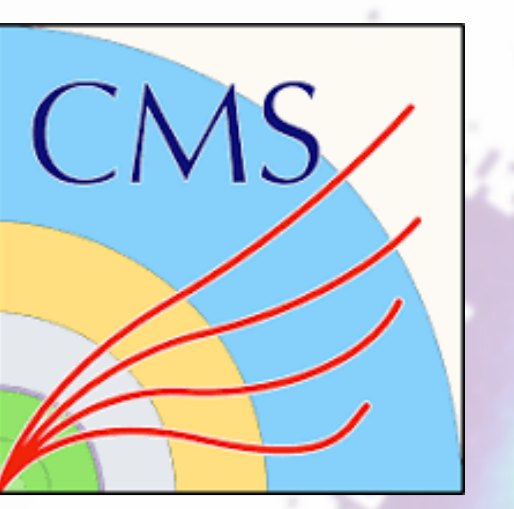


Generating parton-level events from CMS reconstructed events with Conditional Normalizing Flows

Antonio Petre on behalf of the CMS collaboration



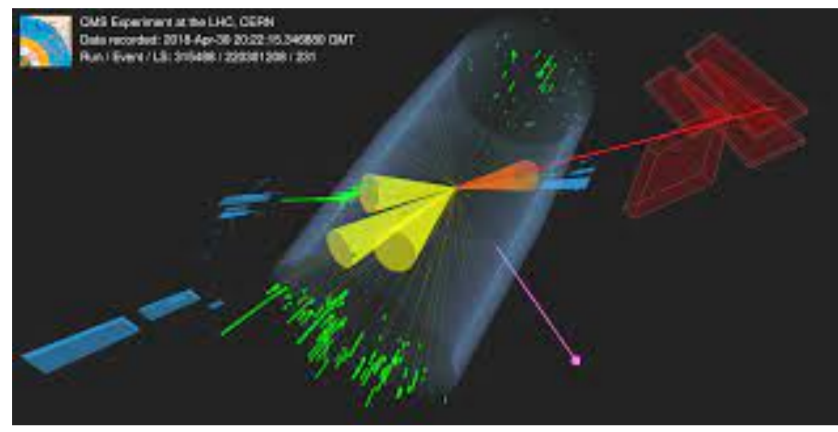
ETH zürich

Matrix Element Method (MEM)

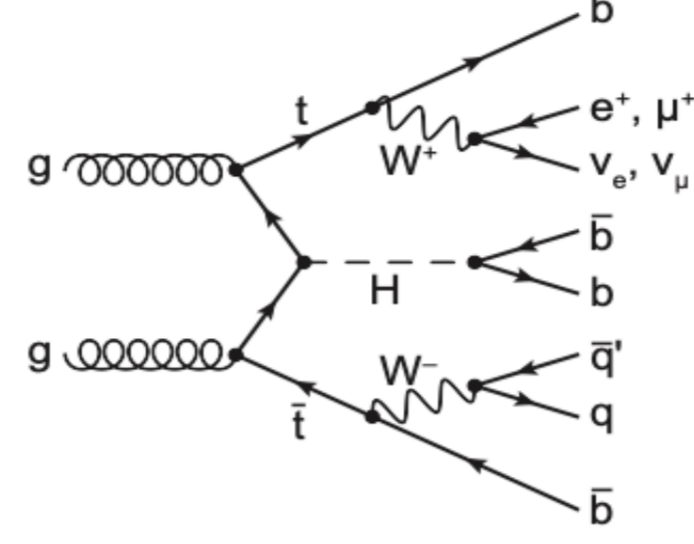
Matrix element method estimates the probability of a single reconstructed event \vec{Y} to be generated by a physical process defined by θ parameters:

$$\mathcal{P}(\vec{Y}|\theta) \propto \int_{\phi} d\vec{X} M(\vec{X}|\theta)^2 \cdot Pdf \cdot W(\vec{Y}|\vec{X})$$

$\vec{Y} =$



$\vec{X} =$



Pros & Cons:

- ✓ It can be used for hypothesis testing or parameter estimation
- ✓ Maximizes the amount of theoretical information for the discriminator
- ✓ It is not bound to a specific process
- ✗ Integral computation is very CPU demanding due to jet-parton matching (combinatorial problem)
- ✗ Many approximations used to speedup the computation e.g. jet-parton alignment

New Method & Normalizing Flows

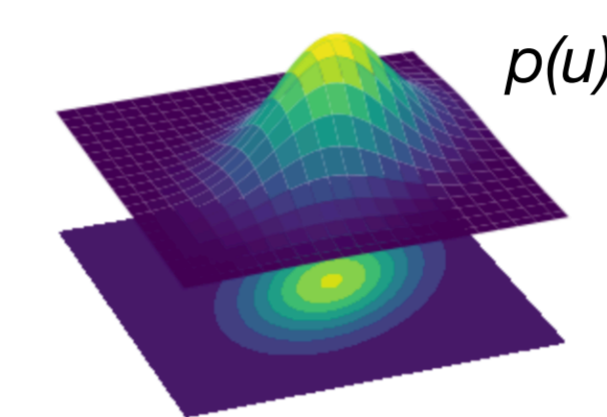
Our goal is to model the conditional probability of parton-level events given a reconstructed event using generative machine learning architectures, more specifically **normalizing flows**:

$$\int_{\phi} d\vec{X} M(\vec{X}|\theta)^2 \cdot Pdf \cdot W(\vec{Y}|\vec{X})$$

Use importance sampling: $\vec{X} \sim \mathcal{P}(\vec{X}|\vec{Y}, \theta)$

$\mathcal{P}(\vec{X}|\vec{Y}, \theta)$ found using **normalizing flows**

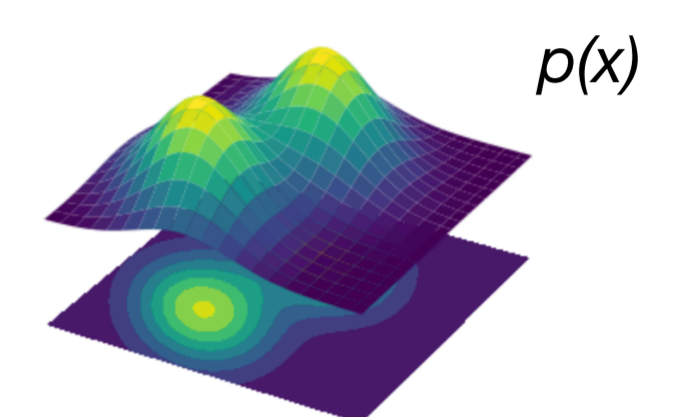
Flow models: Machine-learned maps (transformations) between probability distributions



'Base' distribution
Easy-to-evaluate
Easy-to-sample

$$x = T(u)$$

Flow transformation:
• Invertible
• Tractable Jacobian
• Many params. for optimization



'Target' distribution
Multi-modal

Dataset and Legacy Performance

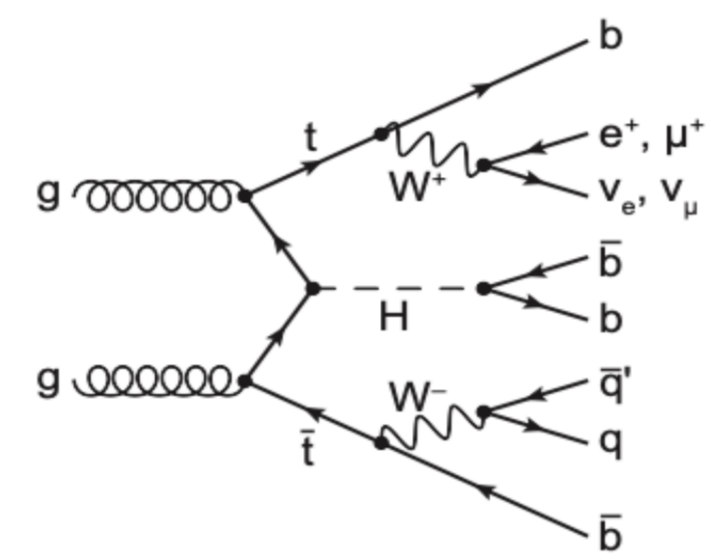
Process used: **single-lepton channel in the ttH(→bb) with an additional radiation**

Events details:

- Generated with a CM energy of 13 TeV
- Pileup profile of LHC Run II (~30-50 simultaneous pp collisions)
- Full CMS detector simulation, including standard RUN II reconstruction

Data selection:

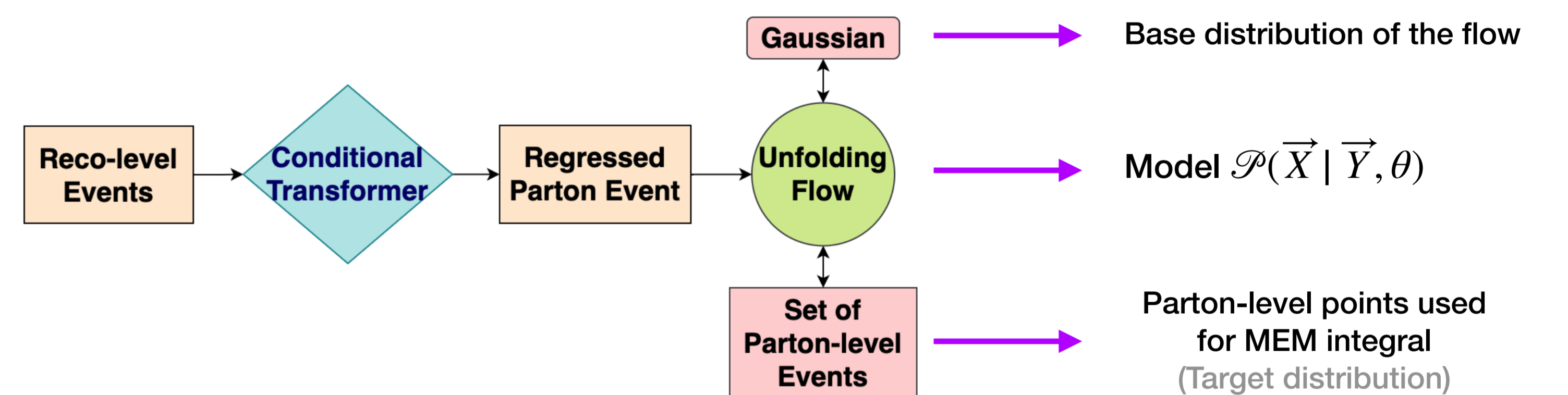
- At least 4 jets with $p_T > 30$ GeV and $|\eta| < 2.4$
- At least 3 jets tagged as b-quark flavour
- One prompt reconstructed lepton with $p_T > 30$ GeV
- MET > 20 GeV



Legacy ttH analysis: **MEM used for discriminating between signal and background**

Computation performance: **~ 1 min/event** → **speedup needed**
normalizing flow is a good candidate

Our Strategy



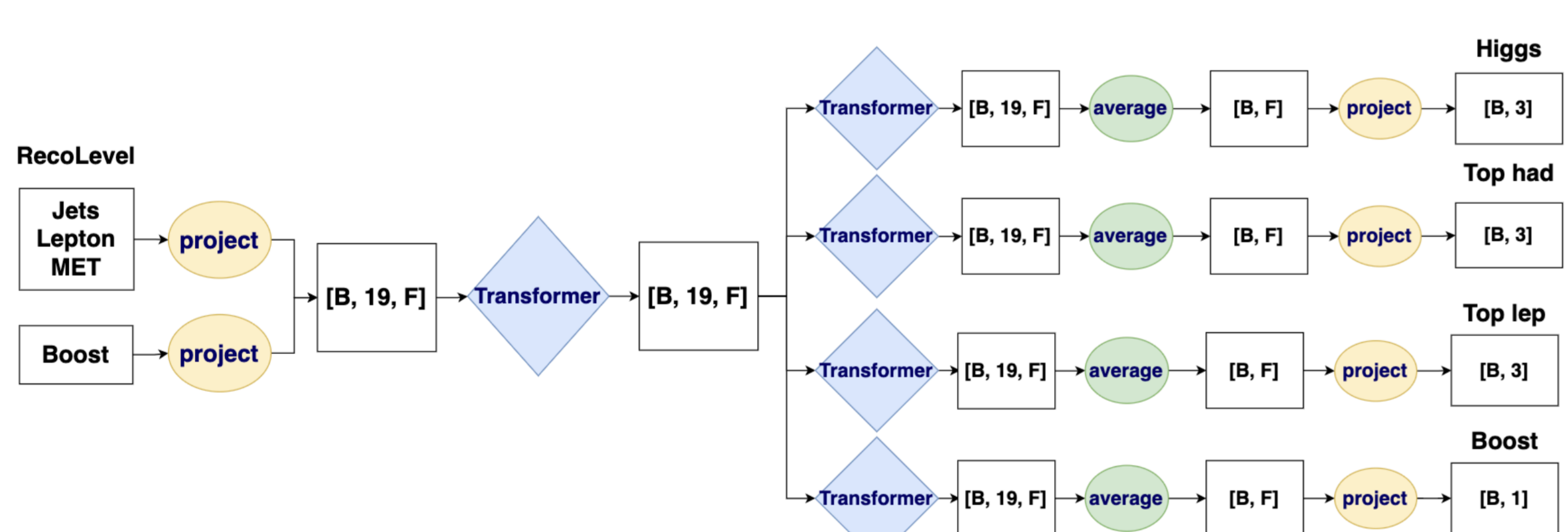
Reco-level Events → Jets + Lepton + MET : p_T, η, ϕ , btag score, SPANET output

Regressed Parton Event → Higgs + two tops + additional radiation : p_T, η, ϕ
In practice, RAMBO parametrization was used

Conditional Transformer → Regress the **parton-level event** for a given **reco-level event**
Extracts a latent information vector which conditions the **Unfolding Flow**

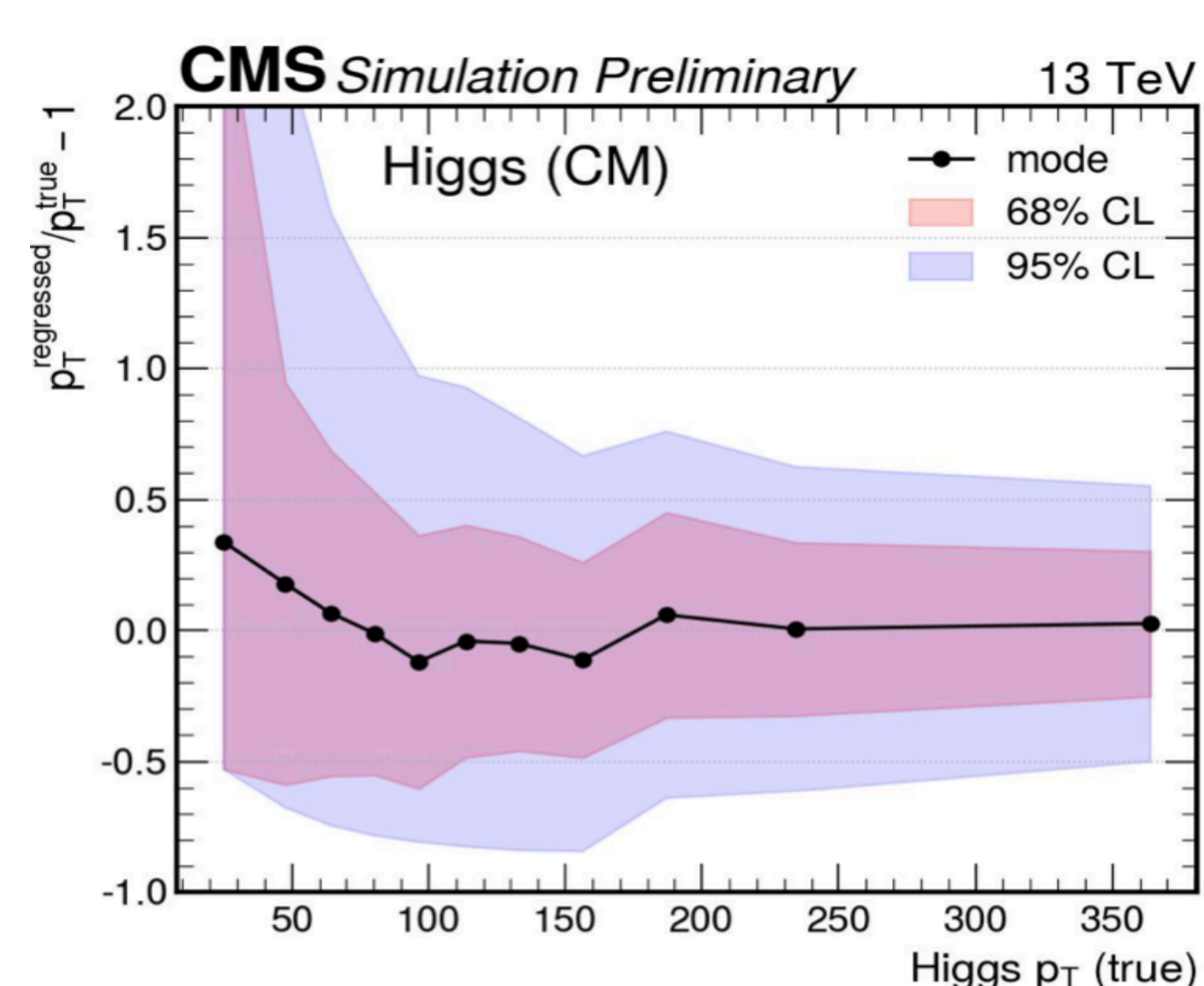
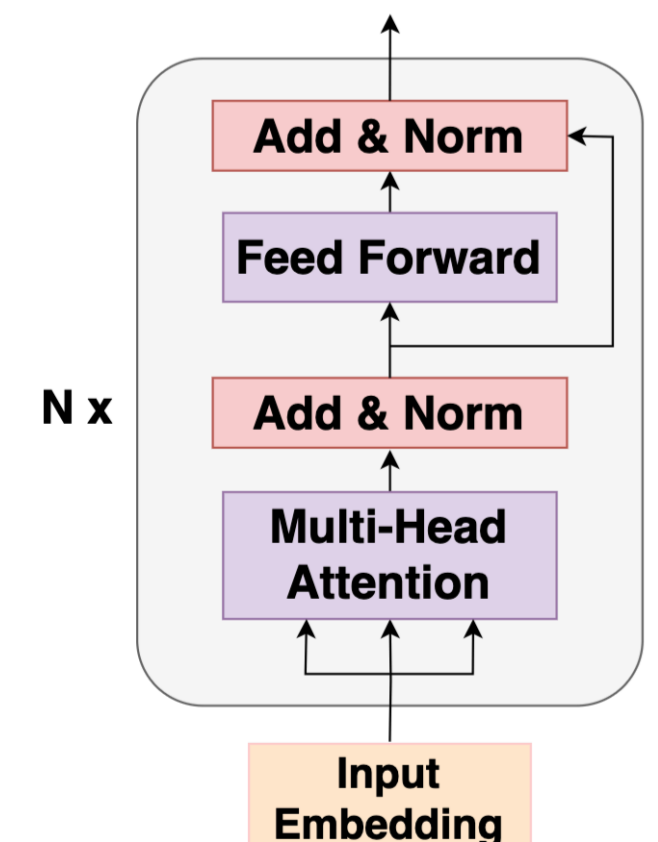
Unfolding Flow →
✓ Generates plausible phase-space points compatible with reco-objects
✓ Reduces assumptions on partons' directions
✓ Handles events with out-of-acceptance final state objects and multiple jet multiplicities

Conditional Transformer

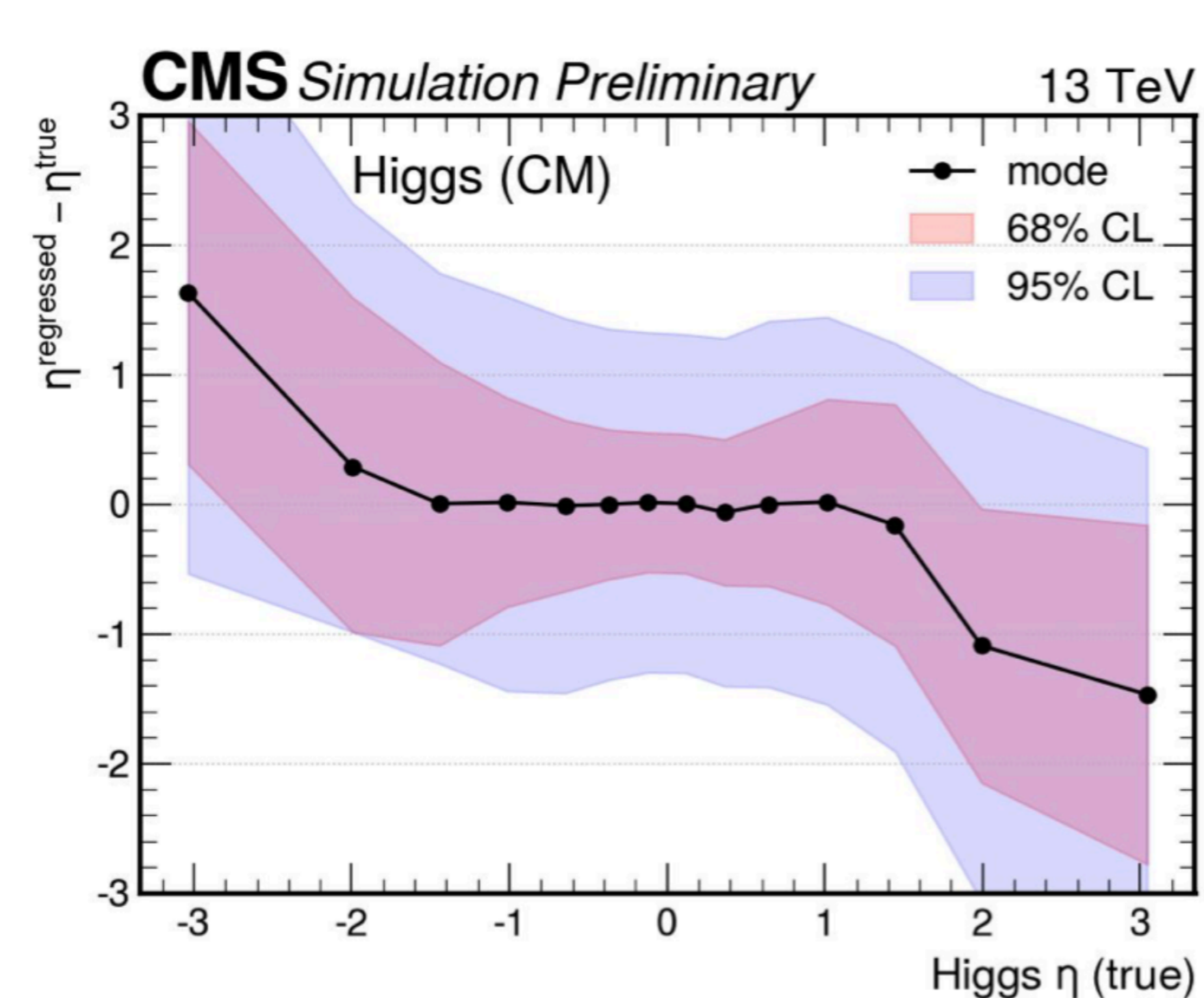


- The main building block is the **Transformer Encoder**
- The training data → partons or reco-objects in the laboratory frame
- The model was pretrained using modified differential multiplier method (MDMM):

- **Main loss L_0** is the Huber loss for partons and boost pz
- **Second loss L_1** is the maximum mean discrepancy (MMD) loss → keep distributions coherent
- **MDMM**: minimize L_0 ensuring $L_1 \leq \epsilon$



Bias mode of the regression for Higgs p_T

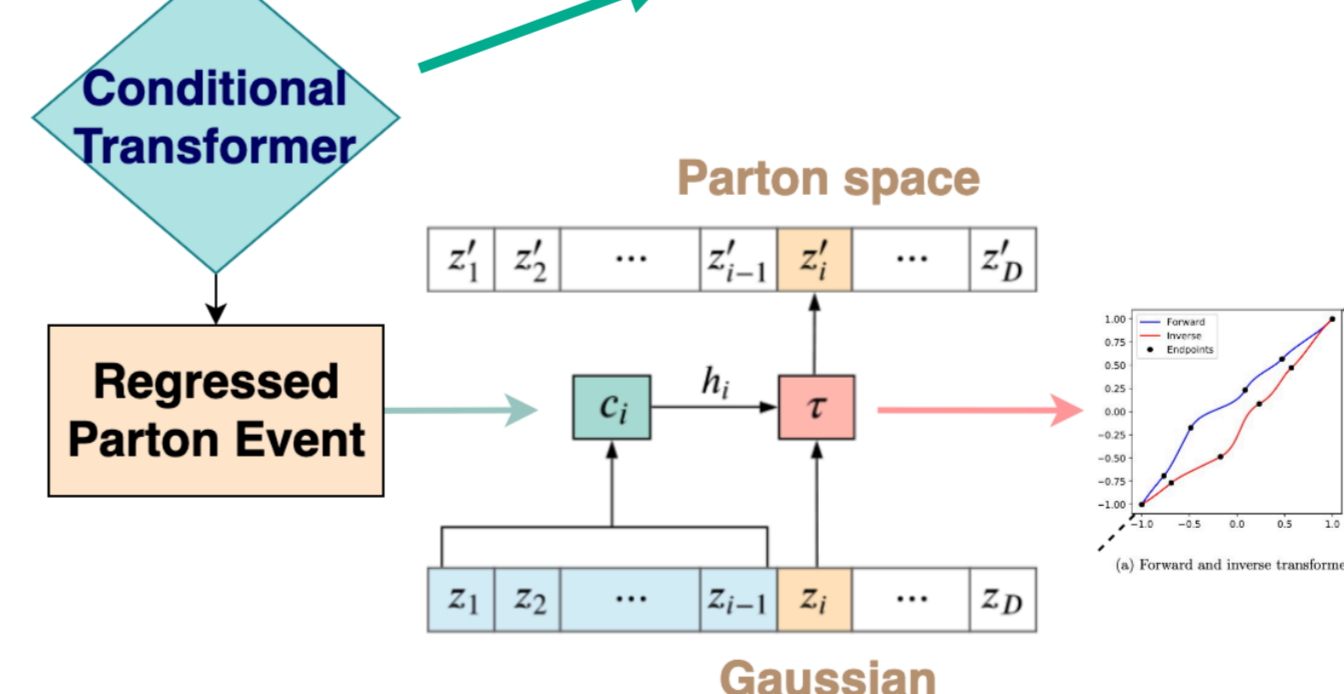


Bias mode of the regression for Higgs η

Unfolding Flow

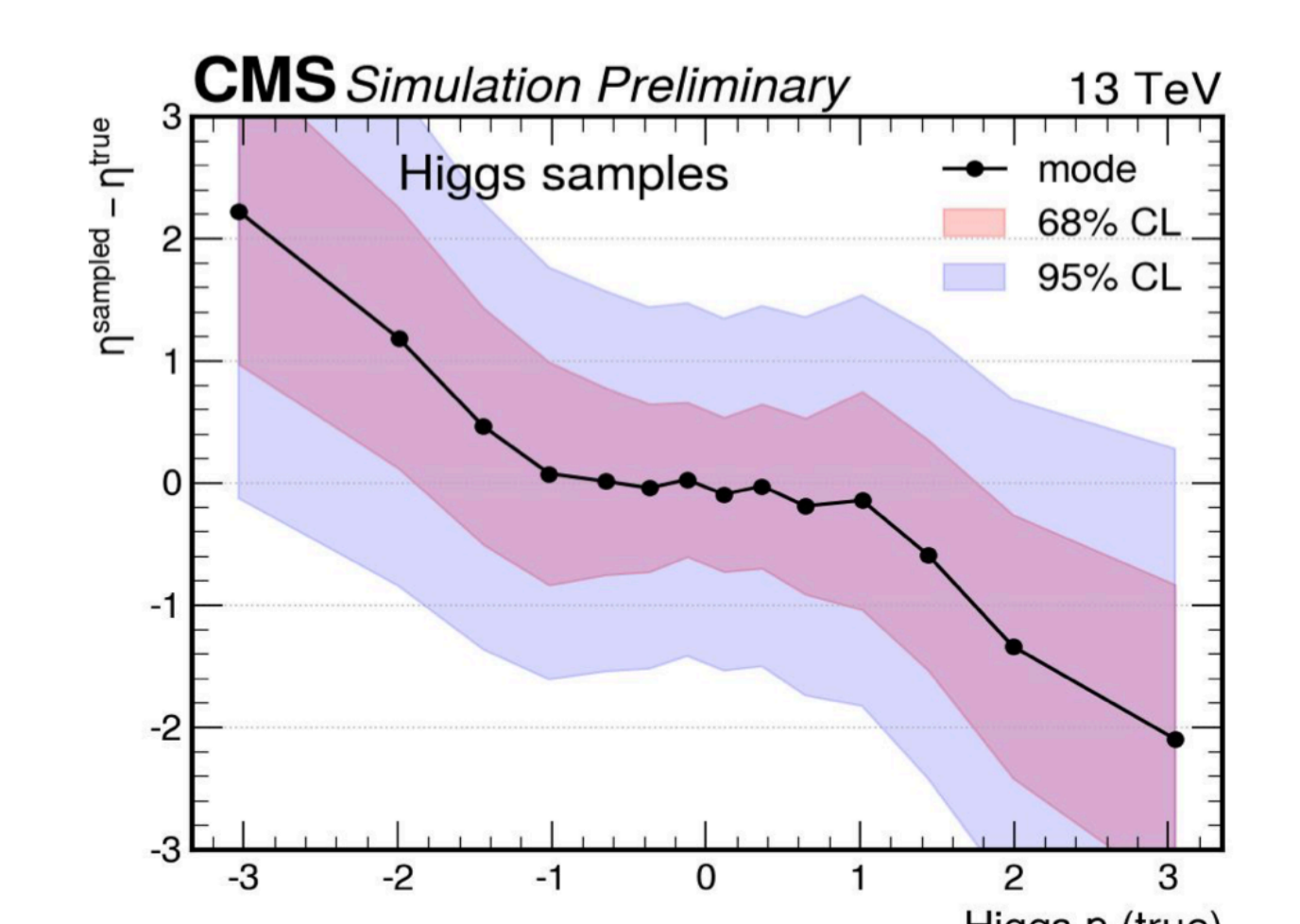
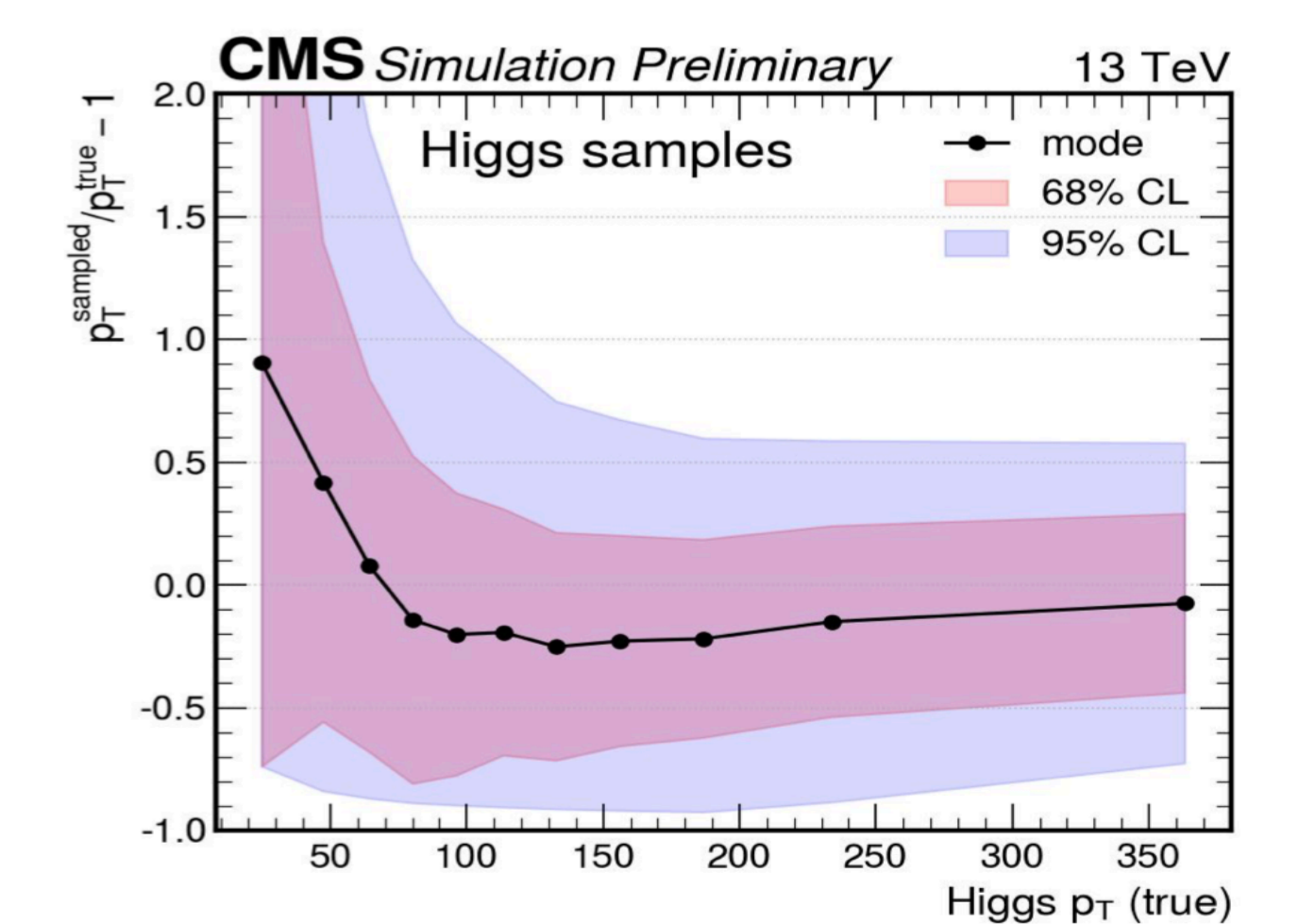
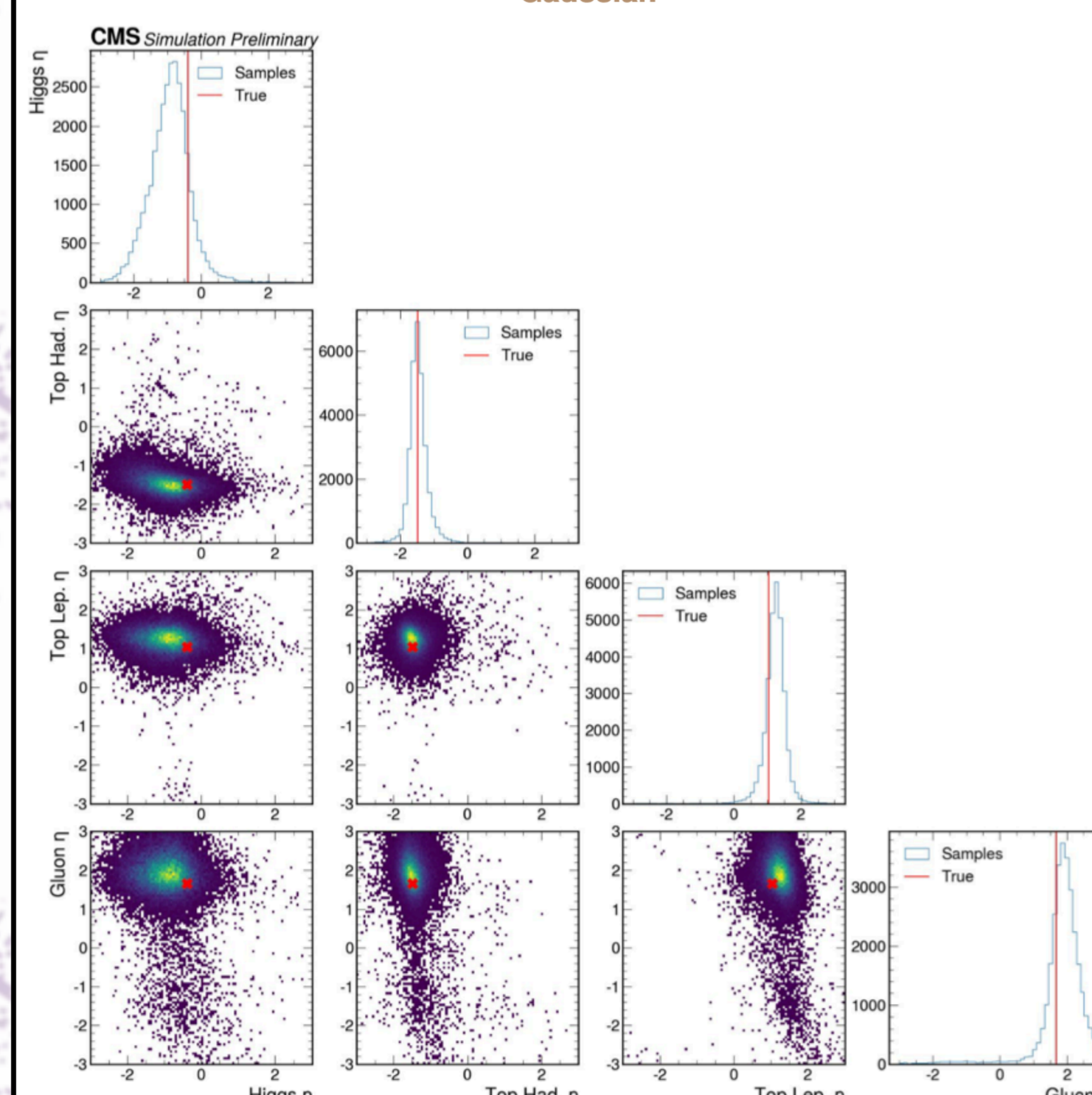


Implemented using **rational quadratic splines (RQS)** with **autoregressive blocks**
weights still updated during **Unfolding Flow** training



Training:

- **maximum likelihood**: evaluate the density of the true partons from the signal MC
- **sampling parton-level events** from the flow and comparing them with the target



Sample 20k parton-level events for one reco-event
Super fast (less than 1 second)

Sample 30 parton events for 1.5M reco-events
Check the quality of the sampled partons

