

TECHNISCHE UNIVERSITÄT MÜNCHEN

Master's Thesis in Informatics

A Novel Transfer Function for Continuous Interpolation between Summation and Multiplication in Neural Networks

Wiebke Köpp





TECHNISCHE UNIVERSITÄT MÜNCHEN

Master's Thesis in Informatics

A Novel Transfer Function for Continuous Interpolation between Summation and Multiplication in Neural Networks

Stetige Interpolation zwischen Addition und Multiplikation unter Verwendung einer geeigneten Transferfunktion in Neuronalen Netzen

Author: Wiebke Köpp

Supervisor: Prof. Dr. Patrick van der Smagt Advisor: Dipl.-Phys. Sebastian Urban

Submission Date: December 15, 2015



I confirm that this master's thes all sources and material used.	is in informatics is my ow	vn work and I have documented
Munich, December 15, 2015		Wiebke Köpp

A Novel Transfer Function for Continuous Interpolation between Summation and Multiplication in Neural Networks

Wiebke Köpp

KOEPP@IN.TUM.DE

Technische Universität München, Department of Informatics

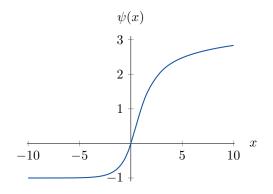


Figure 1. A continuously differentiable solution $\psi(x)$ for Abel's equation (??) with $f(x) = \exp(x)$.

Abstract

Abstract

1. Fractional Iterates of the Exponential Function

1.1. Abel's Functional Equation

Assuming $\psi(x) = x$ for $0 \le x < 1$, we can find a solution for ψ which is valid on all of \mathbb{R} . Since we have $\log(x) < x$ for any $x \ge 1$, multiple applications of $(\ref{eq:condition})$ will, at some point, lead to an argument for ψ that falls into the interval where we assume ψ to be linear. In case x < 0, we can use $\psi(x) = \psi(\exp(x)) - 1$ to reach the desired interval, since $0 < \exp(x) < 1$ for x < 0. In combination, this leads to a piece-wise defined solution for ψ ,

$$\psi(x) = \log^{(k)}(x) + k \tag{1a}$$

with
$$k \in \mathbb{N} \cup \{-1, 0\} : 0 \le \log^k(x) < 1$$
. (1b)

The function is displayed in Fig. 1.

2. Implementation Details

This section first details how the real- and complex fractional exponential functions are implemented in an efficient manner and then discusses their integration in Theano. Considerations about numerical accuracy and improvements in computational efficiency are covered thereafter.

Algorithm 1 Computation of $\psi(x)$.

```
\begin{array}{l} \textbf{if } x < 0 \textbf{ then return } \exp(x) - 1 \\ \textbf{else} \\ k \leftarrow 0 \\ \textbf{while } x > 1 \textbf{ and } k < k_{max} (= 5) \textbf{ do} \\ x \leftarrow \log(x); \ k \leftarrow k + 1 \\ \textbf{end while} \\ \textbf{return } x + k \\ \textbf{end if} \end{array}
```

	CPU		GPU		
r_z	$ \begin{array}{c} e_{rel} \\ [10^{-5}] \end{array} $	t [ms]	$\frac{e_{rel}}{[10^{-5}]}$	t [ms]	M [MB]
0.1	27.45	269.13	31.05	9.29	0.62
0.05	6.757	274.57	9.698	10.38	2.45
0.01	0.278	388.48	1.313	15.80	60.91
0.0075	0.160	421.63	0.987	15.72	108.20
0.005	0.080	478.90	0.685	16.56	243.46
0.0035	0.050	490.75	0.541	16.85	496.60

		CI	CPU		GPU	
r_z	r_n	$\frac{e_{rel}}{[10^{-3}]}$	t [ms]	$\frac{e_{rel}}{[10^{-3}]}$	t [ms]	M [MB]
0.1	0.1	9.865	63.83	9.877	6.58	12.46
0.1	0.05	2.487	123.31	2.497	7.59	24.32
0.05	0.1	9.790	168.43	9.799	7.65	49.52
0.05	0.05	2.456	180.06	2.465	8.54	96.70
0.1	0.01	0.116	187.69	0.138	8.00	119.25
0.05	0.025	0.619	190.78	0.628	8.55	191.03
0.075	0.01	0.101	200.08	0.129	8.72	210.69
0.075	0.005	0.003	212.79	0.005	8.12	420.32

Table 1. Results for the interpolation of $\exp^{(n)}$ with method A (left) and method B (right). r_z and r_n are the sampling resolutions used for choosing the values to be precomputed. t denotes the runtime of computing function values for one million test points and M gives the required memory for the respective setting. All GPU computations have been conducted on a Nvidia Quadro K2200. The used CPU is a Intel Xeon E3-1226 v3 @ 3.30 GHz.

Algorithm 2 Computation of $\psi^{-1}(\psi)$. The value for ψ^{-1} with $\psi < -1$ would mathematically be $-\infty$. Similarly, inputs with $\psi \gtrsim 4.406$ evaluate to values larger than can be represented with single precision floating point numbers. Since infinite values cause problems in gradient-based learning, we restrict ψ^{-1} to the interval [-10, 10]. ψ_{min} and ψ_{max} are set as $\psi_{min} = -0.999955$ and $\psi_{max} = 2.83403$, so that $\psi^{-1}(\psi_{min}) = -10$ and $\psi^{-1}(\psi_{max}) = 10$.

```
if \psi < \psi_{min} then return \psi^{-1}(\psi_{min}) else if x > \psi_{max} then return \psi^{-1}(\psi_{max}) else k = \lceil \psi - 1.0 \rceil if k < 0 then return \log(\psi - k) end if \psi \leftarrow \psi - k while k > 0 do \psi \leftarrow \exp(\psi); \ k \leftarrow k - 1 end while return \psi end if
```