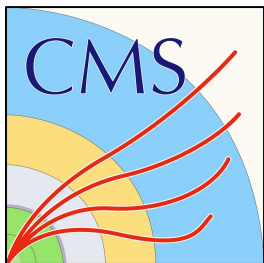


FlashSim: a ML simulation framework



Istituto Nazionale di Fisica Nucleare



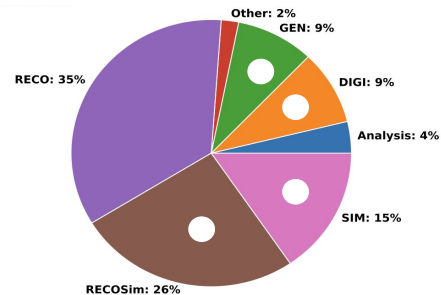
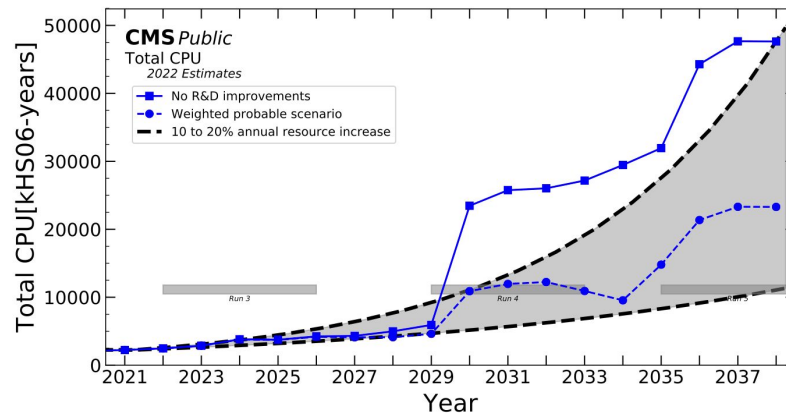
Francesco Vaselli, Filippo Cattafesta, Gloria Cicconofri and Andrea Rizzi
On behalf of the **CMS Collaboration**

Faster simulation frameworks: Motivation

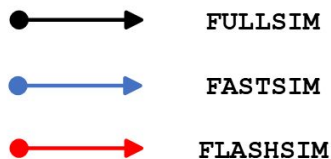
Event simulation is a non-negligible fraction of the total projected CPU need

Faster simulation frameworks are a part of the solution to the computing challenges posed by the HL-LHC era

Machine learning is expected to provide both the speed and the accuracy we need

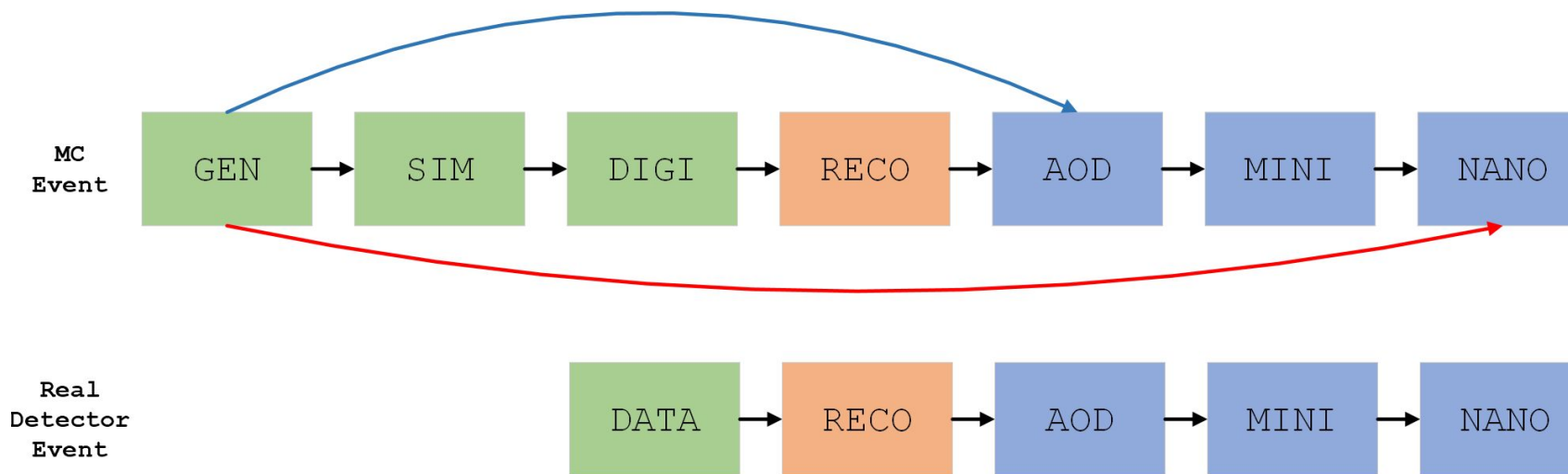


FlashSim means skipping all intermediate steps



We want something:

- fast(er):
Reached 100 Hz/kHz!
- Not analysis specific
- Depending on Gen (not just a *generic* event but *the* event)

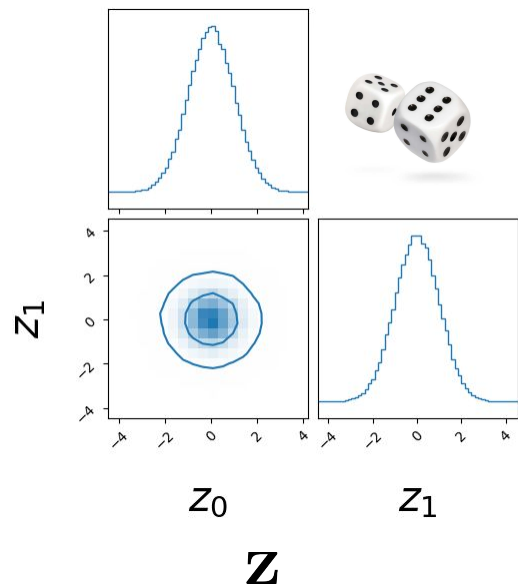


Outline

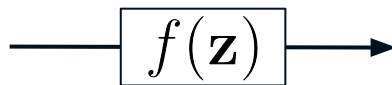
- Building blocks: *Normalizing Flows*
- Objects results:
 - AK4/AK8 Jets
 - Muons
 - Electrons
 - Fakes Jets
- Analysis test
- Fast Gen Variations
- Conclusions

Normalizing Flows: generative model for *pdfs*!

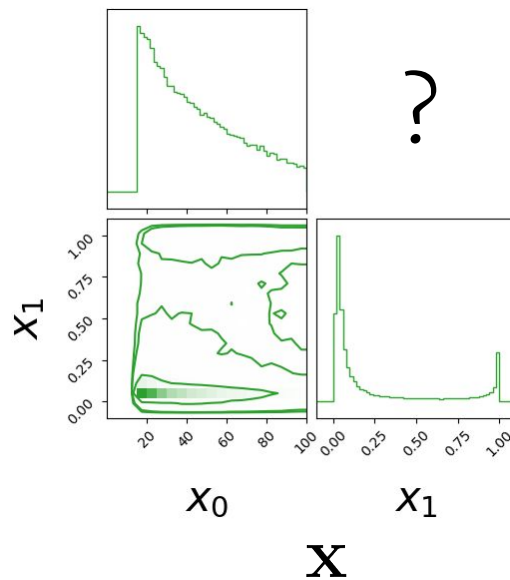
What we know



multi-dimensional gaussian

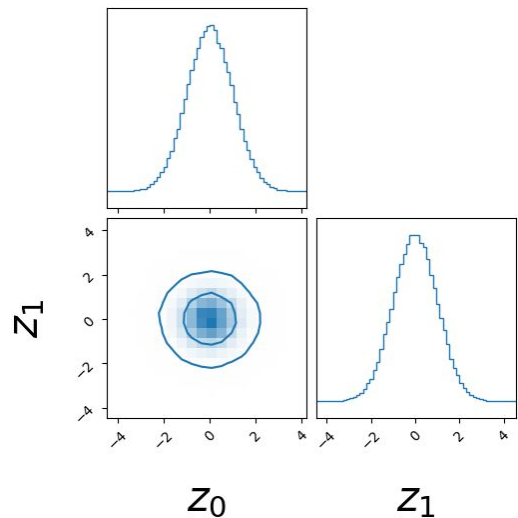


What we need

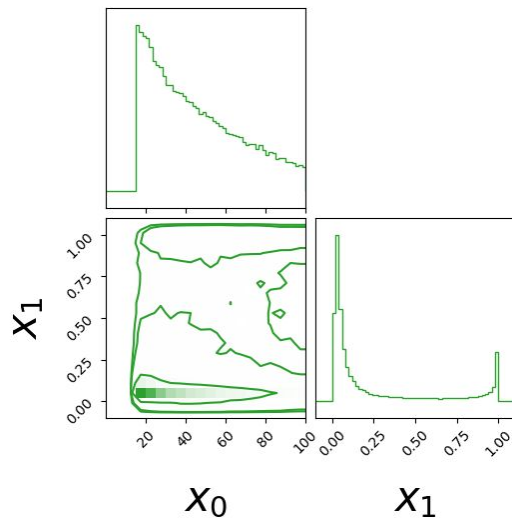


FullSim data, pdf unknown!

Normalizing Flows: generative model for *pdfs*!



$$\leftarrow \boxed{f^{-1}(\mathbf{x})}$$



$$\mathbf{x} = f(\mathbf{z})$$

$$p_x(\mathbf{x}) = p_z(\mathbf{z}) \det \left| \frac{d\mathbf{z}}{d\mathbf{x}} \right|$$

$$\log(p_x(\mathbf{x})) = \log(p_z(f^{-1}(\mathbf{x}))) + \log(\det \mathbb{J}_{f^{-1}}(\mathbf{x}))$$

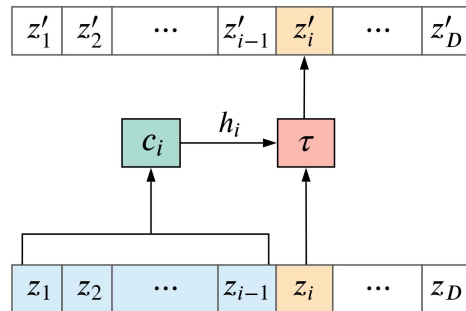
Flows building blocks: *conditioners*

How do we treat the input variables?
Want to correctly model correlations

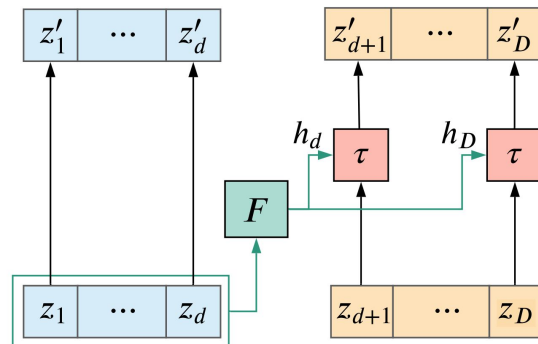
How do we use the other target variables
while transforming another?

Various way to do it!

Autoregressive:



Coupling:



Flows building blocks: *transforms*

Affine:

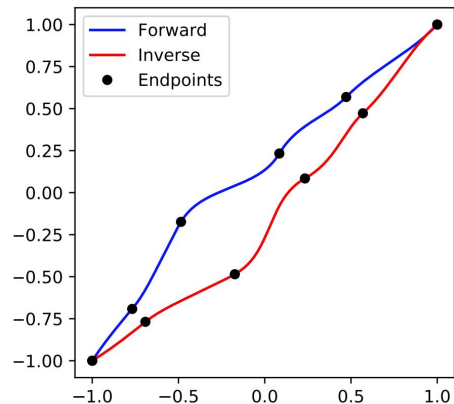
How do we transform the variables?
Various ways to do it (as long as the transformation is invertible!)

Each model is made up of multiple coupling+transformation blocks

This gives us an expressive final transformation with good correlations between variables

$$\tau(z_i; \mathbf{h}_i) = \alpha_i z_i + \beta_i$$

Splines:



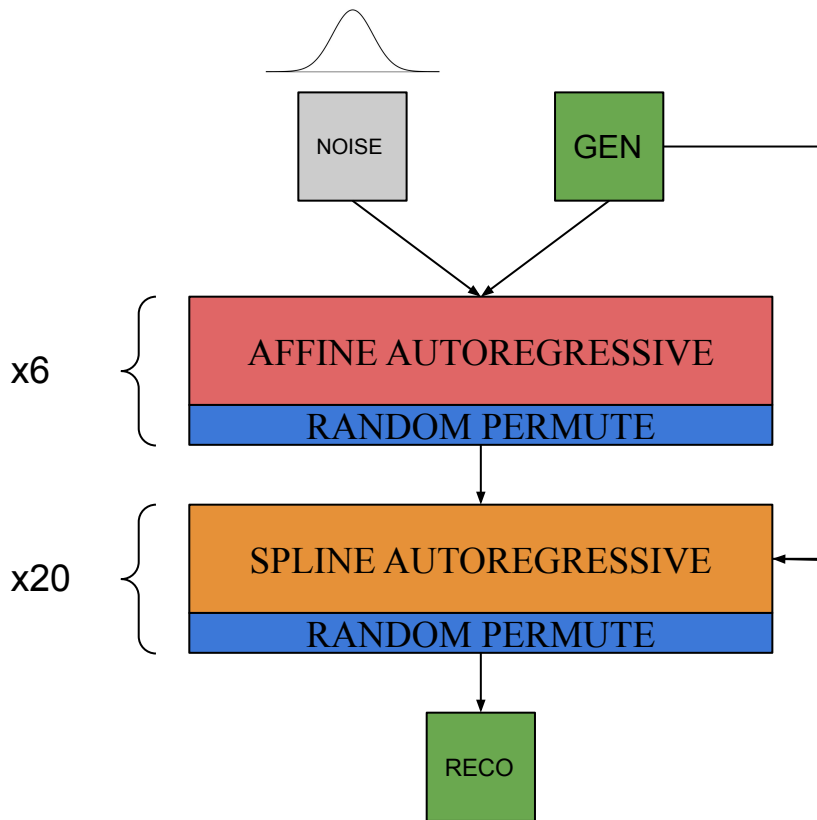
A *mixture* model as our baseline

Conditioning gen variables: 38
Target variables: 47

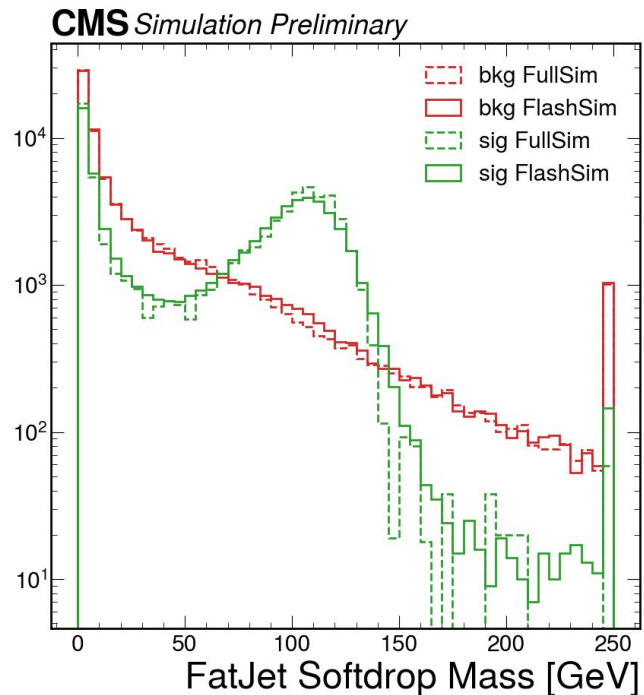
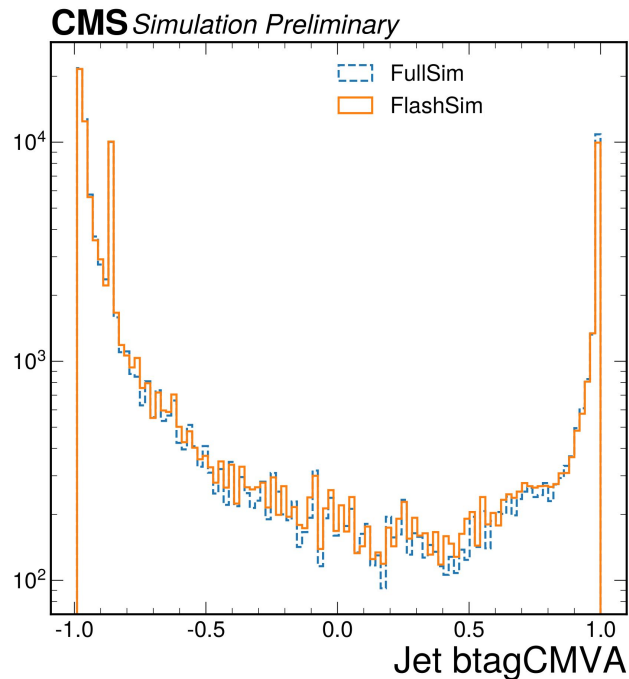
Total params: $O(10e6)$

Trained on $5e6$ objects from ttbar events

Newer model are based only on affine layers, good results with smaller dims



AK4 and AK8 Generator-matched Jets: good closure on 1d distributions

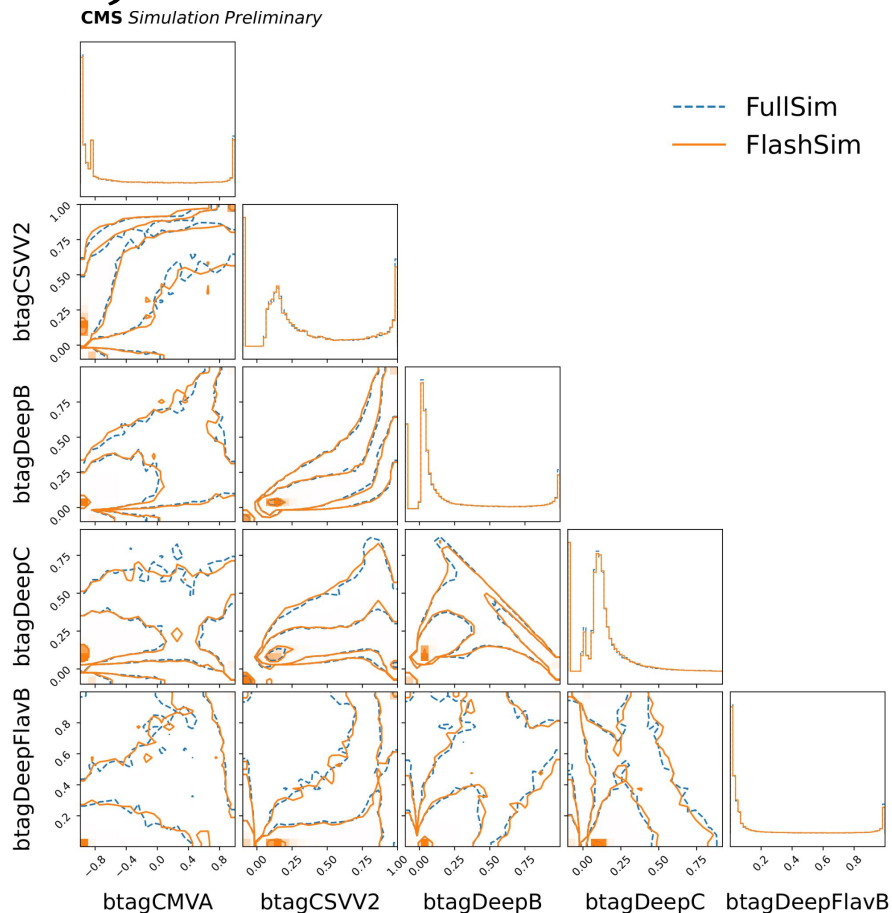


AK₄ Generator-matched Jets: accurate correlations

The corner plot shows
multivariate tagging
algorithms

B-tag, C-tag algorithms
relying on similar jet
features

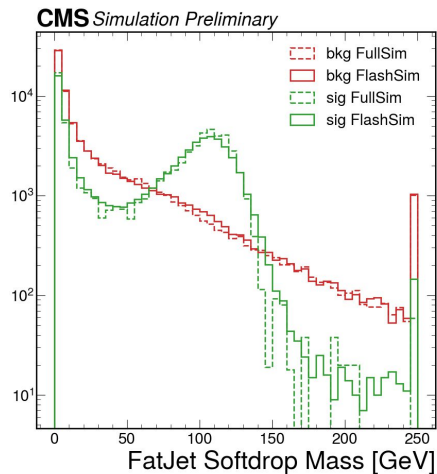
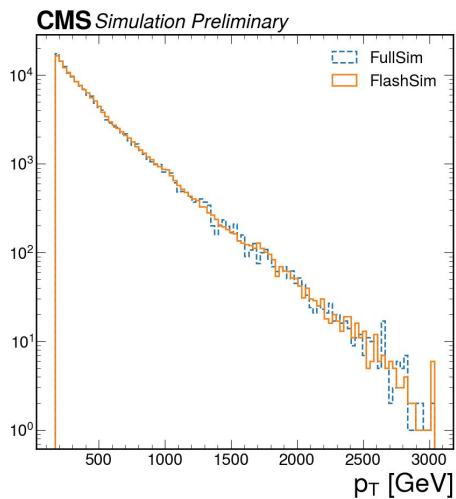
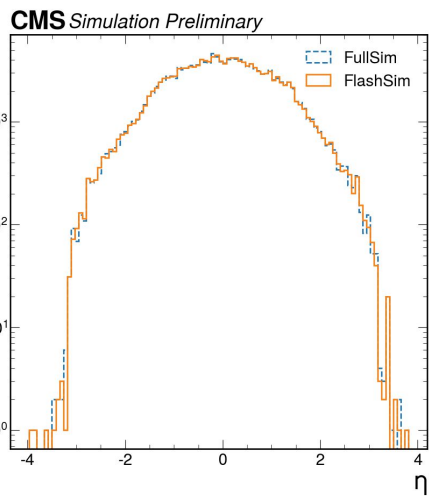
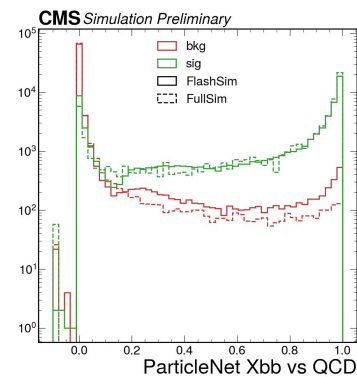
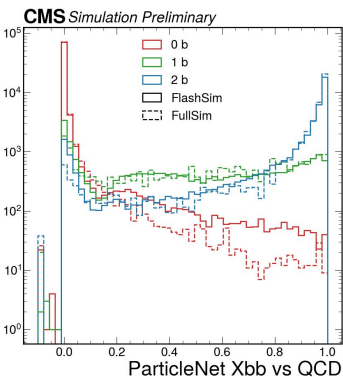
Even without specific
information,
FlashSim learns them!



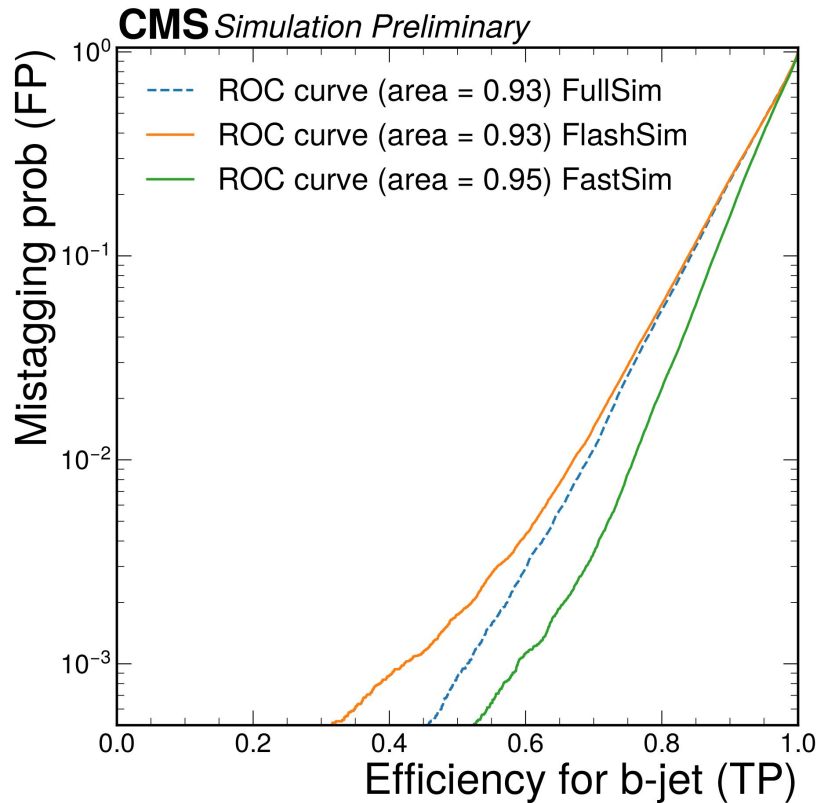
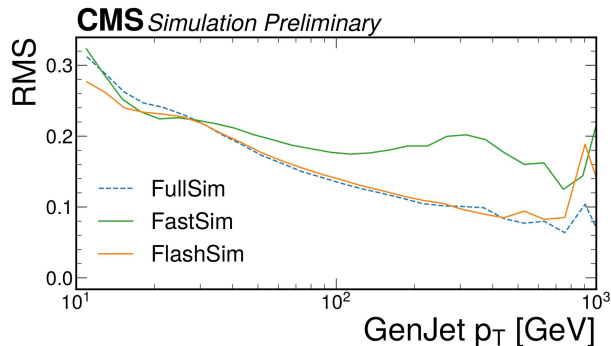
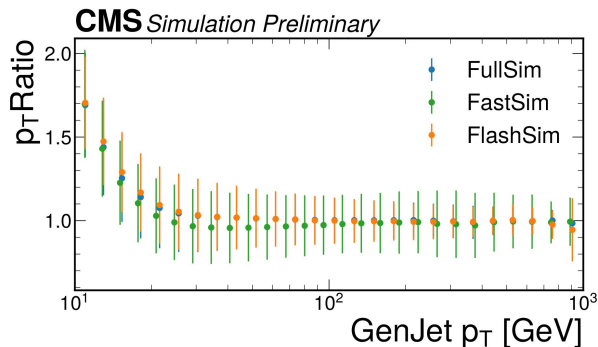
FatJets results

Trained of Phase 2 QCD multijet events and $HH \rightarrow 4b$

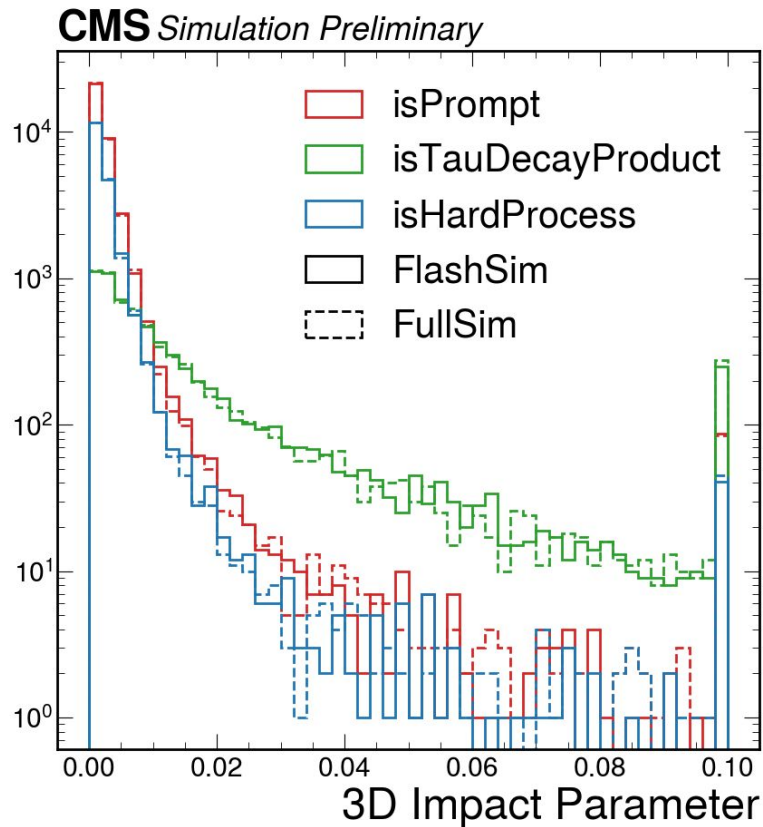
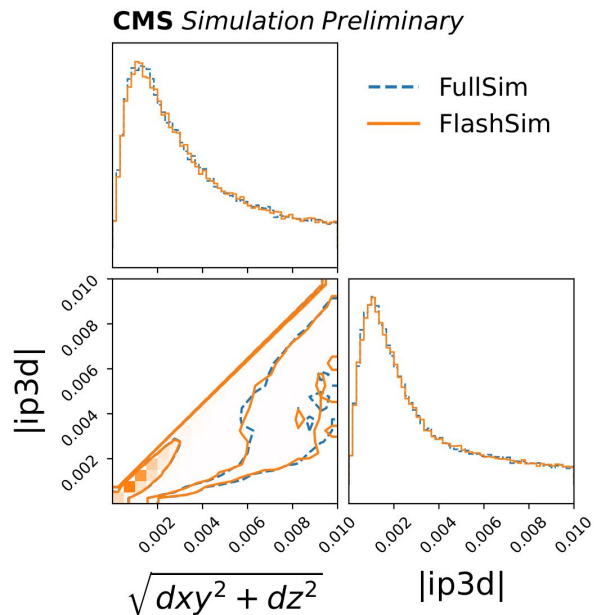
Smaller training sample compared to Run II samples



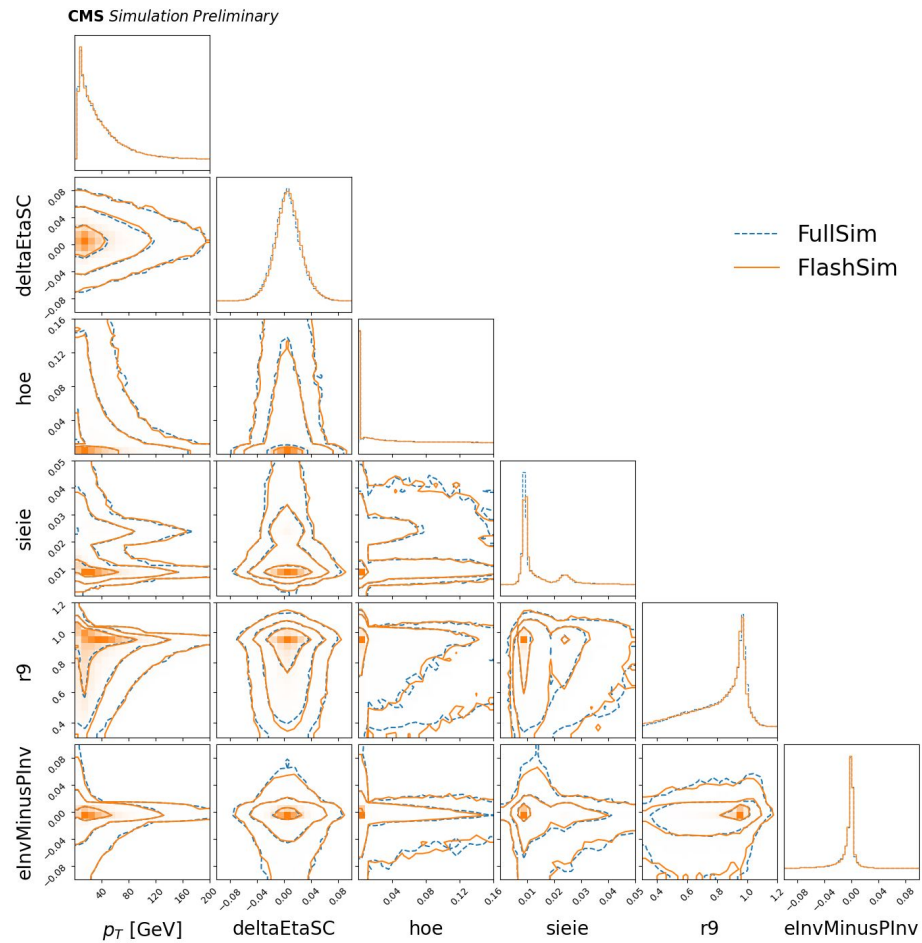
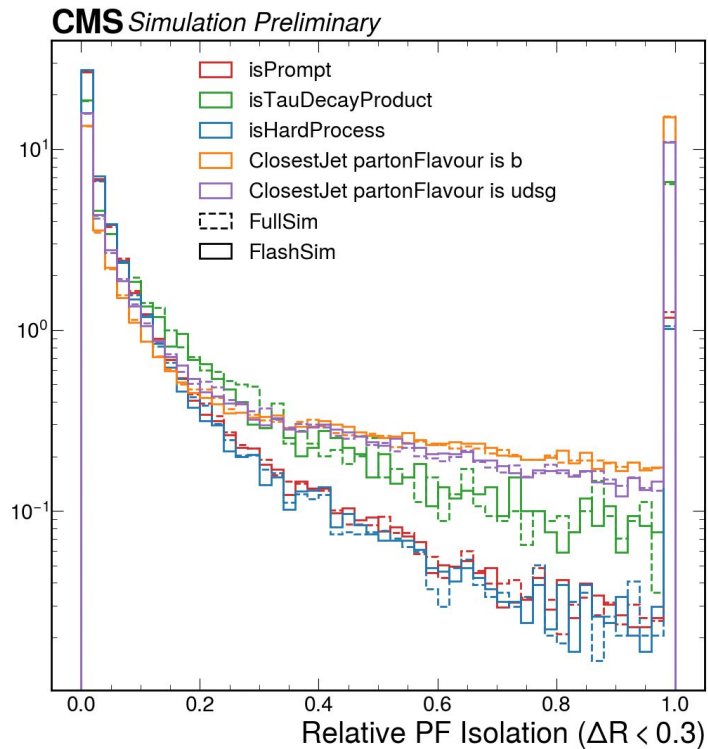
AK₄ Generator-matched Jets *conditioning*



Promising results on Muons



Promising results on Electrons

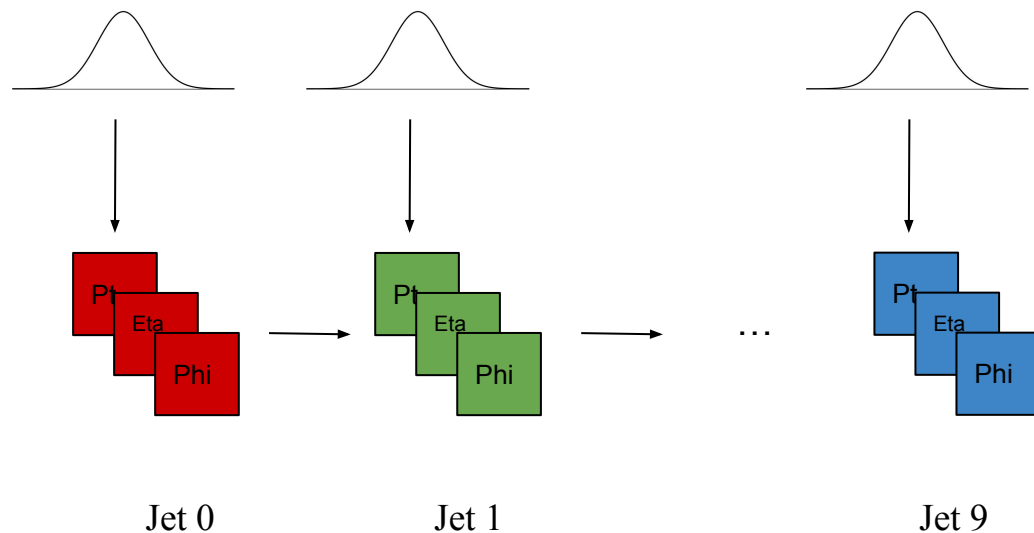


Fakes are a different type of problem: need to experiment

For fake jets there is no “generator” starting point to use!

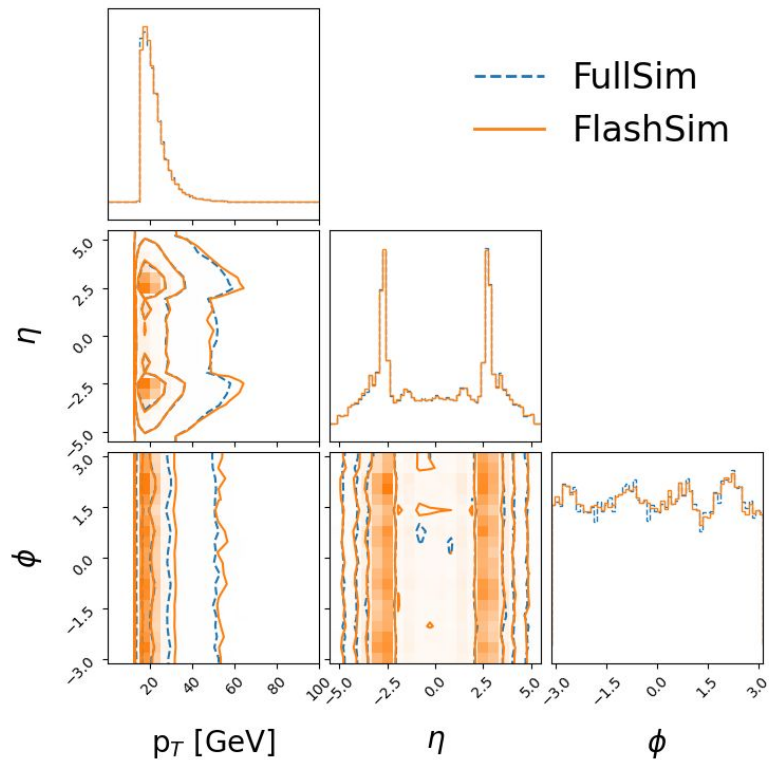
Autoregressive chain on a variable number of fake jets per event

(pT, eta, phi) of jet N depends on those of jets N-1

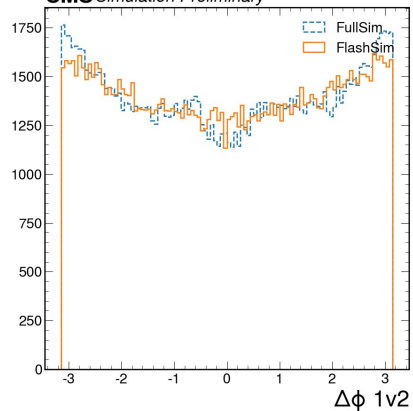


Fake jets results

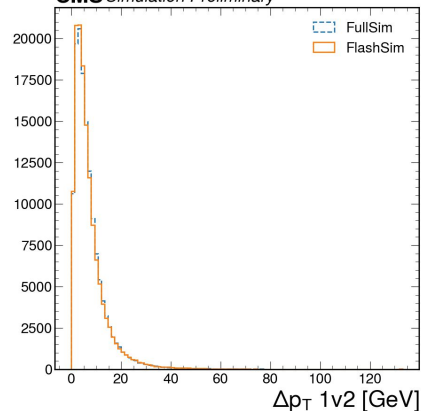
CMS Simulation Preliminary



CMS Simulation Preliminary



CMS Simulation Preliminary



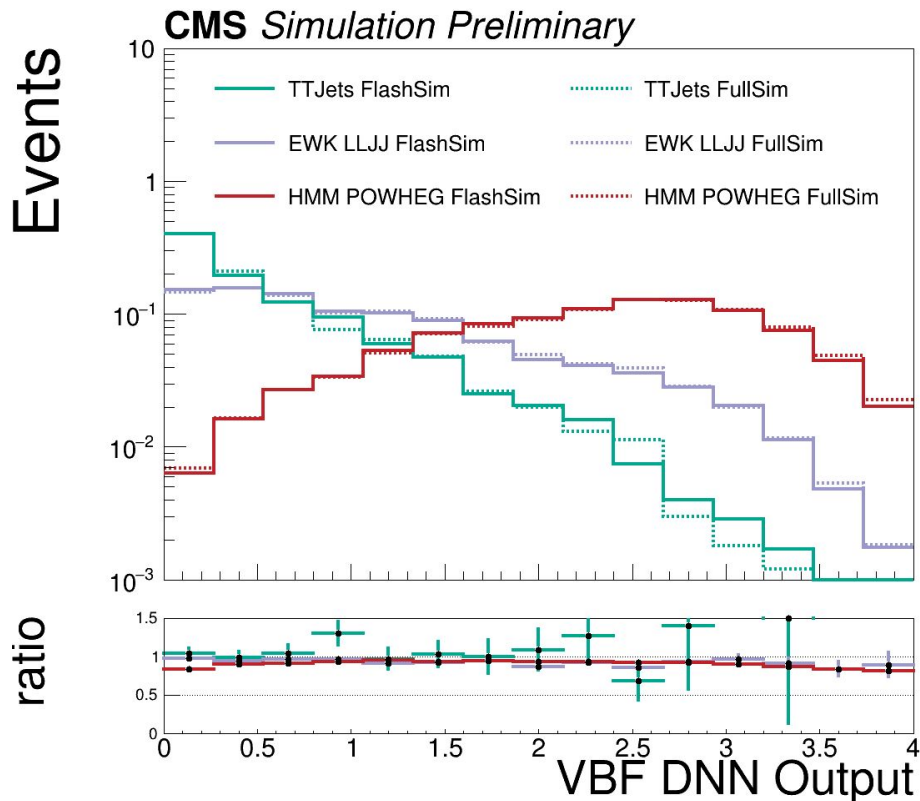
A real-world scenario: Analysis test

Simulate Jets, Muons on 4
different processes:
TTJets, DY+Jets, EWK LLJJ,
 $H \rightarrow \mu^+ \mu^-$

Perform VBF $H \rightarrow \mu^+ \mu^-$
analysis selection

Compute derived
quantities

Compute the DNN score
used in the final fit!

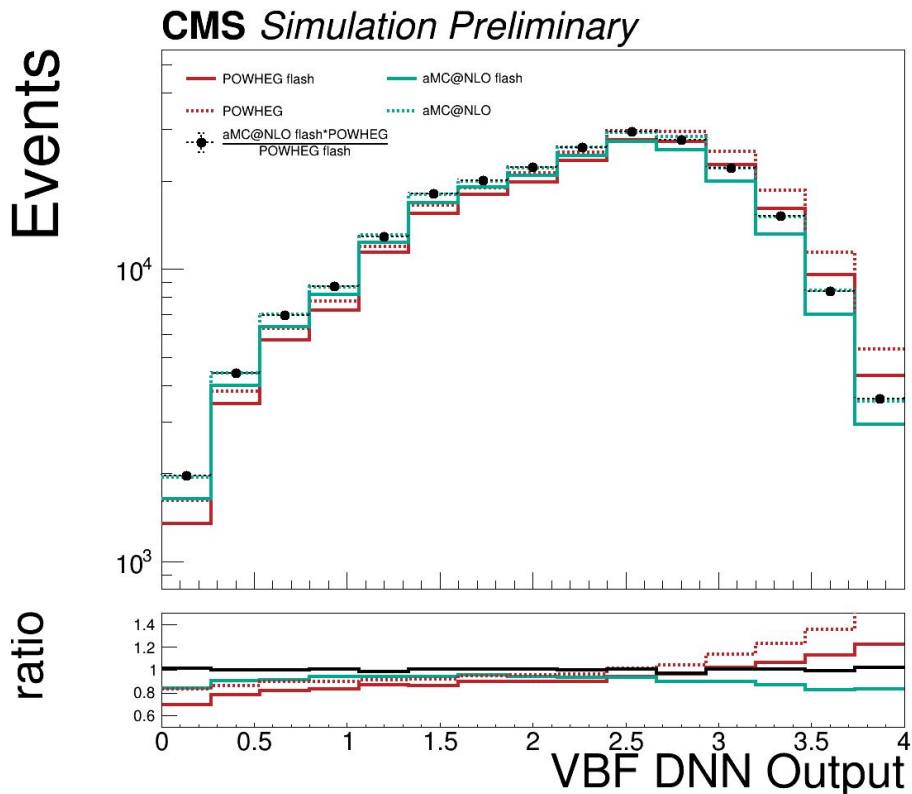


FlashSim for Generator-level variations

Given a FullSim sample with generator e.g. POWHEG
We can compute:

$$VAR_{Full} = REF_{Full} \frac{VAR_{Flash}}{REF_{Flash}}$$

To get quickly another precise sample for a different generator such as aMC@NLO



Conclusions

We compared results with Run II FullSim/ Phase-2 samples for:

- AK4/AK8 Jets
- Muons
- Electrons
- Fakes Jets
- Analysis use case

Still need to:

- R&D base models
- Cover missing Objects
- Crack Fakes

So far, **a fully-ML powered FlashSim simulation frameworks seems achievable!**
Speed of kHz/100Hz

A Public CMS note is available with all the results: <https://cds.cern.ch/record/2858890>

First concept repo:

<https://github.com/francesco-vaselli/FlashSim>

Get in touch!

francesco.vaselli@cern.ch



Backup

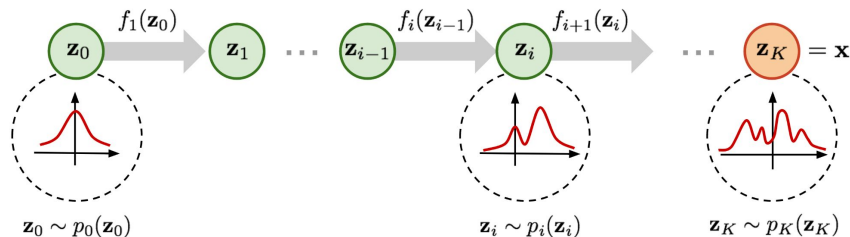
- Normalizing Flows in detail

Normalizing Flows in detail

Learn the $f(z)$ to send $p_{\mathbf{z}}(\mathbf{z})$ into the (*unknown*) data distribution $p_{\mathbf{x}}(\mathbf{x})$

Needs:

- Basic distribution $p_{\mathbf{z}}(\mathbf{z})$, typically Gaussian
- Function called flow $f(z)$ invertible and differentiable, with tractable jacobian (splines are a good choice)



$$p_x(\mathbf{x}) = p_z(f^{-1}(\mathbf{x})) \det \left| \frac{d\mathbf{z}}{d\mathbf{x}} \right|$$

$$\log(p_x(x)) = \log(p_z(f^{-1}(\mathbf{x}))) + \log(\det \mathbb{J}_{f^{-1}}(\mathbf{x}))$$