# Technical documentation

# Reading the datasets

A dataset is read from 2 files: one contains the pixel-data, the other one contains the labels. They are read simultaneously and stored in RAM in a Dataset struct.

## Format of the image files

- 4B magic number
- 4B # images
- 4B # cols
- 4B # rows
- images -> 1B per pixel

# Format of the label files

- 4B magic number
- 4B # labels
- labels -> 1B per label

# Model breakdown

We use a fully-connected feed-forward deep neural network. A general case can be found in the image below.

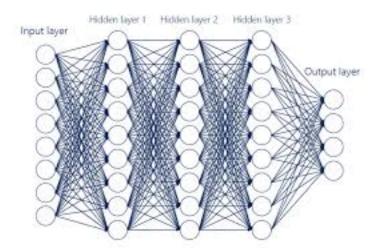


Figure 1: Image of a neural network

## Input layer

A flattened vector of the input pixels normalized to [0:1] interval.

# Hidden layers

These are defined by the user. Their size and activation function can be customized.

## Output layer

As this problem is a classification problem, the output layer's activation function is softmax. The output layer has 10 neurons, as we have 10 classes.

# Implementing the feed-forward algorithm

For every layer, we compute  $\underline{\phantom{a}}$   $f(W^*x+b)\underline{\phantom{a}}$ , where f is the activation function, W is the weight matrix, x is the previous layer's value-vector and b is the bias vector.

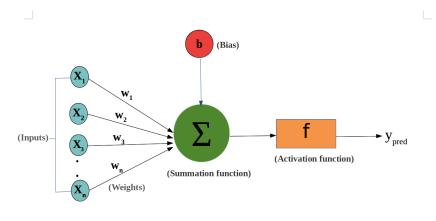


Figure 2: Image of the feed-forward algorithm

# Implementing backpropagation

We use cross-entropy for the loss-function, as it is great for classification prob-

Articles for better understanding:

- Backpropagation
- Softmax and cross-entropy loss

We first compute the loss, then we compute the derivative of the loss with respect to the neurons in each layer. We do this first, so we can use the chain-rule later with ease. Then we adjust the weights and biases using the chain rule. This is easy, as we just use the previously computed derivatives.

# Overview of the program's structure

The program reads the arguments, determines the actions, it would have to take (training, testing of accuracy, showing images), and does them sequentially. The other parameters just customize this basic flow (eg.: by setting the dataset's folder). ## Saving / loading models

Each model can be saved/loaded to/from a file. Usually we use the .model extension, but it can differ.

## The file-structure of the .model extension

- 1B # layers
- for each layer:
  - 4B inputDim
  - 4B outputDim
  - inputDim \* outputDim \* 8B weights
  - outputDIm \* 8B biases

# Important types

#### Model (struct)

Defines a model, holds its layers in a heap-allocated array.

# Layer (struct)

Defines a layer, holds its dimensions, and its weights and biases in heap-allocated arrays.

## Dataset (struct)

Defines a dataset, conatins its images and labels in a heap-allocated LabeledImage array.

### LabeledImage (struct)

Defines an image with its label. Contains its pixels in a stack-allocated array. Should not be copied much!

#### Result (struct)

Defines a prediction-result. Holds 10 stack-allocated probabilities.

# **API** documentation

# dataset.h ReadInt int ReadInt (FILE\* fp); Summary: Reads 4 bytes from a file and converts them into a signed int32 t ReadByte uint8\_t ReadByte (FILE\* fp); Summary: Reads 1 byte from a file FreeDataset void FreeDataset (Dataset dataset); Summary: Frees up a dataset (with all its images) GetPixelOfImage double GetPixelOfImage(LabeledImage image, int x, int y); Summary: Returns the pixel value of an image at the given coordinates ${\bf Print Labeled Image}$ void PrintLabeledImage (LabeledImage image); Summary: Prints the given image into stdout ReadDatasetFromFileDataset ReadDatasetFromFile (const char\* imagePath, const char\* labelPath); Summary: Reads 2 files (one for the images, one for the labels) and zips them into a single dataset main.c main int main (int argc, char \*\*argv) { **Summary:** The entry point of the program.

# manager.h

#### GetDefaultSetup

```
ProgramSetup GetDefaultSetup ();
```

Summary: Sets the program into default mode.

#### **ProcessArgs**

```
ProgramSetup ProcessArgs (int argc, char** argv);
```

**Summary:** Reads the command line arguments (passed as args here), and determines the mode, the program should be running in.

## **PrintImagesInfinitely**

```
void PrintImagesInfinitely (Dataset dataset);
```

**Summary:** Prints images to stdout with their label for as long as the given dataset lasts

#### **TestXORProblem**

```
void TestXORProblem (
    int numNeuronsInHiddenLayer,
    ActivationFunction activationFunction,
    int iterations,
    double learningRate,
    double maxDeviationFromResult
);
```

**Summary:** Tests the integrity of the code, by running the XOR problem on it. If the backpropagation code is written successfully, the model should be able to learn a non-linear function (the XOR function).

## **GetAccuracy**

```
double GetAccuracy (Model model, Dataset dataset);
```

Summary: Calculates the accuracy of a model on a dataset.

## ${\bf Fit Model}$

```
void FitModel (
    Model model,
    Dataset trainSet,
    Dataset testSet,
    uint8_t numEpochs,
    double learningRate,
```

```
bool saveContinuously,
    const char* savePath
);
Summary: Fits a model with the given parameters.
{\bf Print Images With Predictions}
void PrintImagesWithPredictions (Model model, Dataset dataset, bool onlyWrongs);
Summary: Prints images to stdout with their labels and the given model's
predictions.
model.h
Sigmoid
double Sigmoid (double x);
Summary: The sigmoid function f(x) = 1 / (1 + e^-x)
SigmoidDer
double SigmoidDer (double x);
Summary: Derivative of the sigmoid function
ReLU
double ReLU (double x);
Summary: The ReLU function f(x) = x >= 0? x : 0
ReLUDer
double ReLUDer (double x);
Summary: The derivative of the ReLU function
LReLU
double LReLU (double x);
Summary: The Leaky ReLU function
LReLUDer
double LReLUDer (double x);
```

Summary: The derivative of the LReLU function

#### **PrintModel**

```
void PrintModel(Model model);
Summary: Prints a model to stdout
```

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# ${\bf Print Model Layout}$

```
void PrintModelLayout(Model model);
```

Summary: Prints a model's architecture / layout to stdout

#### **PrintResult**

```
void PrintResult(Result result);
```

Summary: Prints a result's probabilities to stdout

#### FreeModel

```
void FreeModel(Model model);
```

**Summary:** Frees up all the memory the given model uses (all weights, biases, layers, etc)

### FreeLayer

```
void FreeLayer(Layer layer);
```

Summary: Frees up all memory the given layer uses

# InitModelToRandom

```
void InitModelToRandom (Model* model, double randomRange);
```

**Summary:** Initializes the given model's weights and biases to a random value between [-randomRange, +randomRange] This will override all weights and all biases, so should only be called on uninitialized models!

# CreateModel

```
Model CreateModel(int numHiddenLayers, ...);
```

Summary: Sets up the layers of a model from dimension and activation fn lists

- ...: size, activation fn, size2, activation fn2, activation fn3 The number of additional parameters must be at least 1 (activation fn of last layer) @return model instance

## ${\bf Create Model From Layout}$

```
Model CreateModelFromLayout(LayerLayout* layout);
```

Summary: Creates an uninitialized model from a linked-list of LayerLayouts

#### ${\bf Make Value Buffer For Model}$

```
double** MakeValueBufferForModel (int numLayers);
```

**Summary:** Allocates a buffer with dimensions defined by the model's layers. It does not create a buffer for the -1st vector (the image)

• model: @return pointer to the buffer

#### FreeValueBuffer

```
void FreeValueBuffer (Model model, double** buffer);
```

Summary: frees the value buffer

#### Predict

```
Result Predict(Model model, double* input, double** out_neuronValues);
```

**Summary:** Forwards the given image through the given model, and optionally saves the SUMS of the neurons to a buffer (not the values after the activation function)

- model: the model, with which we want to predict
- input: input image's data
- out\_neuronValues: the SUMS!!! of the neurons, so  $sum(w^*n + b)$

Returns: the prediction result

#### CalculateCost

```
double CalculateCost(uint8_t label, double* resultValues);
```

**Summary:** implements cross-entropy algorithm cross-entropy: -sum(y[i] \* log(s[i])) for i: 0..NUM\_CLASSES

Returns: Cross-entropy cost of the given image with the label

## **BackPropagate**

```
void BackPropagate(
    Model model,
    double** neuronValues,
    LabeledImage* image,
    double learningRate,
    double** preallocatedDerBuffer,
```

```
double* out_cost
);
```

Summary: Adjusts the model's weights

• **neuronValues**: holds the values of the neurons, filled during the prediction phase

## ${\bf Fit Model For Image}$

```
bool FitModelForImage(
    Model model,
    LabeledImage* image,
    double learningRate,
    double** preallocatedValueBuffer,
    double** preallocatedDerBuffer,
    double* out_cost
);
```

**Summary:** Runs predict and then backpropagation on the model with the given image and target.

- model: it is ok if it's not a pointer, because the layers variable will still point to the same memory address.
- **preallocatedDerBuffer**: a preallocated buffer to store the derivatives of the neurons. It is cheaper this way.
- preallocatedValueBuffer: a preallocated buffer to store the values of the neurons.

**Returns:** whether the result is ok or  $\inf/-\inf/nan -> true = good$ , false = stop learning.

#### ${\bf Calculate Avg Cost For Model}$

```
double CalculateAvgCostForModel (Model model, LabeledImage* images, int numImages);
Summary: Runs the model with the given image and gives back the cost
```

#### ${\bf GetPredictionFromResult}$

```
int GetPredictionFromResult(Result result);
Summary: argmax
```

#### **IsResultOk**

```
bool IsResultOk (Result result);
```

Summary: Determines whether the result's probs are valid numbers.

# ${\bf Save Model To File}$

```
void SaveModelToFile (Model model, const char* filePath);
```

Summary: Writes the model's weights and biases to a file with the given path.

# ${\bf Load Model From File}$

```
Model LoadModelFromFile (const char* filePath);
```

 ${\bf Summary:}\ {\bf Reads}\ {\bf the}\ {\bf given}\ {\bf file}\ {\bf and}\ {\bf returns}\ {\bf the}\ {\bf model}\ {\bf stored}\ {\bf in}\ {\bf it}.$