

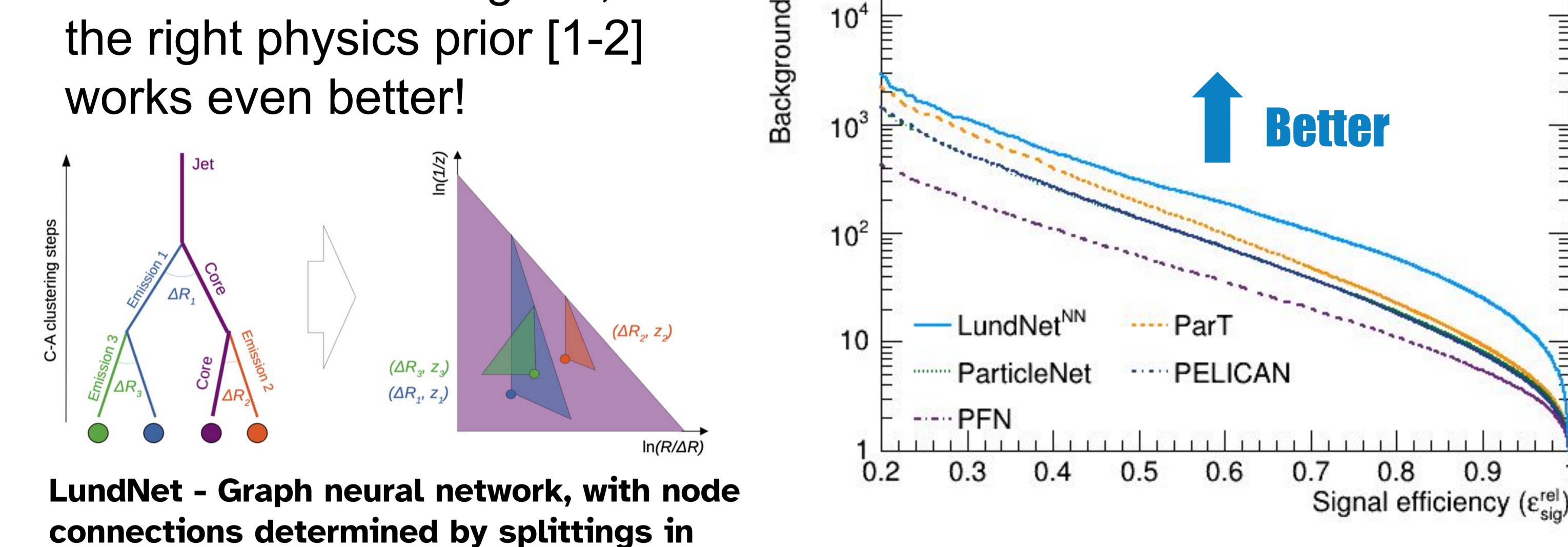
Performance versus uncertainty in boosted top tagging with the ATLAS detector Kevin Greif, for the ATLAS Collaboration. BOOST, July 29th 2024 kgreif@uci.edu

Performance

What is the most effective machine learning model for tagging boosted top quarks?
Transformers work great, but

Lund jet plane and node features:

ε_{bkg}^{rel}	10 ⁶ ATLAS Simulation Preliminary	
(1/	$= \sqrt{s} = 13$ TeV, <i>top</i> tagging	-
ction	10 ⁵ anti-k, R=1.0 UFO Soft-Drop CS+SK jets	
ece	Ē p _τ > 350 GeV, η < 2.0	E
р	4 0 4	200
-		



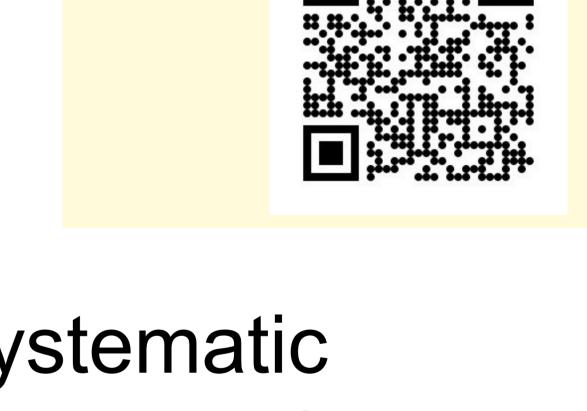
Public data:

Versus

Have an idea for controlling uncertainties? These datasets are public!

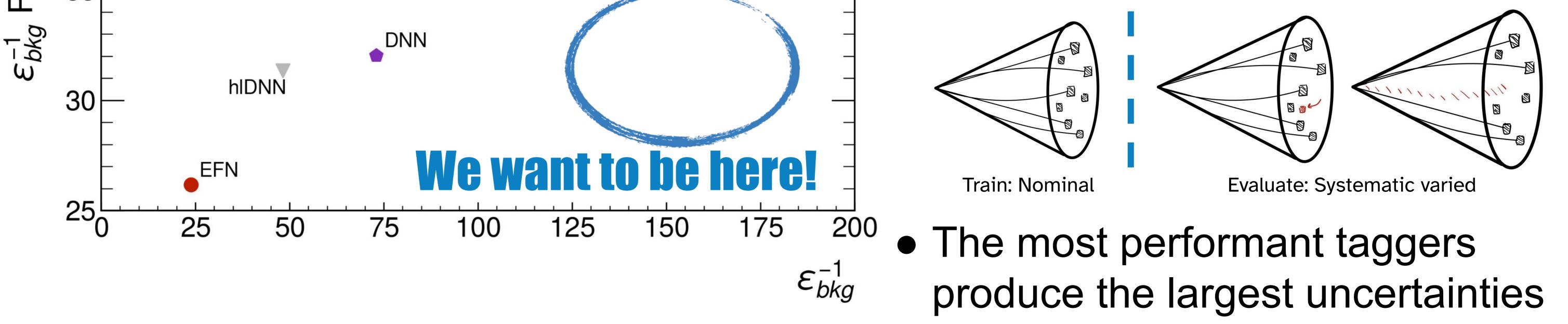
	50]
ative Uncertainty [%]		ATLAS Simulation		_
		\sqrt{s} = 13 TeV, Pythia8		
	45	anti-k _t R = 1.0 UFO CSSK Jets		_
		$\varepsilon_{sig} = 0.5$	ParticleNet	_
	40			_
		ResNet50		
		PFN		
Rel	35			

 $= k_t, \Delta, z, m, \psi$



Treatment of systematic uncertainties is crucial for interpreting physics results
Can approximate the size of systematic uncertainties associated with each tagger using "bottom-up" approach [3]:

uncertainty



References: [1] F. A. Dreyer and H. Qu, *Jet tagging in the Lund plane with graph networks*, JHEP **03** (2021) 052, arXiv: 2012.08526 [hep-ph] [2] ATLAS Collaboration, *Tagging boosted W bosons applying machine learning to the Lund Jet Plane*, ATL-PHYS-PUB-2023-017, 2023 [3] CERN preprint: CERN-EP-2024-159