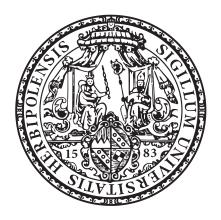
# **PNN Framework - 2023**

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Explanation for the implemented framework - PNN2023

Chair for Computer Science VI University of Würzburg Programming with Neural Networks

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## 1. Structure and Functionalities

The source code has the following structure and functionalities:

- src
  - data/MNIST/raw
    - \* .\*?(.gz)?: all the raw data downloaded via pytorch
  - layer
    - \* ActivationLayer.py: Can be initialised with the following activation functions (by passing the string mentioned in the brackets):
      - · ReLu ('relu')
      - · Sigmoid ('sigmoid')
      - · Tanh ('tanh')
    - \* Conv2DLayer.py: Takes four initial integers:
      - $\cdot$  x the size of the rows
      - $\cdot$  y the size of the columns
      - · channels the number of channels
      - · numFilters the total number of filters to be applied

Each filter is initialised according to a uniform distribution between -1.0f and 1.0f and a bias of 0.

- \* FlattenLayer.py: A utility layer that takes the expected incoming shape for initialisation; flattens incoming tensors of the given shape into a one-dimensional vector (or a two-dimensional one with shape (1, n)).
- \* FullyConnectedLayer.py: Takes an integer each for the input's shape and the wanted output's shape. The weight matrix' values are initialised randomly (according to a uniform distribution) between -1.0f and 1.0f. The bias is initialised with zeros.
- \* InputLayer.py: Expects a list (or any stacking of lists) of floats and turns it into a corresponding list of Tensors.
- \* Layer.py: Abstract class that defines all needed functions of each layer.
- \* LossLayer.py: Incorporates two types of losses:
  - · Cross entropy loss
  - · Mean squared error loss
- \* Pooling2DLayer.py: Takes the shape of the filter and the respective stride for initialisation. Aggregates according to the filter and stride in two possible ways:
  - · Taking the maximum of the overlapped filter
  - · Taking the average of the overlapped filter
- \* SoftmaxLayer.py: The usual softmax.

#### model

\* Tensor.py: Describes a layer's values, deltas, and shape.

#### network

- \* Network.py: Describes the neural network. After initialising an empty network, the aforementioned layers can be added progressively to create a fully functional network. The corresponding input and label(s) have to be passed to the forward function to get a result/loss for an input. During this process, a cache is created which saves the results of the forward pass. The cache runs the respective updates on each layer when running the backward pass (which needs a learning rate). Aside from that, there is an optional debug flag (on initialisation) to enable prints of the state of the net during the respective passes.
- \* SGDTrainer.py: Applies the standard gradient descent to a neural network with the given data. The hyperparameters (e.g. number of epochs, learning rate, etc.) can be adjusted on initialisation. Furthermore, just like the neural net, there is a debug flag to enable print statements and another loss flag that prints the current loss at a rate given by the loss\_batch parameter (the loss flag has a higher priority than the debug flag). (There is also another flag, the dev flag, which enables an early break if the improvement from the last epoch is too little.)
- \* UpdateMechanism.py: Enum taken from the lecture.
- Dummy.py: Messing around with neural nets and layers.
- LoadDataMNIST.py: Loads the MNIST dataset via torch(vision). There is a variant each for the CNN and the FNN as they work with different dimensionalities.
- MNIST-CNN.py: Runs a certain CNN on the MNIST dataset and evaluates it via accuracy; usually runs on one certain config (1 epoch, 0.01 learning rate).
   Due to the long training time, there are some optional functions:
  - \* pickle\_run: Loads an already trained net and runs through the evaluation of the net.
  - \* save\_current\_net: Saves one trained net that has the best performance among all other ones after the complete training procedure.
- MNIST-FFNN.py: Runs a certain FFNN on the MNIST dataset and evaluates it via accuracy; has some configs to run multiple times which should be adjusted one's need in the first few lines of the main function.
- MNIST-Torch.py: Torch playground to create nets and run them through MNIST.

### 1. Structure and Functionalities

#### • test

- test\_activation.py: Unit tests for the relu and sigmoid activation layer as well as the softmax layer.
- test\_layer.py: Unit tests for the fully connected, convolutional, and pool layer.
- test\_losses.py: Unit tests for the loss layers, e.g. cross entropy and mean squared error loss.

#### a) FFNN

```
# Assumption of correctly imported classes etc.
# initialises the standard gradient descent trainer with certain
# hyperparameters and the debug flag
# --> forces print statements during the network optimisation
sgd = SGDTrainer(learningRate=1, amountEpochs=3, debug=True, loss=False)
# initialising an empty network and adding consecutive layers
# one after another
net = Network()
# fully connected layer of 3 rows and 3 columns
net.addLayer(FullyConnectedLayer(3, 3))
# activation layer sigmoid by passing the according string
net.addLayer(ActivationLayer('sigmoid'))
# fully connected layer of 3 rows and 2 columns
net.addLayer(FullyConnectedLayer(3, 2))
# softmax layer
net.addLayer(SoftmaxLayer())
# cross entropy loss
net.addLayer(CrossEntropyLoss())
# the data the net will be trained on with sgd
# only consists of one pair (input, labels)
data = [([0.4183, 0.5209, 0.0291], [0.7095, 0.0942])]
# trains the net with the sgd trainer
sgd.optimize(net, data, [], [])
```

#### b) CNN

```
# Assumption of correctly imported classes etc.
# initialises the standard gradient descent trainer with certain
# hyperparameters and the debug flag
# --> forces print statements during the network optimisation
sgd = SGDTrainer(learningRate=0.01, amountEpochs=1, debug=True, loss=False)
# initialising an empty network and adding consecutive layers
# one after another
net = Network()
# convolutional layer of 9 rows, 9 columns, and 1 channel
# while using 4 filters
net.addLayer(Conv2DLayer(9, 9, 1, 4))
# activation layer relu by passing the according string
net.addLayer(ActivationLayer('relu'))
# pooling layer that uses averaging as aggregation mechanism
# filter of size 4x4 and a stride of 4x4
net.addLayer(AvgPooling(filter_shape=(4, 4), stride=(4, 4)))
# flattening layer to transform the result into 1d
net.addLayer(FlattenLayer(input_shape=(4, 5, 5)))
# fully connected layer of 100 rows and 10 columns
net.addLayer(FullyConnectedLayer(100, 10))
# softmax layer
net.addLayer(SoftmaxLayer())
# cross entropy loss
net.addLayer(CrossEntropyLoss())
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# only consists of one pair (input, labels)
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