SQNL: A New Computationally Efficient Activation Function

Ms. Adedamola Wuraola, Dr. Nitish Patel

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Motivation

Motivation

Concept Mapping

Properties

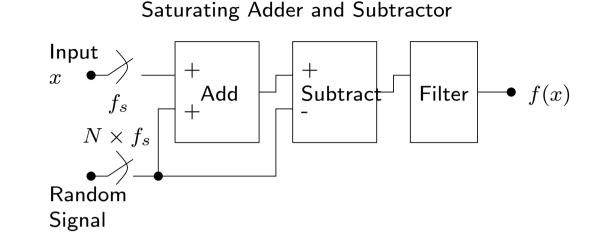
Performance Conclusions

Hardware ANN

- Playback (forward pass)
- Training (backpropagation)

 FPGA based, maybe ASIC based (neither Neuromorphic nor GPU/TPU styled)

Building a non-linearity: Concept



- Random source is uniformly distributed
- Adder and subtractors are saturating
- Filter can be complex or simple.

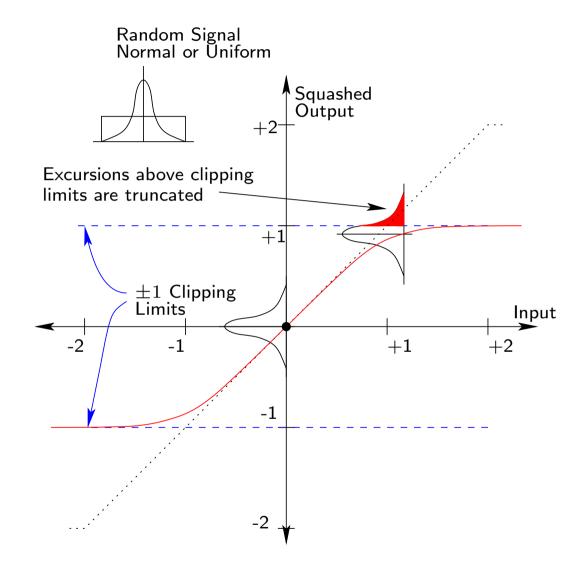


Mapping

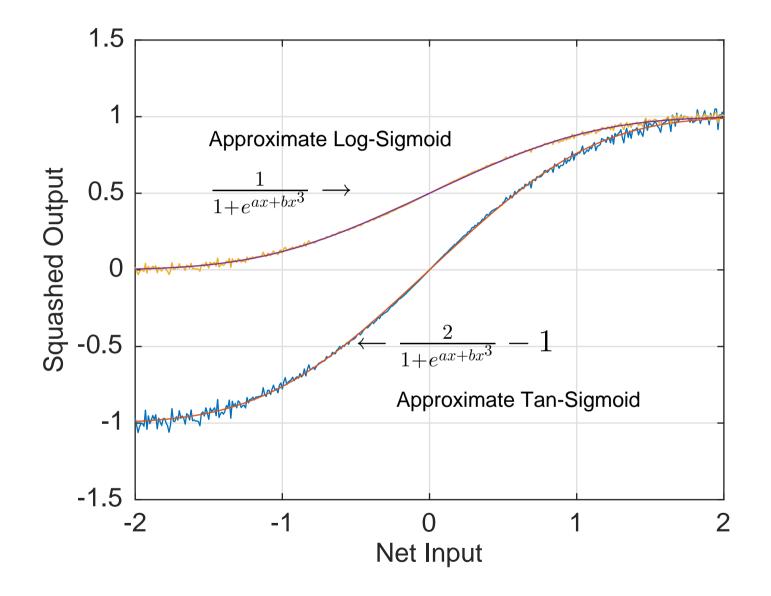
Properties

Performance Conclusions

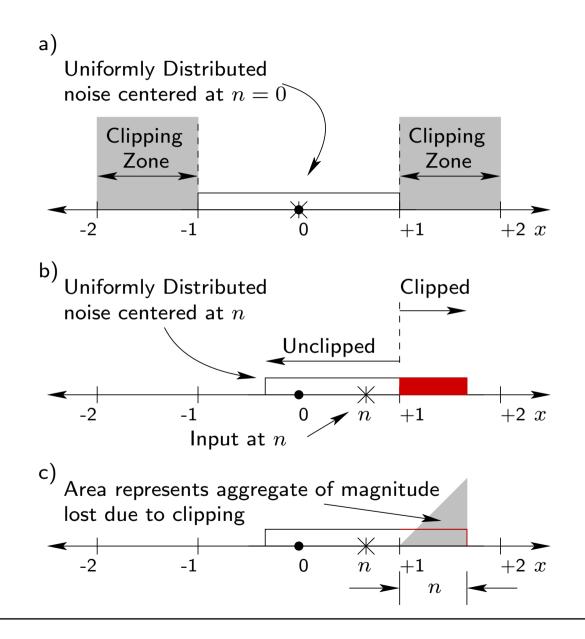
Concept (cont)



Concept (cont)



Concept (cont)



Mapping Function

The saturation results in a clipping o(x) that is given by

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$$o(x)\Big|_{x=n} = \frac{1}{2} \int_{1}^{1+n} (x-1)dx = \frac{n^2}{4}$$

The resultant effect is a non-linear mapping defined by

$$f_B(x) = \begin{cases} 1 & :x > 2.0\\ x - \frac{x^2}{4} & :0 \le x \le 2.0\\ x + \frac{x^2}{4} & :-2.0 \le x < 0\\ -1 & :x < -2.0 \end{cases}$$

Properties of the SQNL

The function has been named the Square-Law Non-Linear (SQNL) function due to its inherent square operation.

Simple Non-Linearity The square law is, arguably, the simplist non-linearity

Symmetrical and Continuous It is symetrical around zero and it is continuous between $-\infty$ and $+\infty$

Linear Derivative The derivative of the SQNL is linear.

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Performance

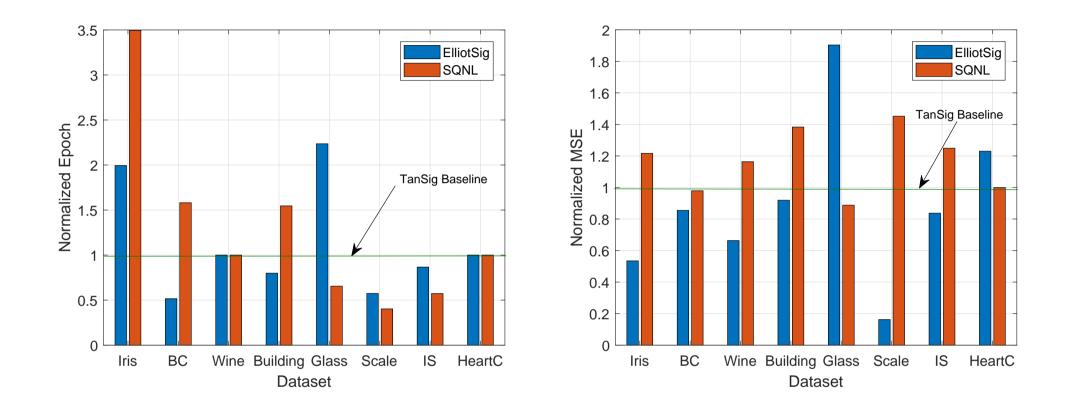
Motivation Concept Mapping Properties Performance Conclusions

Is this mapping comparable to other similar mappings?

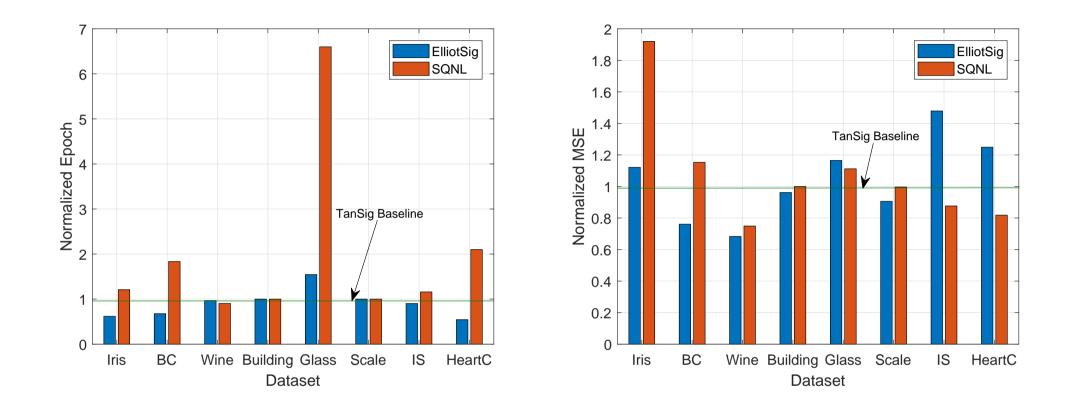
Method: By attempting to quantify both convergerce speed and the ability to generalise

- 100 networks trained: 100 weight-sets stored and reused with every experiment
- Only mapping function changed
- Both epochs and mean square error logged

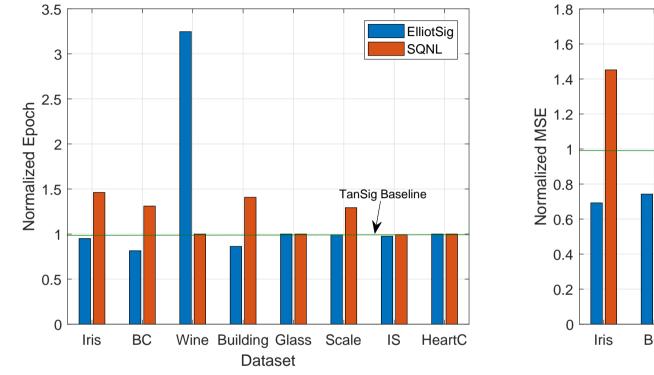
Backpropagation

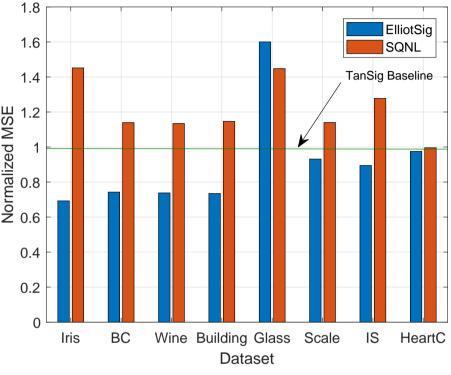


Levenberg-Marquandt



Resilient Backpropagation





Convergence Speed On MNIST Dataset

Function	Epoch
ElliotSig	17.74
TanSig	8.45
SQNL	8.11

Conclusions

Motivation Concept Mapping Properties Performance Conclusions

- The SQNL is a simple non-linearity
- Mathematically speaking, both forward and derivative functions are simple
- The SQNL seems to perform better.
- However, the variation in performance suggests a strong data set dependence
- Importantly, the SQNL is not inferior to the well established TanSig.
- Digital circuit implementations are possible