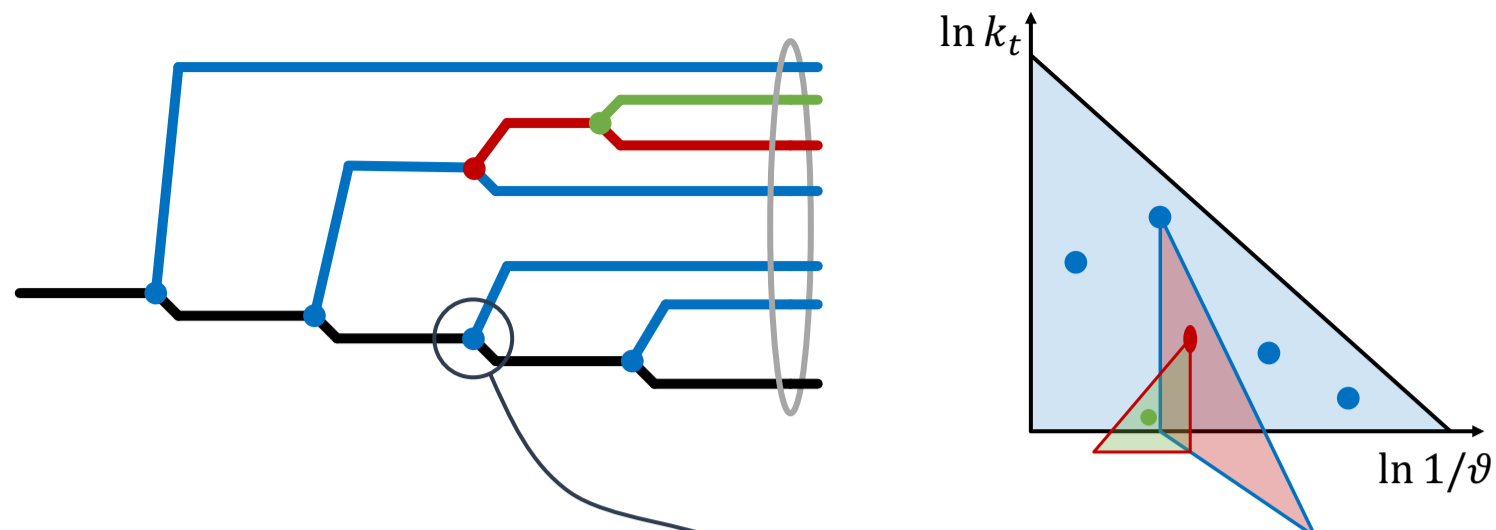


Adam Takacs* University of Bergen (Norway)
 Frederic Dreyer University of Oxford (UK)
 Gregory Soyez IPhT, CNRS, CEA Saclay (France)
 Based on [arXiv:2112.09140](https://arxiv.org/abs/2112.09140)



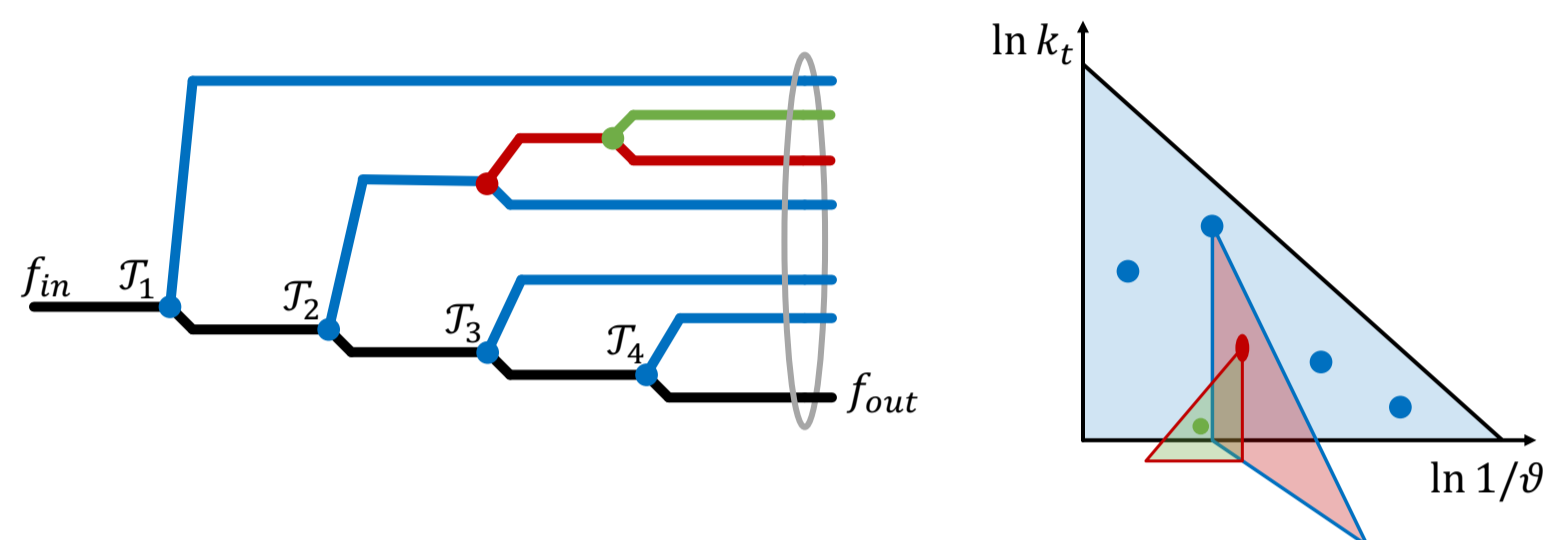
Cambridge/Aachen reclustering (angular ordering)



One step in the declustering: $\mathcal{T}_i = \{\Delta_i, k_{t,i}, z_i, \dots\}$

2

Primary Lund declusterings



Primary declusterings: $\mathcal{L}_{prim} = [\mathcal{T}_1, \dots, \mathcal{T}_n]$

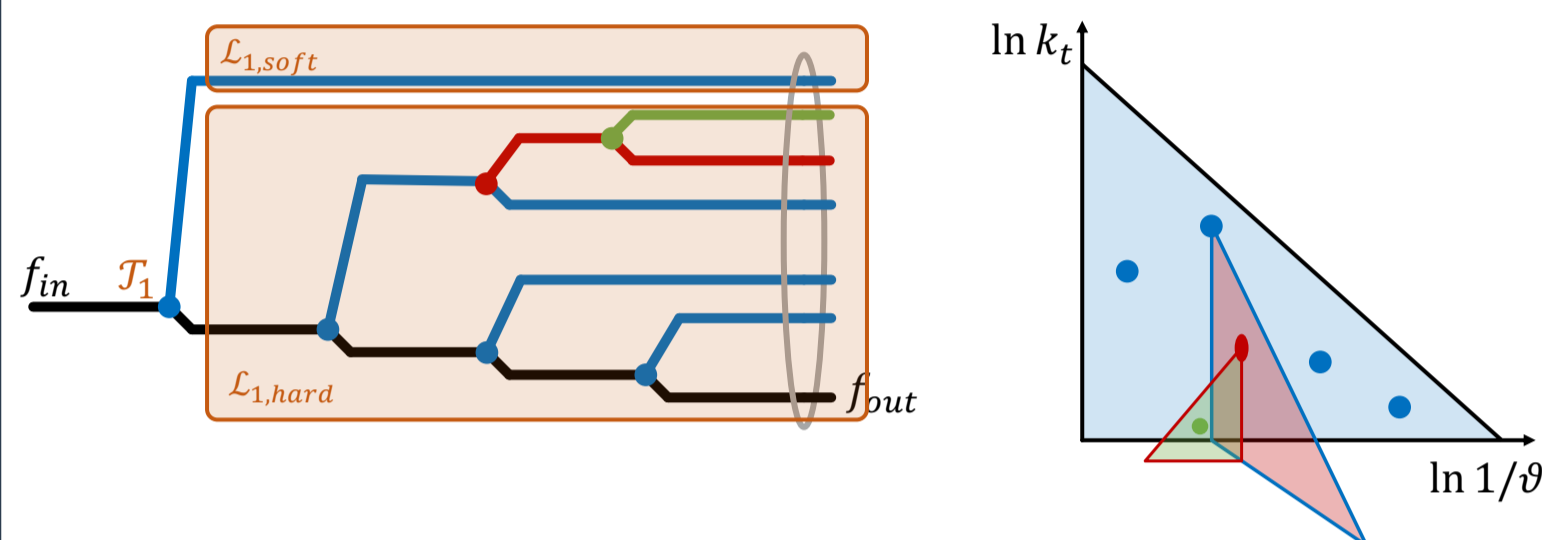
Probability of \mathcal{L}_{prim} :

$$p(\mathcal{L}_{prim}, f_{out} | f_{in}) = S^{(n+1,n)} P^{(n)} S^{(n,n-1)} \dots P^{(1)} S^{(1,0)}$$

no emission
splitting

3

The Lund tree



Full tree (recursively):

$$\mathcal{L}_{tree}(j) = [\mathcal{T}_j, \mathcal{L}_{j,hard}, \mathcal{L}_{j,soft}]$$

4

The likelihood ratio

Optimal discriminants at NLL:

$$\mathbb{L}_{prim} = \frac{p_q(\mathcal{L}_{prim})}{p_g(\mathcal{L}_{prim})}$$

$$\mathbb{L}_{tree} = \frac{p_q(\mathcal{L}_{tree})}{p_g(\mathcal{L}_{tree})}$$

Ingredients to NLL:

- changing flavor
- collinear-hard emissions
- running coupling
- commensurate angle emissions

[A.Lifson,G.Salam,G.Soyez,2007.06578]

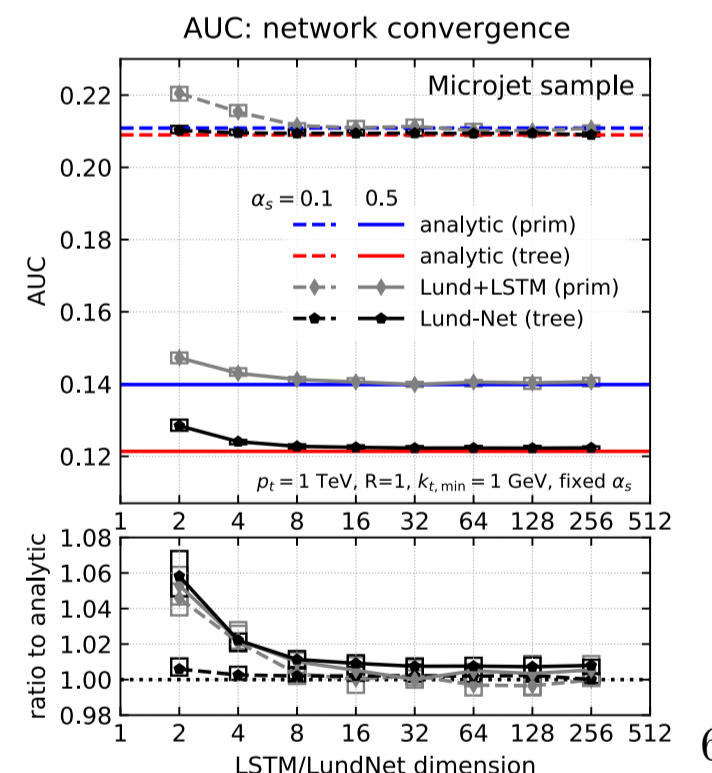
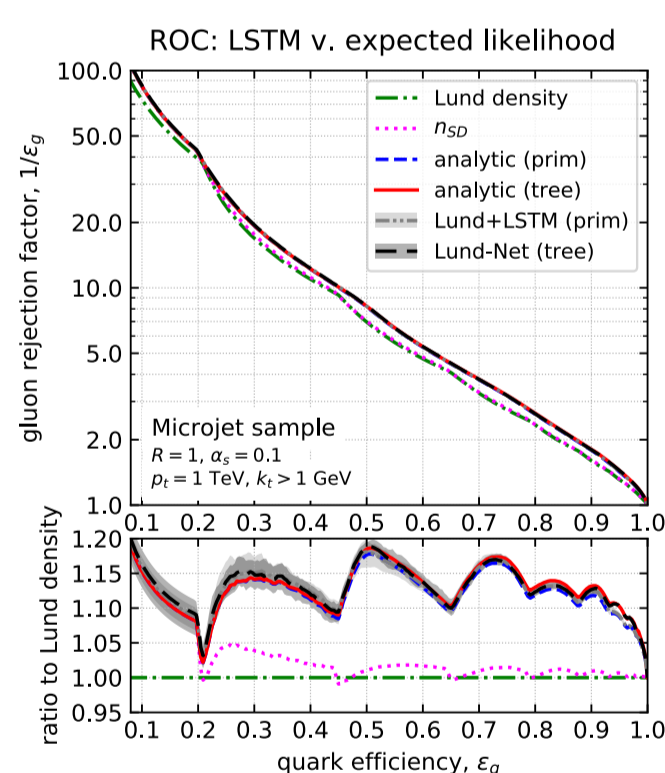
Machine Learning techniques:

1. LSTM network (\mathcal{L}_{prim}) [F.Dreyer,G.Salam,G.Soyez,1807.04758]
2. Lund-net graph network (\mathcal{L}_{tree}) [F.Dreyer,H.Qu,2012.08526]

5

Equivalence of ML and analytic

Collinear limit \rightarrow ML = analytic!

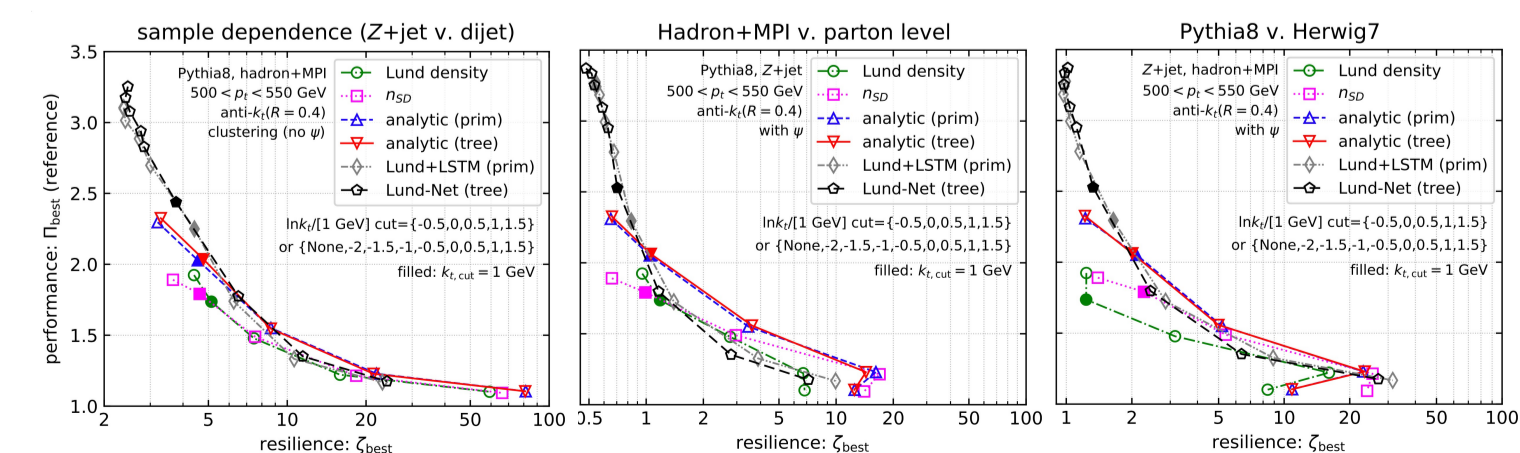


6

Equivalence of ML and NLL

Studies:

- sensitivity to the parton shower (Pythia8 vs. Herwig7)
- sensitivity to the hard process (dijet vs. Z+jet)
- sensitivity to non-perturbative effects (MPI, hadronization)

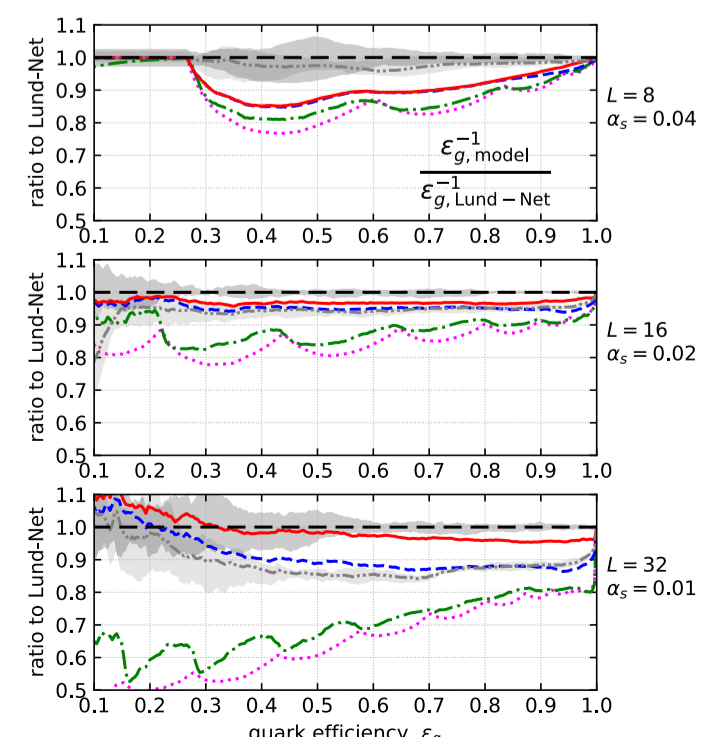
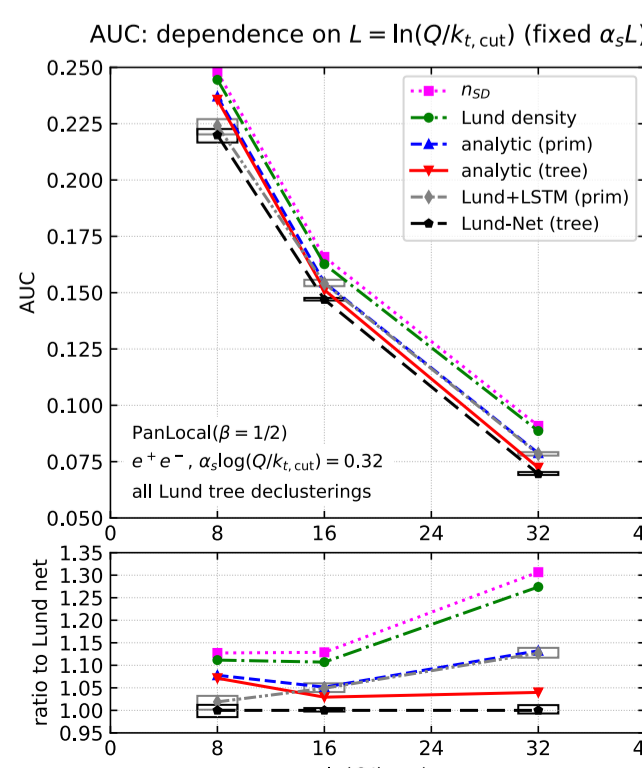


Lund-ML > Lund analytic > ISD
 tree > primary > ISD

7

Equivalence of ML and NLL

PanScales NLL parton shower: ML \rightarrow NLL [2002.11114]



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