# ML<sub>4</sub>Science

Week 3 meeting

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## Outline

- 1. Improvements of NN of last time
  - Sigmoid removed
  - Different way to standardize data
- 2. New networks
  - NN with data of last time
  - NN with new data and masking to remove zeros
  - NN with new data without masking



### Architecture

As discussed in the email of this week, we decided to change a little bit the architecture of our network. After some trials, we understand that the following architecture works really well.

```
self.11 = nn.Linear(num-feature,64,dtype=dtype)

self.reLU1 = nn.ReLU()

self.12 = nn.Linear(64,128,dtype=dtype)

self.reLU2 = nn.ReLU()

self.13 = nn.Linear(128,256,dtype=dtype)

self.reLU3 = nn.ReLU()

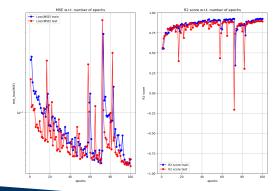
self.14 = nn.Linear(256,1,dtype=dtype)
```



### Results 1

We firstly decided to work with the same data as last time to have a better comparison on architectures. So we firstly fixed  $LH_0$  and we worked with all z.  $R_{test}^2 \approx 0.85$ .

 $\it X$ : Normal standardization,  $\it Y$ : Division for  $10^{10}$ 





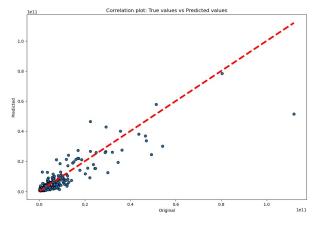


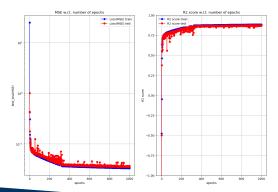
Figure 2: Correlation plot old data.



#### Results 2

Then we moved on new data with masking in order to remove all the zeros (still, fixed  $LH_0$  and we worked with all z).  $R_{test}^2 \approx 0.85$ .

X: Normal standardization, Y:  $log_{10}(Y)$ .





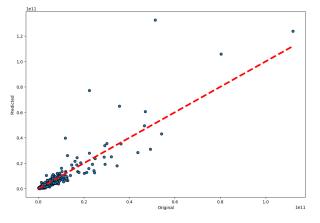


Figure 4: Correlation plot, new data and masking.



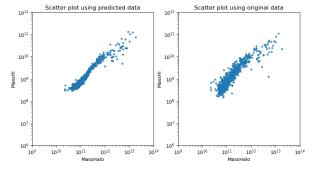
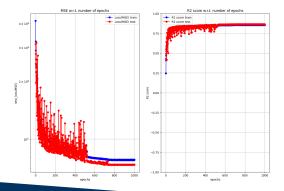


Figure 5: Cloud of points, new data and masking



## Results 3

Finally, in order to generalize at the whole cloud of points, we tried to train the same model as before without masking (still, fixed  $LH_0$  and we worked with all z).  $R_{test}^2 \approx 0.85$ . X: Normal standardization, Y:  $log_10(Y)$ .





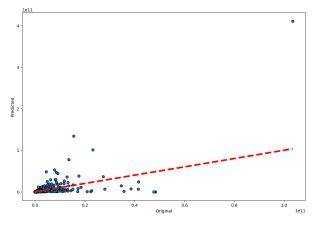


Figure 7: Correlation plot, new data and NO masking.



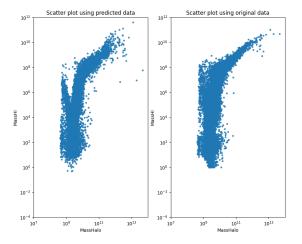


Figure 8: Cloud of points, new data and NO masking



#### Conclusions

From these trials we concluded that:

- Leaky Relu increases loss, slow down the train and decrease the  $R^2_{test}$ .
- Dropout increases loss, slow down the train and decrease the  $R^2_{test}.$
- We believe that now the standardization is quite good and also the results3 tell us that our NN learn at least the unit of the output ( $loss_{test} \approx 0.8$ ).
- The architecture *mirror* does not work well for this problem.
- This model cannot be generalized for different LH, so also the cosmological constants play an important role.

