

DNN normalization in DLKit study

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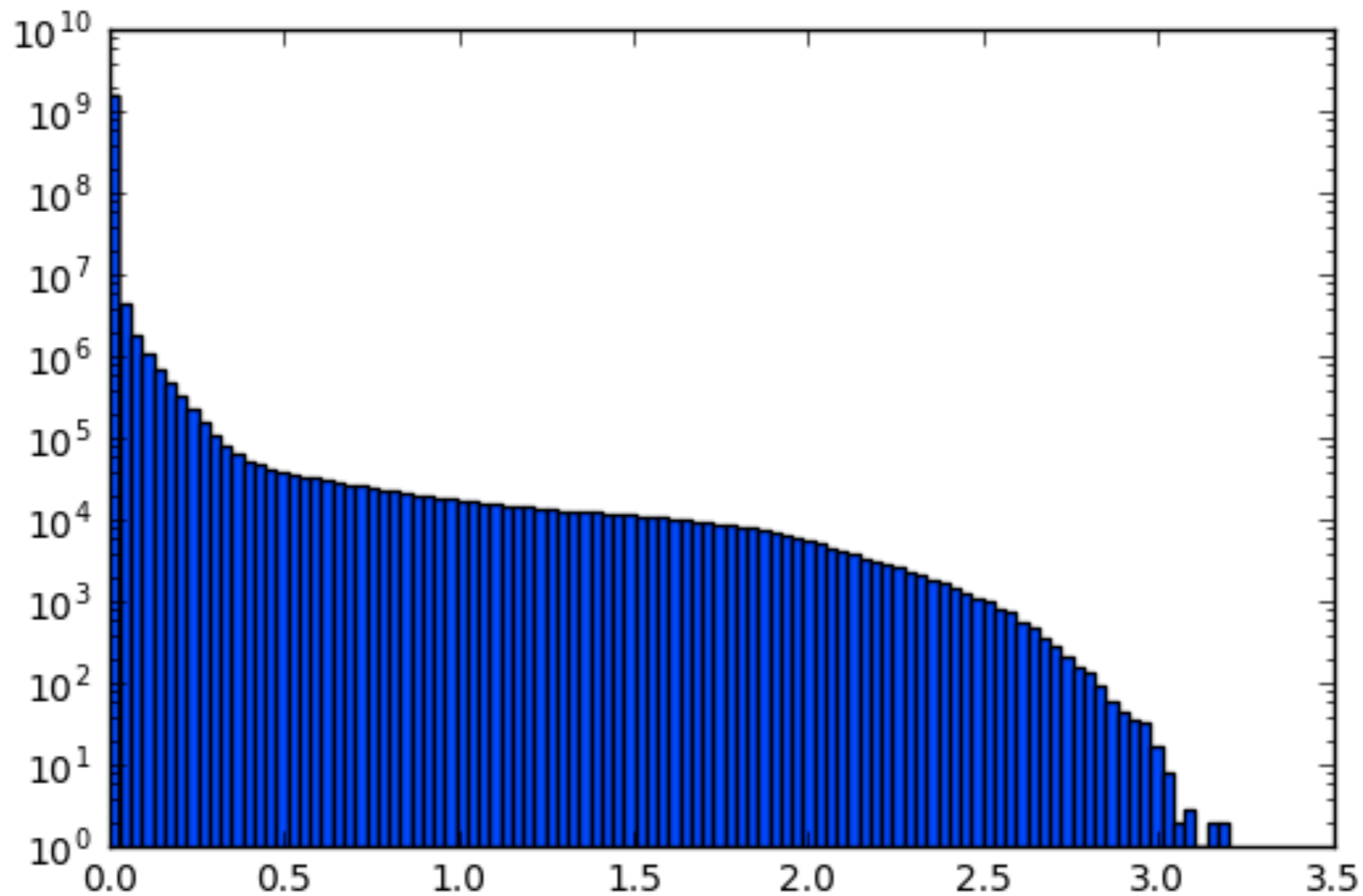
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with **Amir Farbin (UTA)**



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ECAL x-data



Maximum 3.2 with our old normalization factor of 1/150.

New nonlinear normalization

- I setup a hyper-parameter scan over width, depth, and the ECAL and HCAL normalization types, trying the
 - 1/150 past default
 - 1/480 (1/3.2 * past default)
 - “nonlinear” function suggested by Amir below:

$$X' = \tanh(\text{sign}(X) \ln(\text{abs}(X) + 1) / 2)$$

Scan results

```
In [7]: # Make a Table of all relevant parameters, sort by 1,2,then 0 columns.
# Note: Parameters are optional... but the columns and rows will be not optimally sorted.
ScanTable(MyModels,['Model Name', 'Width', 'Depth', 'Epochs', 'Ele_AUC', 'Pi0_AUC', 'ChPi_AUC',
```

Model Name	Width	Depth	Epochs	El
e_AUC Pi0_AUC ChPi_AUC Gamma_AUC				
-----	-----	-----	-----	-----
Width=256 Depth=1 ECALNorm=480.4 HCALNorm=1235.5	256	1	18	
0.5000 0.5000 0.5000 0.5000				
...				
Width=32 Depth=4 ECALNorm='nonlinear' HCALNorm='nonlinear'		32	4	10
0.9948 0.9458 0.9986 0.9530				
Width=128 Depth=3 ECALNorm='nonlinear' HCALNorm=150.0		128	3	8
0.9924 0.9466 0.9982 0.9535				
Width=512 Depth=3 ECALNorm='nonlinear' HCALNorm='nonlinear'		512	3	8
0.9952 0.9478 0.9994 0.9527				
Width=64 Depth=4 ECALNorm='nonlinear' HCALNorm='nonlinear'		64	4	12
0.9952 0.9501 0.9989 0.9559				
Width=64 Depth=3 ECALNorm='nonlinear' HCALNorm='nonlinear'		64	3	8
0.9909 0.9504 0.9977 0.9541				
Width=128 Depth=3 ECALNorm='nonlinear' HCALNorm='nonlinear'		128	3	8
0.9953 0.9514 0.9992 0.9566				
Width=512 Depth=4 ECALNorm='nonlinear' HCALNorm=150.0		512	4	6
0.9932 0.9520 0.9980 0.9567				
Width=256 Depth=3 ECALNorm='nonlinear' HCALNorm='nonlinear'		256	3	8
0.9957 0.9530 0.9993 0.9558				

$\approx .996$ $\approx .95$ $\approx .999$ $\approx .96$

The best models tend to use the “nonlinear” normalization.

Thoughts

- Using a nonlinear normalization function should be equivalent to using a particular nonlinear activation function at the first layer.
- So maybe one would want to use the linear l_1 normalization and push the nonlinearities to the activation, but maybe this is a nice combination in the end.
- In CaloDNN/Models.py, the Fully3DImageClassification model we are using uses *relu* normalization for every layer except the last, where it uses *softmax*.