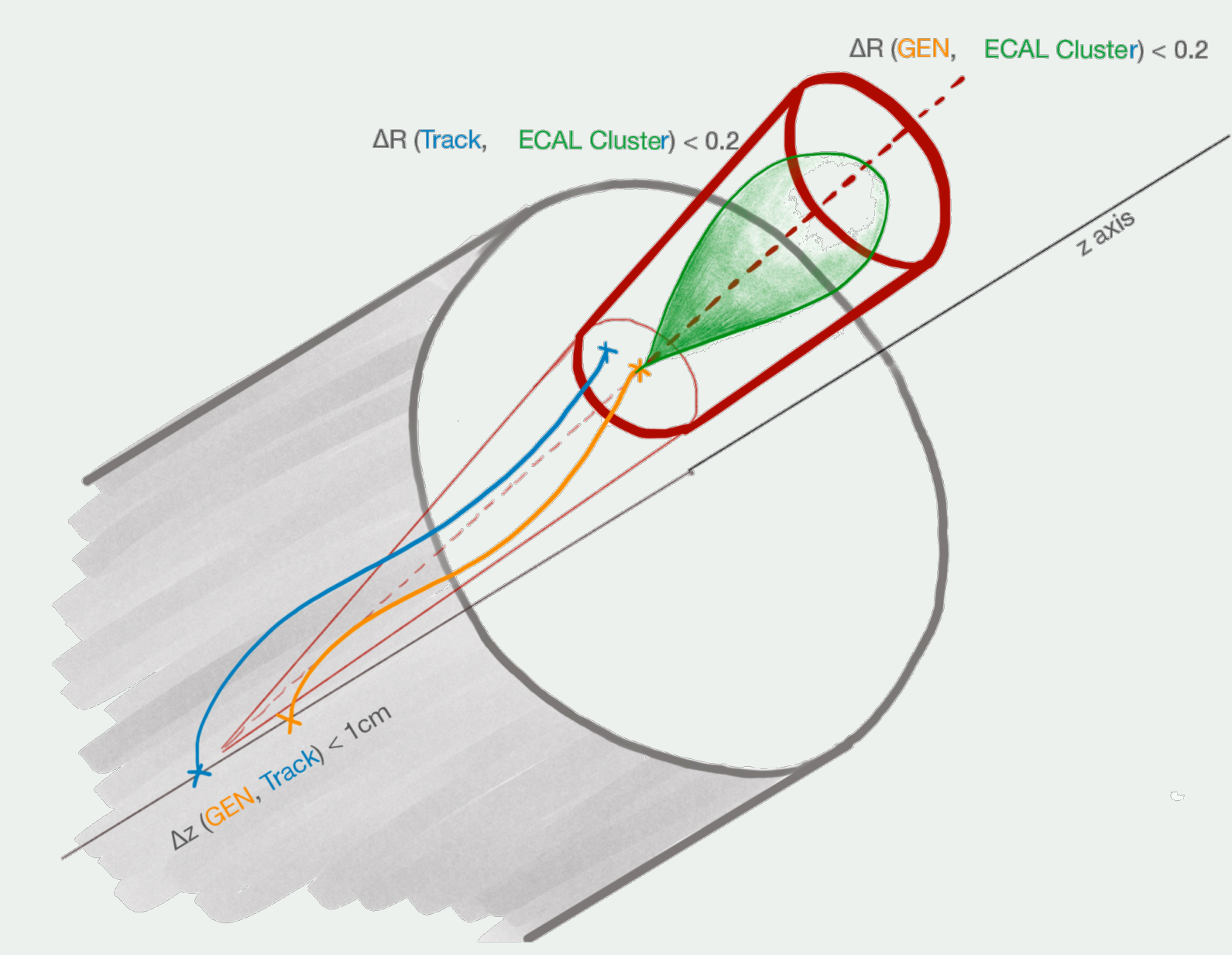
2 L1 Scouting

40MHz **online analysis system** that complements the L1 Trigger capturing the events that would be lost otherwise.

It exploits improved efficiency of low- p_T trigger objects not fitting standard menu (too much for rate budget)

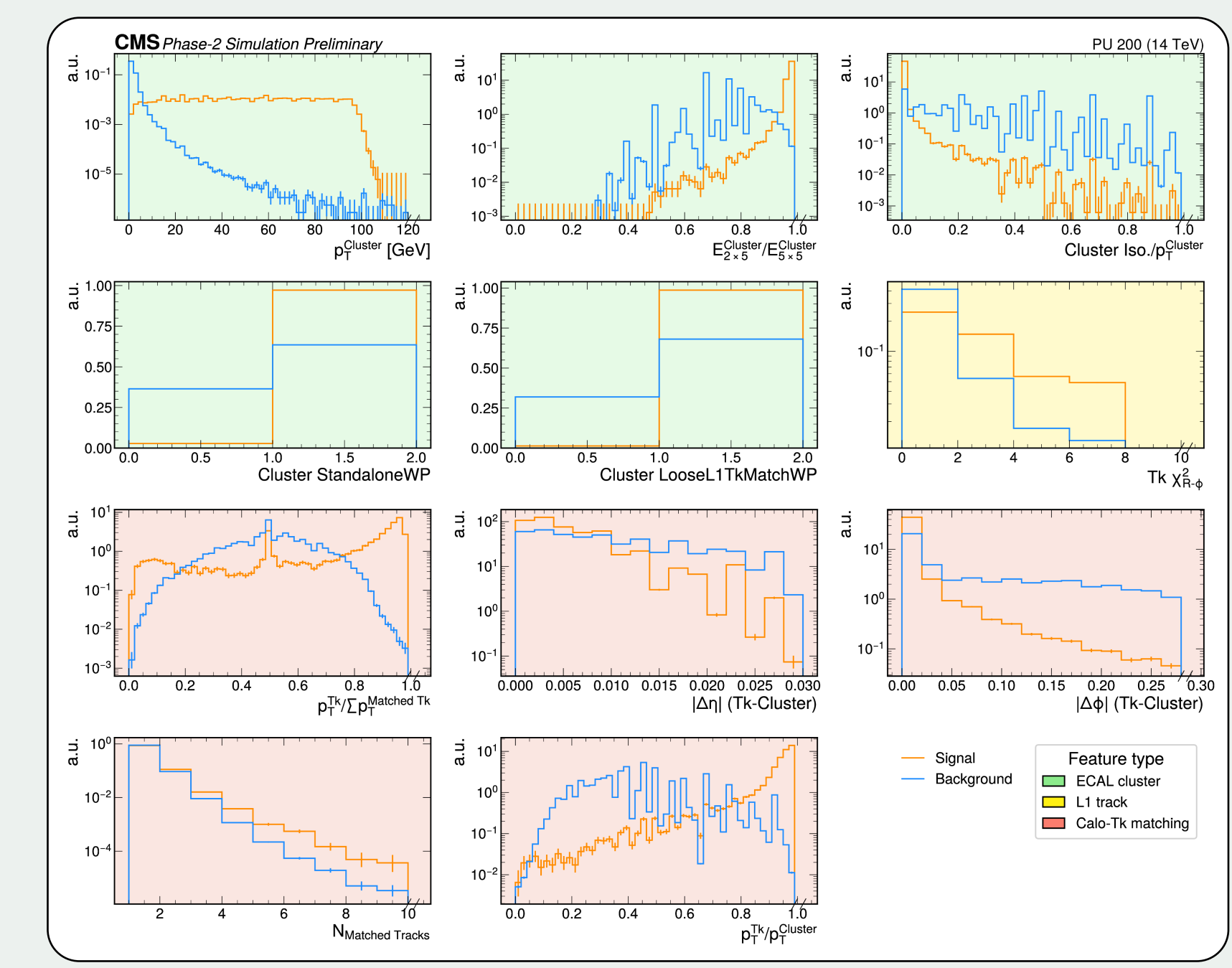
3 Track-matched electron candidates

Track-matched Electron candidates (TkEle) are built by the **correlator trigger**, linking tracks to calorimeter clusters with an **elliptical matching** in η - ϕ



- Baseline approach:**
- $p_T^{Tk} > 10$ GeV
 - $(|\Delta\eta| < 0.025, |\Delta\phi| < 0.07)$ for $|\eta| < 0.9$
 - $(|\Delta\eta| < 0.015, |\Delta\phi| < 0.07)$ for $0.9 < |\eta| < 1.479$
- Tight geometrical match (Elliptic)**

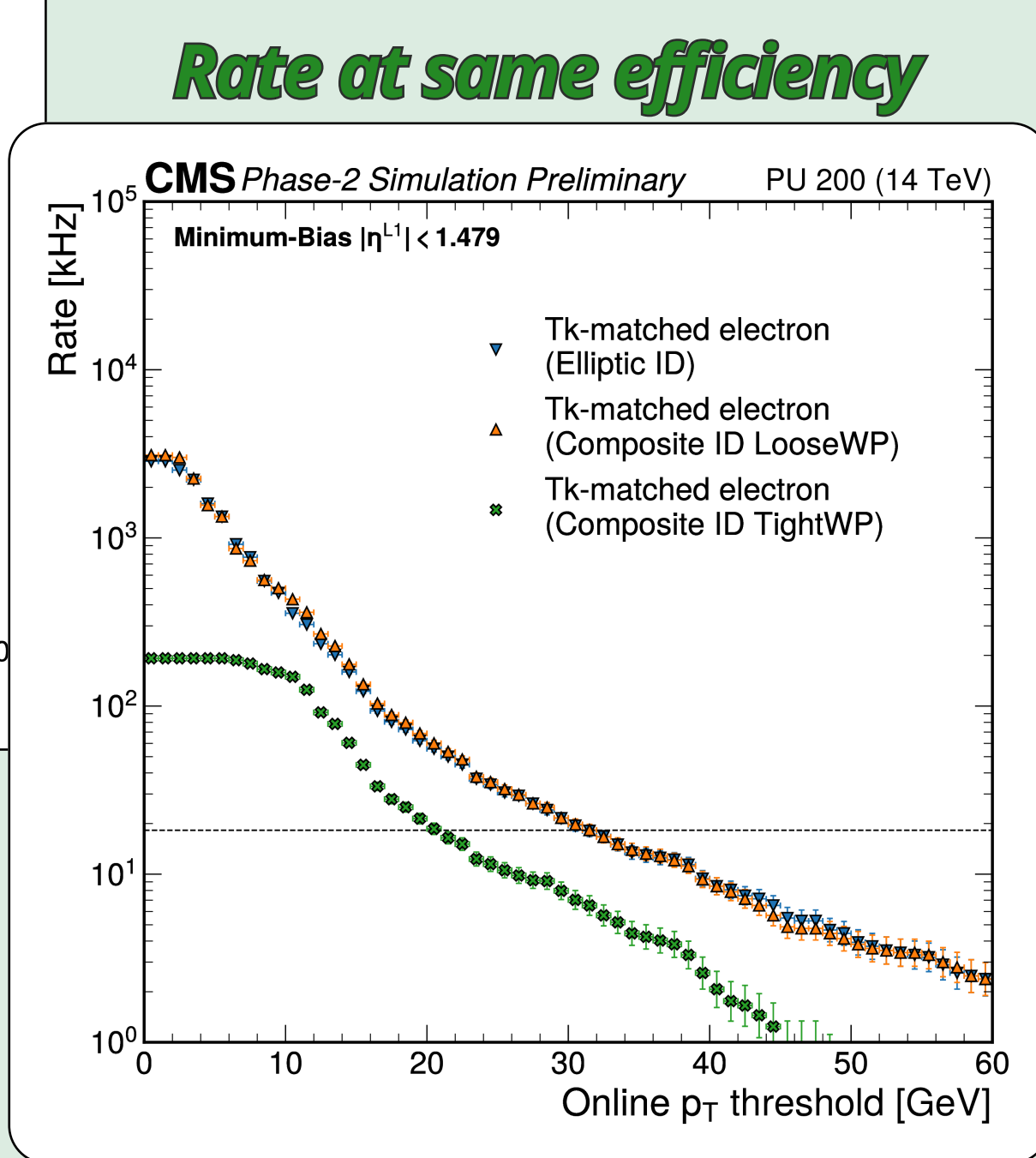
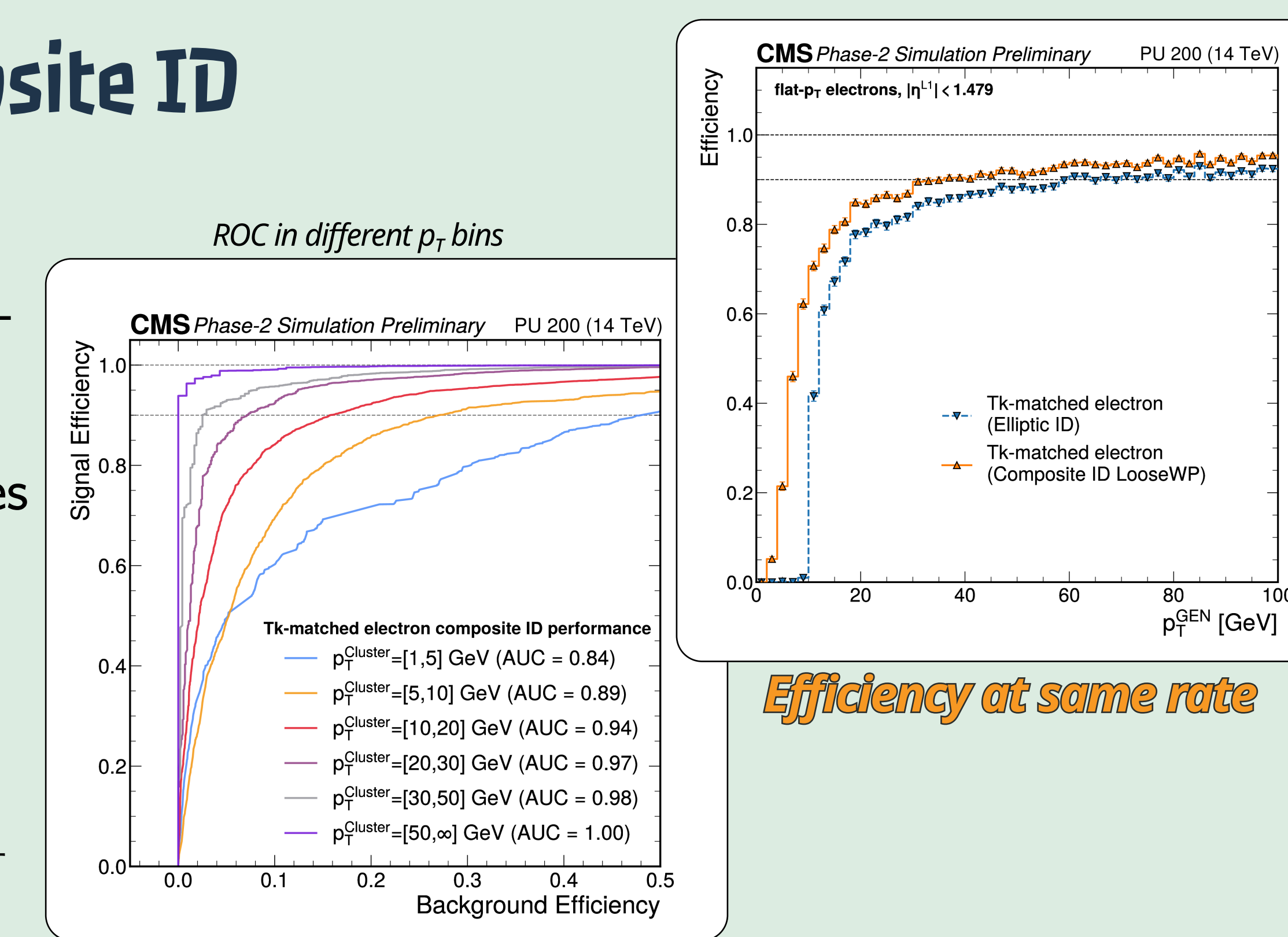
- Novel approach:**
- Loose elliptical match
 - $(|\Delta\eta| < 0.03, |\Delta\phi| < 0.3)$
 - Candidates classified by a BDT (**Composite ID**)



4 Identification: Composite ID

- BDT classifier**
- **Signal:** TkEle matched to generated electrons in a flat- p_T , flat- η sample at PU200
 - **Background:** TkEle candidates in a PU200 MinBias sample

Background candidates are reweighted in order to **flat the p_T distribution** to avoid the model from imposing a tight cut on the p_T



Quantization & Implementation

The models have been synthesized in firmware using **Conifer** and **Verilog**

All the input features are rescaled to the $[-1, 1]$ interval to use less bits

```
cfg = {
  "Input Precision" : ap_fixed<8,1>,
  "Score Precision" : ap_fixed<11,4>,
  "Board" : "Xilinx Virtex UltraScale+ VU13P",
  "Clock" : 4.166 #ns (248 MHz)
}
```

	BRAM	DSP	FF	LUT
SLR	0.0%	0.0%	0.9%	7.5%
Total	0.0%	0.0%	0.02%	1.9%

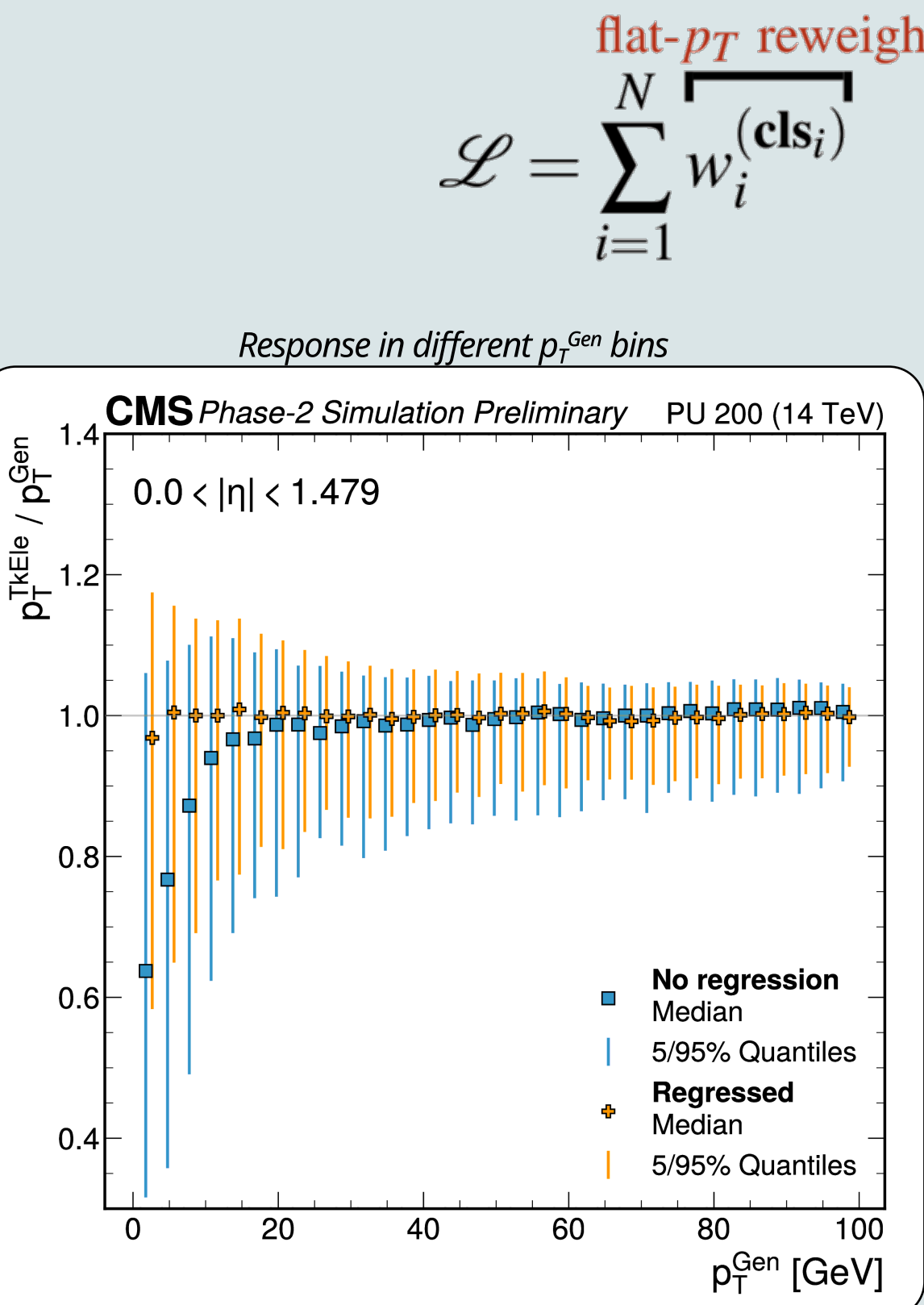
	Clock cycles	Latency
Clock (4.166 ns)	7	29.17 ns

```
cfg = {
  "Input Precision" : ap_fixed<11,1>,
  "Score Precision" : ap_fixed<12,3>,
  "Board" : "Xilinx Virtex UltraScale+ VU13P",
  "Clock" : 4.166 #ns (248 MHz)
}
```

	BRAM	DSP	FF	LUT
SLR	0.0%	0.0%	1.13%	6.2%
Total	0.0%	0.0%	0.03%	1.5%

	Clock cycles	Latency
Clock (4.166 ns)	6	25.0 ns

5 p_T regression



L1 signal regression term

$$\text{cls}_i \frac{|y_i - p_{T_i}^{\text{Gen}} / p_{T_i}^{\text{Calo}}|}{\text{Med} \left[\left| 1 - p_{T_j}^{\text{Gen}} / p_{T_j}^{\text{Calo}} \right| \right]}$$

Background penalization term

$$(1 - \text{cls}_i) \cdot (\alpha + \beta \text{Min}[p_{T_i}^{\text{Calo}}, p_T^{\text{thr}}]) \cdot \text{Max}[0, y_i - t]$$

Target y_i : $p_T^{\text{Gen}} / p_T^{\text{TkEle}}$ (output of the model = p_T corrective factor)

Penalization linear in p_T (up to p_T^{thr}) Penalize output $\geq t$

$\alpha = 8, \beta = 1.05 \text{ GeV}^{-1}, t = 0.975, p_T^{\text{thr}} = 15 \text{ GeV}$

Signal term:

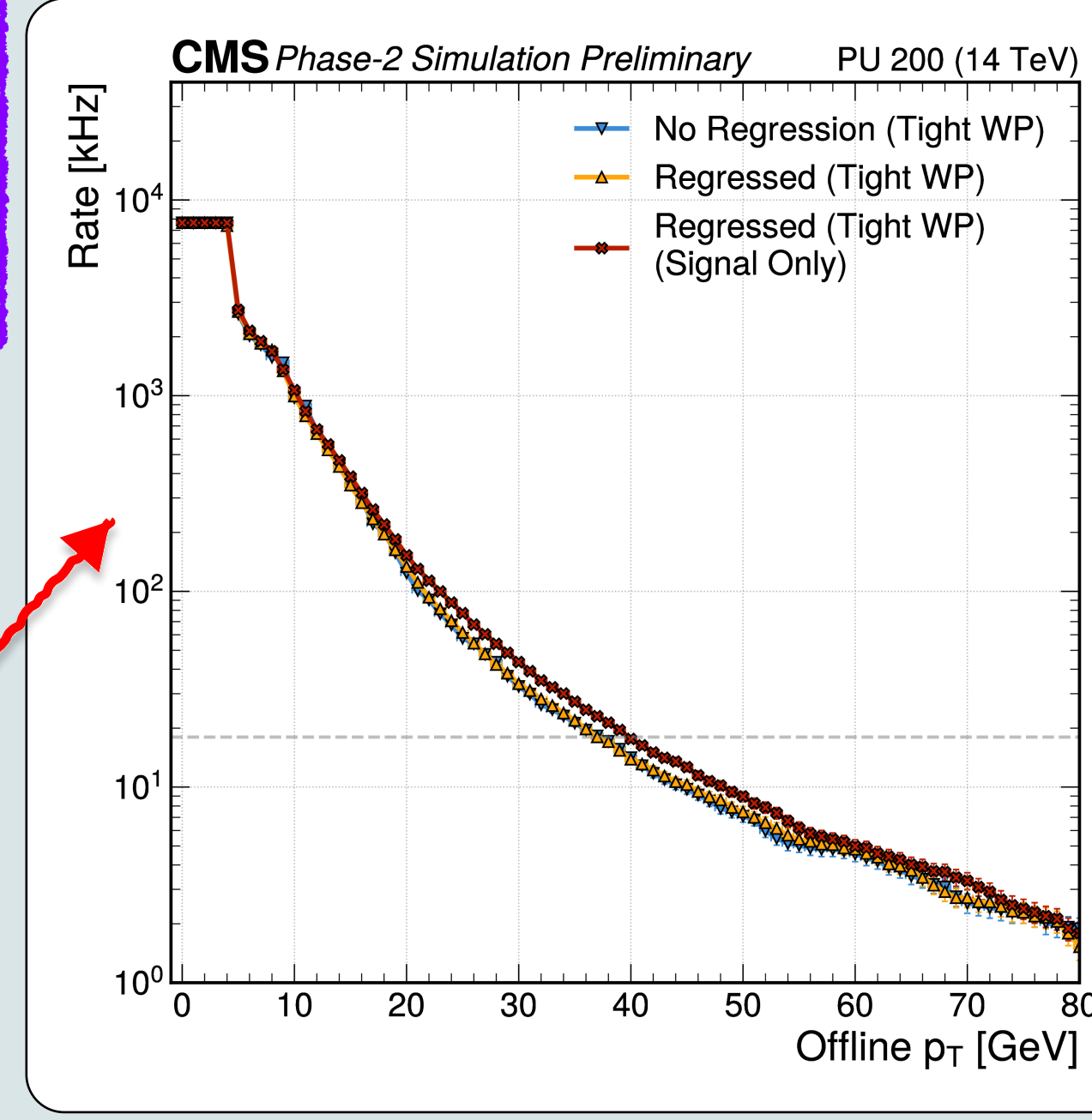
- L1 regression loss (insensitive to outliers, optimize median response)
- Heteroscedasticity-adjusted (Non constant variance)

Background term:

- p_T increment is penalized
- Penalization linear in p_T (up to p_T^{thr}) (Higher p_T background TkEle are signal-like+ typical trigger threshold at 12/25/36 GeV)

Avoid the model from increasing the p_T of background electrons more than signal electron (i.e. increase the rate)

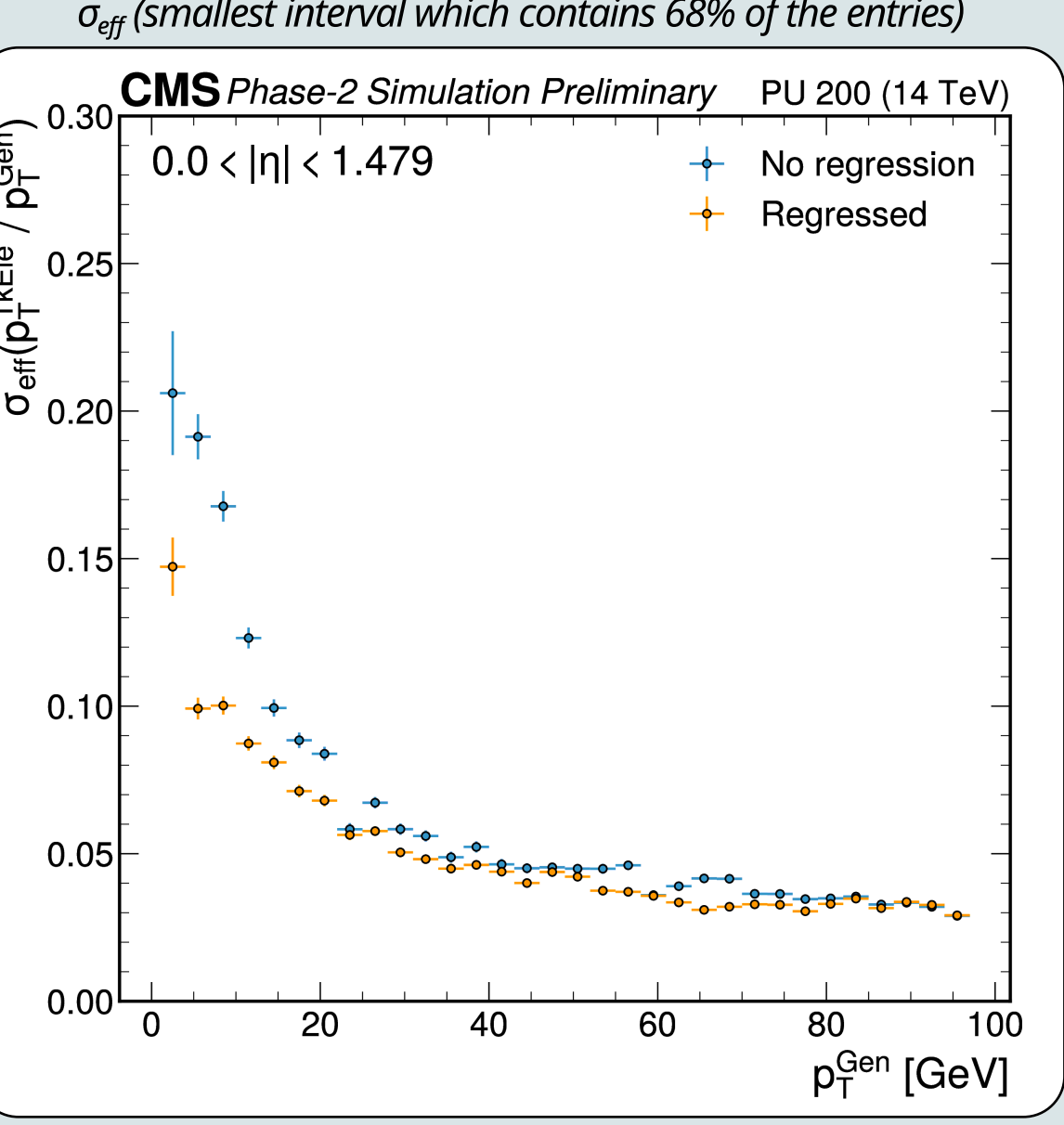
Feature Importance:



The loss has **always null second derivative**

XGBoost employs the **Newton-Raphson** method to perform the minimization, which require a **positive definite Hessian**.

- using the L1 loss, XGBoost find the **split** of each node using a **quantile based approach** instead of Newton-Raphson
- The python API does not allow to change the split method: **the proposed loss require to be implemented in C++, recompiling XGBoost.**



L1 Scouting benchmark

Hidden Abelian Higgs model

Low-mass Drell-Yan produced Dark-Photon Z_d that decays to a pair of electrons

5 mass points were generated.

Selections:

- TkEle matched to generated electrons
- Opposite sign pair
- $p_T > 4$ GeV
- ID score > -0.3
- $|\Delta\eta_z| < 0.65$

$\sigma_{\text{eff}}(m_{ee})$ 23% lower after the regression

