

# Cloud Atlas

## An LstmEncoder for UHECR AirShowers

G. Becuzzi   L. Papalini

July 2022

# Table of Contents

# Table of Contents

Questo lo fa la Lush

# Dataset, first glance

The dataset is composed of  $10^5$  simulated events:

- 9x9 grid of detectors
- most intense detector at the center
- 80 frames of time series (40 MHz sampling rate)
- 1 frame of times of first arrival

The single record shape is then  $(80 + 1, 81)$

The grid is hexagonal (adjacency matrix available) but we neglect the structure of the detector since the net can learn it.

The `pd4ml` package splits by default in 70% train 30% test.

# Table of Contents

# Split the dataset

Using a generator (`keras.utils.Sequence`)

- inherit multiprocessing features
- has default callbacks

The dataset is splitted *record by record* for index shuffling

The effect of the high reading time from memory ( $\approx 3ms$ ) is mitigated by `keras` multiprocessing

For the design of the net it is convenient using `numpy` structured arrays

# Split the dataset: `funky_dtype`

Data is extracted: from a conceptually *ihomogeneous* list (activity time series together with times of arrival) to  
 $(80 + 1, 81) \rightarrow [(\text{"toa"}, (9, 9, 1)), (\text{"timeseries"}, (80, 9, 9))]$   
Data can be accessed depending on what is needed



# DataFeeder class

# DataFeeder class

Augmentation class e amici

# Resolution

The reference article suggest using the resolution:

resolution

defined as the standard deviation of the distribution bla bla bla

We point out that

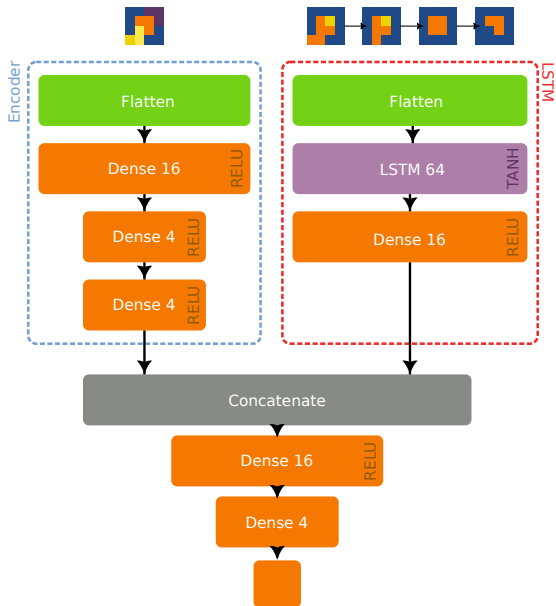
$$\sigma^2 = \frac{1}{N} \sum_i (\delta_i - \bar{\delta})^2$$

is a valid estimator only if  $\bar{\delta} = 0$ , for which the adopted resolution is equal to the *RMSE* of the distribution

$$RMSE^2 = \frac{1}{N} \sum_i (x_i - \hat{x}_i)^2$$

Thus we preferred the latter as a measure of the estimate.

# Table of Contents



# Overview on the network

# Encoder for time of arrivals



i graficini loss accuracy ecc ecc

si spiega che cos'è

# LSTM for the time series

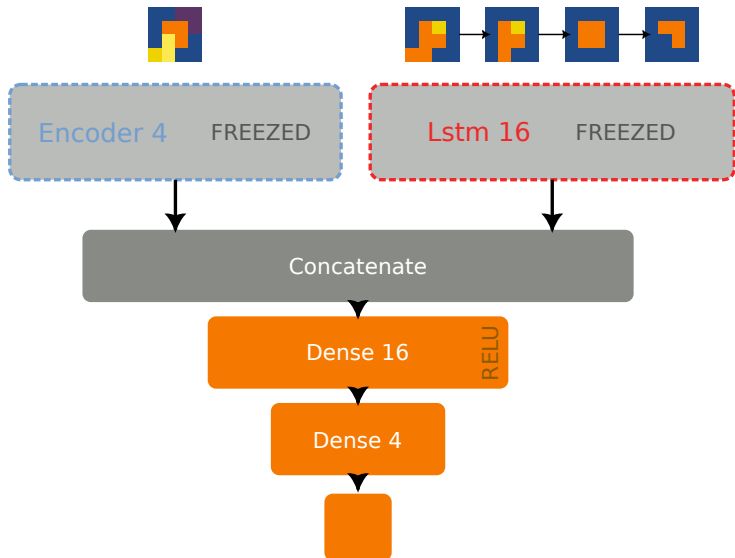
si fa vedere come abbiamo fatto noi

# LSTM performance

same

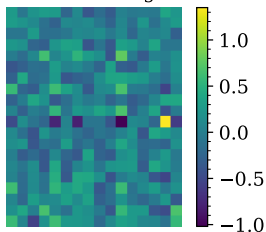
# Concatente + dense layers

# Subnets train freezing

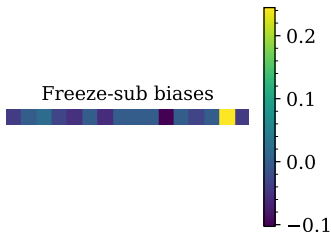


# Subnets train freezing

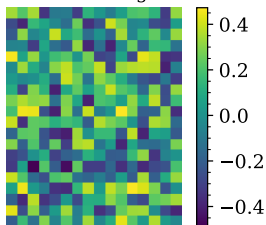
Freeze-sub weights



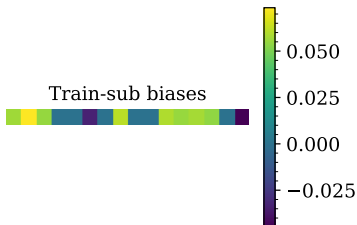
Freeze-sub biases



Train-sub weights



Train-sub biases



# Network's output



# Hyperparameters tuning

# Whole Network performance

# Test setup on CircleCI

Danke Schon