

## Fiche TP1 : Training an MLP from scratch

- 1- Write the function **init\_params(nx, nh, ny)** that initialize the weights of an MLP based on nx, nh, and ny that presents the number of neurons in each layer : input (x), hidden (h), and output (y). All weights must be initialized following the normal distribution with an average of 0 and a standard deviation of 0.3.
- 2- Write the function **forward(params, X)** that receive a batch of data X with a size of  $n_{batch} \times n_x$  and the different parameters in **params**. The function returns  $\hat{Y}$  and intermediate values.
- 3- Write the function **loss\_accuracy(Yhat, Y)** that returns the loss and accuracy.
- 4- Write the function **backward(x, params, outputs, Y)** that compute the loss gradients regarding parameters and stock them in a dictionary.
- 5- Write the function **sgd(params, grads, eta)** that apply mini batch sgd and update the network parameters based on gradients and the learning rate (eta).
- 6- Write the training steps based on the previous function and display Accuracy and loss graphs.
- 7- Train the network on the Hand written digit recognition dataset MNIST.

**Ps:** Use matrix multiplications. Activation functions: Tanh (for hidden layers) and Softmax for the Output. Loss function: Cross entropy.

**Parameters:** 0.1, 128, 50.

- Load data
- Initialize the Network
- For  $I = 1 \dots n_{epoch}$ 
  - Random the data
  - For  $J = 1 \dots N/n_{batch}$ 
    - Load a batch of data
    - Forward Propagation on the batch
    - Compute the loss on the batch
    - Backpropagation on the batch
    - Apply SGD to update parameters

- Plot the accuracy and loss histories.