

The Fast Calorimeter Challenge 2022: Results & The Road Ahead

— Villa Mondragone, Frascati, Italy —

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- We now have seen 12 talks with different approaches
- All of them better than midjourney \Rightarrow

\Rightarrow The big Question:
How do they compare to each other?



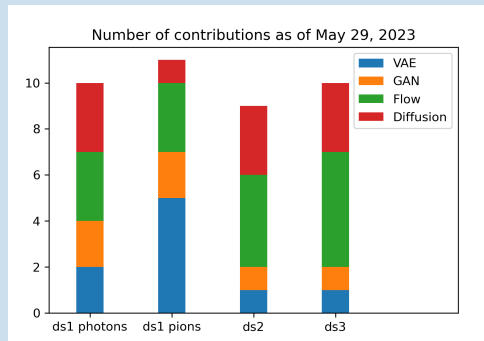
“Calorimeter Simulation”

A total of 40 samples have been submitted

- Until Monday afternoon, 40 samples have been submitted.
- Not enough time to run all comparisons.

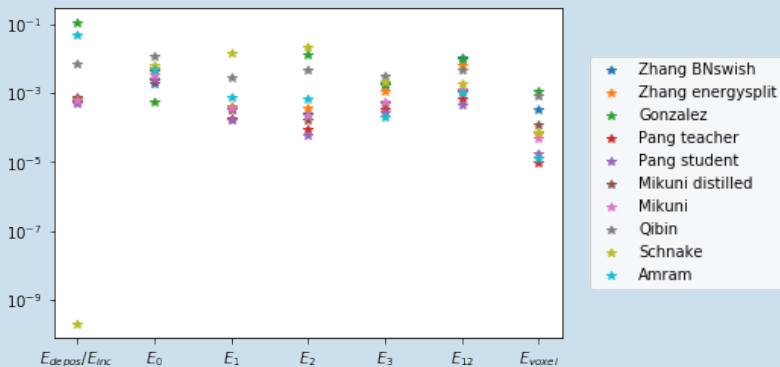
⇒ A few histogram χ^2

⇒ Focus on Multi-class classifier for now:
Train on submission 1 vs. submission 2 vs. ... vs. submission n
Which submission is GEANT4 the closest to?



Histograms ds1 photons

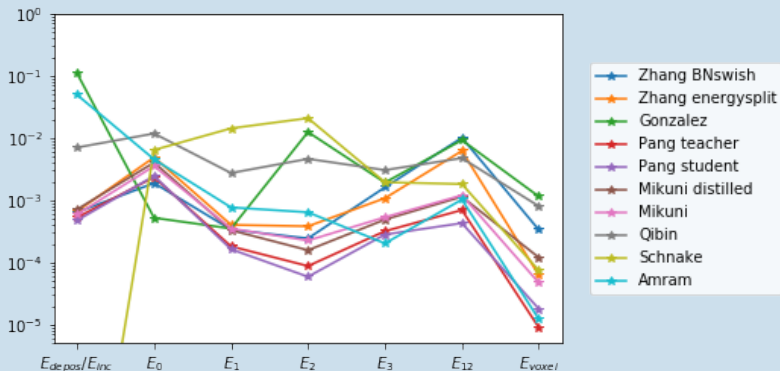
- Looking at histogram χ^2 of $E_{tot}/E_{inc}, E_{layer0}, E_{layer1}, E_{layer2}, E_{layer3}, E_{layer12}, E_{voxel}$



⇒ Keep in mind: I didn't check E_{inc} -dist, normalizations, doubles, some sets are imbalanced, ...

Histograms ds1 photons

- Looking at histogram χ^2 of $E_{tot} / E_{inc}, E_{layer0}, E_{layer1}, E_{layer2}, E_{layer3}, E_{layer12}, E_{voxel}$



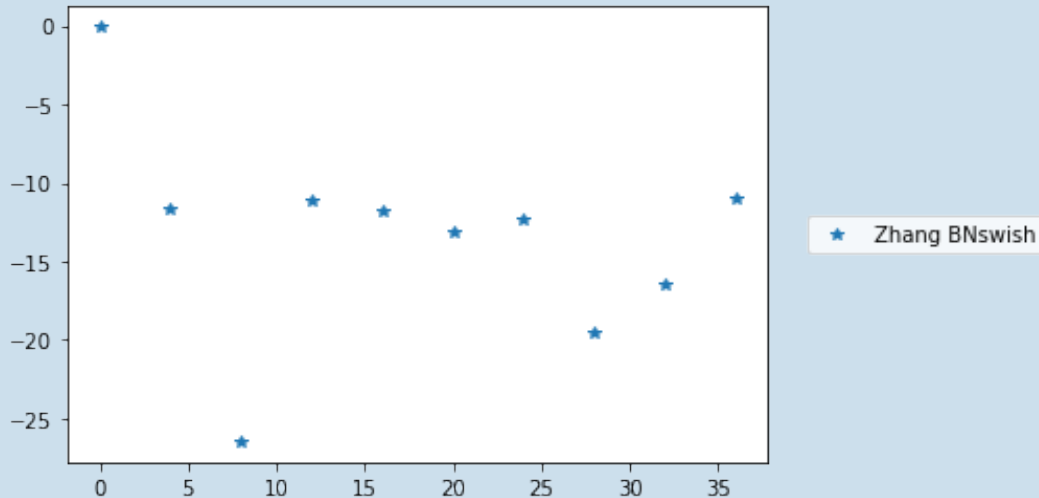
⇒ Keep in mind: I didn't check E_{inc} -dist, normalizations, doubles, some sets are imbalanced, ...

Architecture and Training

- Simple DNN with 2 hidden layer of 2048 neurons.
 - features: $\log_{10} E_{inc}, \mathcal{I}_a / E_{inc}$
 - Cross Entropy loss, ADAM optimizer, 50 epochs (val loss min around 15)
- ⇒ Keep in mind: I didn't check E_{inc} -dist, normalizations, doubles, thresholds, ...
pion data has doubles, some sets are imbalanced, only 1 training

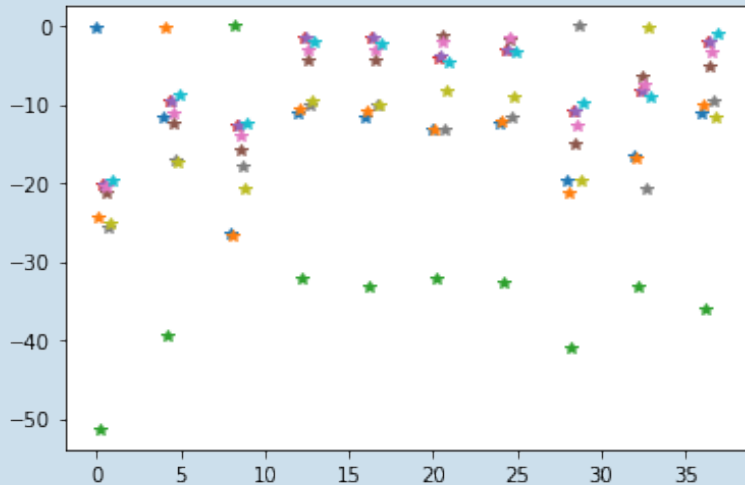
Log-posterior ds1 photons

submission vs submission: what we see: $\text{mean log } p(\text{model} | \text{data})$



Log-posterior ds1 photons

submission vs submission:

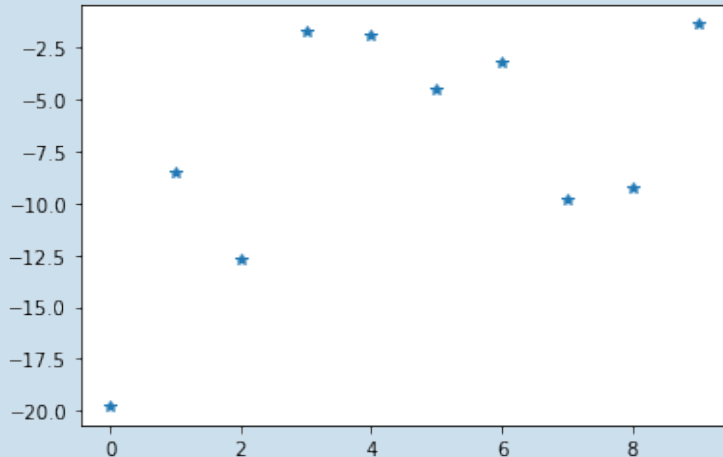


- ★ Zhang BNswish
- ★ Zhang energysplit
- ★ Gonzalez
- ★ Pang teacher
- ★ Pang student
- ★ Mikuni distilled
- ★ Mikuni
- ★ Qibin
- ★ Schnake
- ★ Amram

Log-posterior ds1 photons

submission vs geant:

Geant Truth



Zhang BNswish: -19.78

Zhang energysplit: -8.57

Gonzalez: -12.70

Pang teacher: -1.76

Pang student: -1.90

Mikuni distilled: -4.57

Mikuni: -3.18

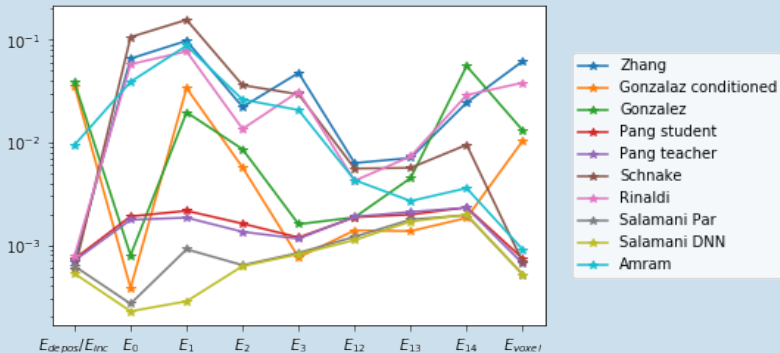
Qibin: -9.84

Schnake: -9.29

Amram: -1.39

Histograms ds1 pions

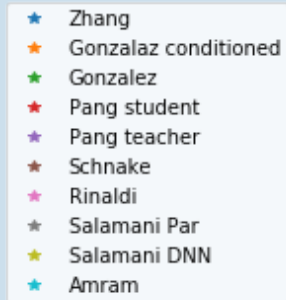
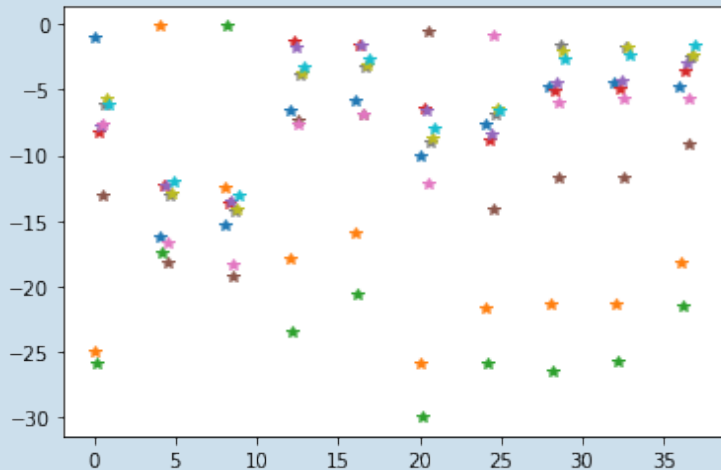
- Looking at histogram χ^2 of $E_{tot} / E_{inc}, E_{layer0}, E_{layer1}, E_{layer2}, E_{layer3}, E_{layer12}, E_{layer13}, E_{layer14}, E_{voxel}$



⇒ Keep in mind: I didn't check E_{inc} -dist, normalizations, doubles, some sets are imbalanced, (only 1 training), ...

Log-posterior ds1 pions

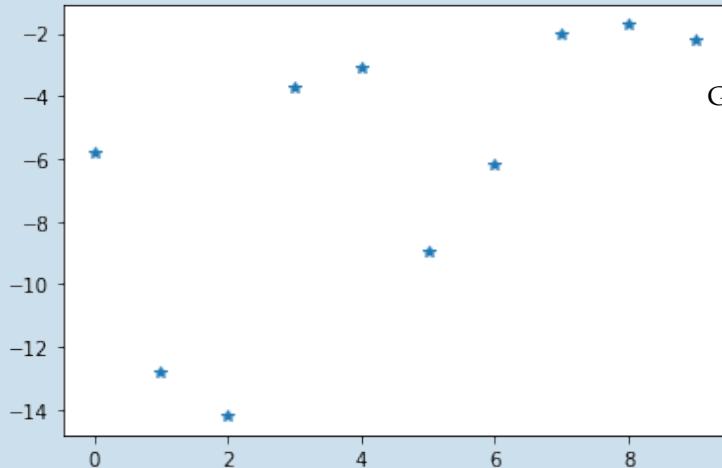
submission vs submission:



Log-posterior ds1 pions

submission vs geant:

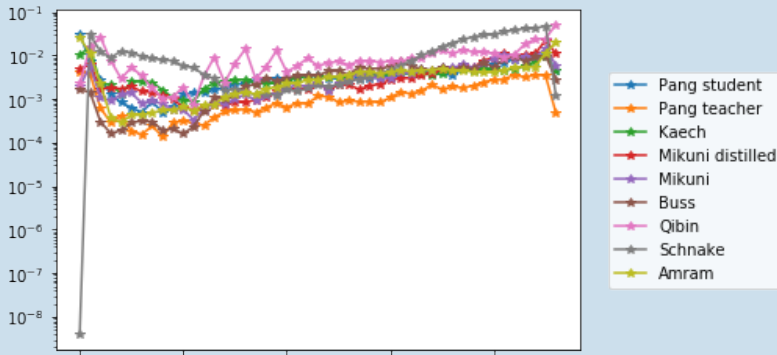
Geant Truth



Zhang: -5.81
Gonzalaz conditioned: -12.83
Gonzalez: -14.19
Pang student: -3.74
Pang teacher: -3.09
Schnake: -8.94
Rinaldi: -6.19
Salamani Par: -2.00
Salamani DNN: -1.72
Amram: -2.20

Histograms ds2

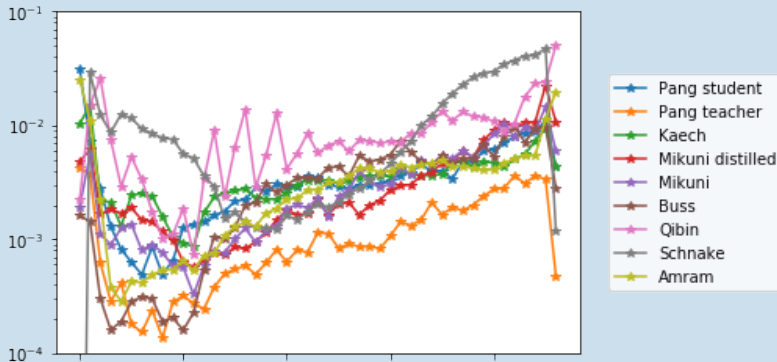
- Looking at histogram χ^2 of $E_{tot} / E_{inc}, E_{layer0}, E_{layer1}, \dots, E_{layer44}, E_{voxel}$



⇒ Keep in mind: I didn't check E_{inc} -dist, normalizations, doubles, some sets are imbalanced, (only 1 training), ...

Histograms ds2

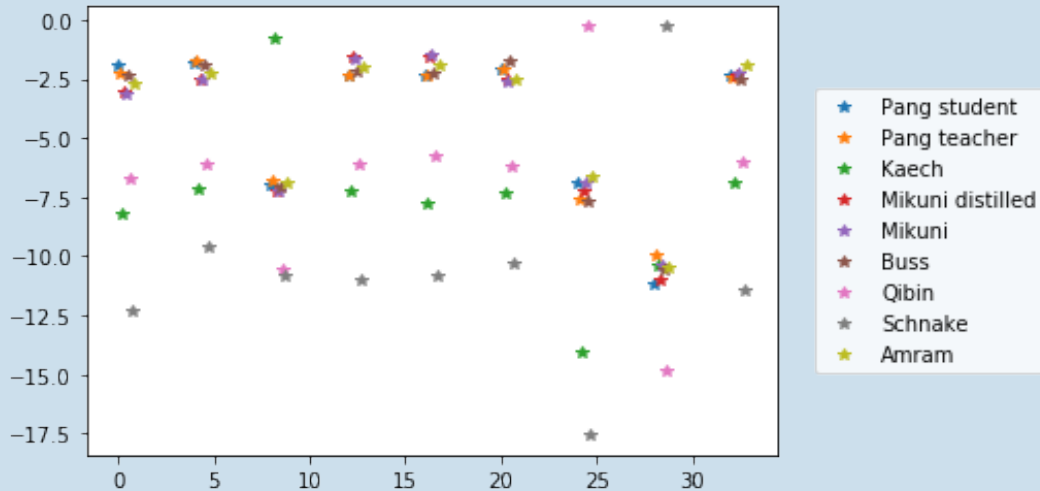
- Looking at histogram χ^2 of $E_{tot} / E_{inc}, E_{layer0}, E_{layer1}, \dots, E_{layer44}, E_{voxel}$



⇒ Keep in mind: I didn't check E_{inc} -dist, normalizations, doubles, some sets are imbalanced, (only 1 training), ...

Log-posterior ds2

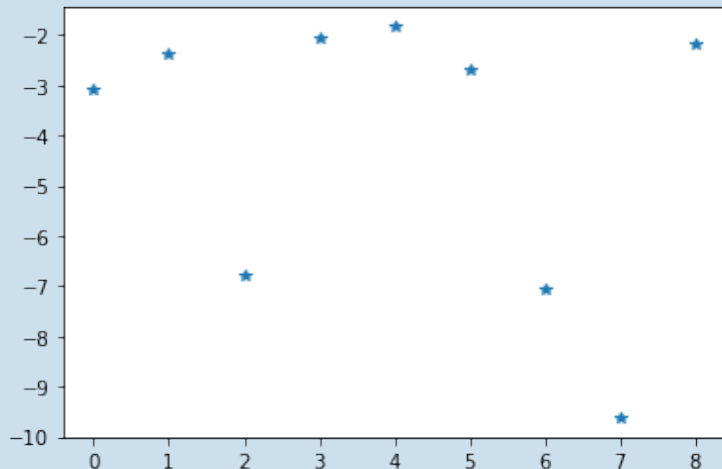
submission vs submission:



Log-posterior ds2

submission vs geant:

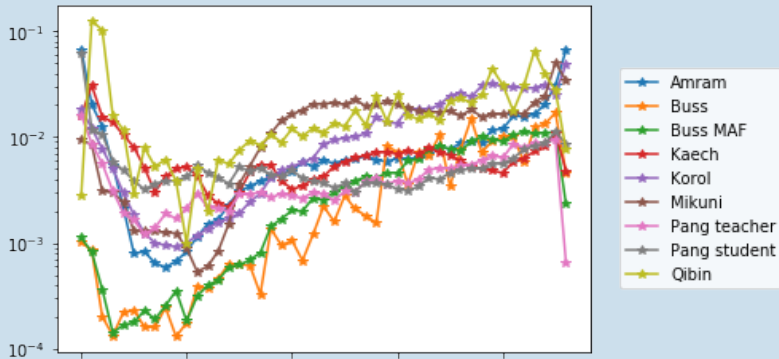
GEANT Truth



Pang student: -3.07
Pang teacher: -2.37
Kaech: -6.79
Mikuni distilled: -2.07
Mikuni: -1.84
Buss: -2.69
Qibin: -7.06
Schnake: -9.62
Amram: -2.17

Histograms ds3

- Looking at histogram χ^2 of $E_{tot} / E_{inc}, E_{layer0}, E_{layer1}, \dots, E_{layer44}, E_{voxel}$



⇒ Keep in mind: I didn't check E_{inc} -dist, normalizations, doubles, some sets are imbalanced, ...

Log-posterior ds3

submission vs submission
submission vs geant

Not enough time / RAM;-)

Next steps:

- We will publish a new dataset for ds1 pions: train + eval
- We will run the detailed comparison of more metrics.
- ⇒ I will contact you more often with questions, maybe we should have a common channel in the ML4Jets SLACK?
- We will summarize our results in a joint paper, to be submitted to EPJC or Rept.Prog.Phys.
- ⇒ We need your (final!) samples by end of July!
- ⇒ We need your (final!) models by end of July!
- ⇒ We need your write-up by end of July!

- ⇒ Have it on arXiv for ML4Jets in November.

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- Each contribution should write 1-3 pages about Architecture, Preprocessing and Training
- Each group of contributions (GANs, Flows, ...) should coordinate the general model introduction
- We will draft the results based on everyone's feedback/contributions

The Summary Paper — Results section

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- Histograms of evaluate.py, sparsity, ...
- timings / number of trainable parameters / memory usage on same hardware
- multiclass and sample vs. Geant classifiers, sliced by E_{inc} , model type, large/small weights, ...
- correlation plots of the above

The Fast Calorimeter Challenge 2022: Results & The Road Ahead

- Thank you to everyone who contributed to the challenge — via samples, ideas, discussions, ...!!!
- Some fun is behind us, but there is a lot more ahead:
 - ⇒ Re-training / fine-tuning of models
 - ⇒ in-depth comparisons
 - ⇒ the write-up
- A Big Thank You to Michele Faucci Giannelli, Marco Vanadia, Umberto De Sanctis, and all local organizers!