



On-line recognition of handwritten mathematical symbols

Bachelor's Thesis of

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I declare that I have developed and written the enclosed thesis completely by myself, and have not used sources or means without declaration in the text.
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(Martin Thoma)

Acknowledgement

TODO







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1. Introduction

Handwriting recognition is the problem of finding a proper textual representation given a handwritten symbol or sequence of symbols.

In off-line handwriting recognition, all algorithms have to work on pixel image information of the handwriting. On-line handwriting recognition on the other hand can use the information how symbols were written.

2. Baseline system

A system for symbol recognition was already written and is described in [Kir10].

3. Preprocessing

TODO

4. Artificial Neural Nets

Artificial neural networks (ANNs) are models for classification that were inspired by the brain. They consist of artificial neurons and have a lot of different subtypes like Feed Forward Neural Nets.

4.1. Artificial neurons

Artificial neurons are inspired by biological neurons.

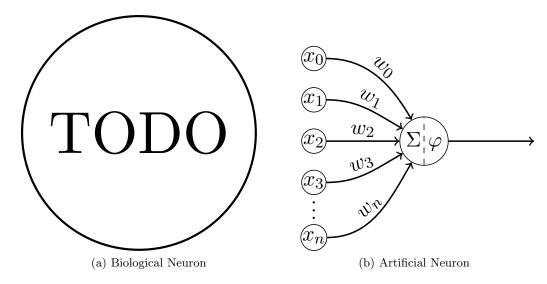


Figure 4.1.: Both neurons receive weighted input, apply a function to that and give output

They receive at least one input and at least one output. Those inputs might get weighted as well as the output.

The neurons apply a function to the sum of all weighted inputs. This function is called activation function.

An artificial neuron using the unit step function (see section 4.3.1) is called a *perceptron*.

The artificial neuron sums all weighted inputs $x_i \cdot w_i$ up and applies its activation function f to it.

4.2. Feedforward Neural Nets

Feedforward neural nets don't have loops.

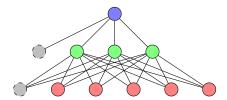


Figure 4.2.: Feedforward artificial neural network

4.3. Activation functions

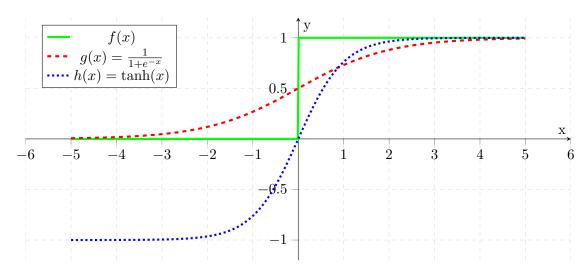


Figure 4.3.: The unit step function f, the sigmoid function g and the hyperbolic tangend h.

4.3.1. Unit step function

Not so good, because it's not differentiable. Therefore, the backpropagation algorithm cannot be used.

4.3.2. Sigmoid function

Is great because it is infinitely often differentiable.

4.3.3. Hyperbolic tangent

Also differentiable, but gradient descent converges faster (sometimes?)

5. Evaluation

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5.1. Section 1

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5.2. Section 2

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5.3. Section 3

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6. Conclusion

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Bibliography

[Kir10] D. Kirsch, "Detexify: Erkennung handgemalter latex-symbole," Diploma thesis, Westfälische Wilhelms-Universität Münster, 10 2010. [Online]. Available: http://danielkirs.ch/thesis.pdf

Glossary

ANN artificial neural network. 7

Appendix

A. First Appendix Section

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Figure A.1.: A figure

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