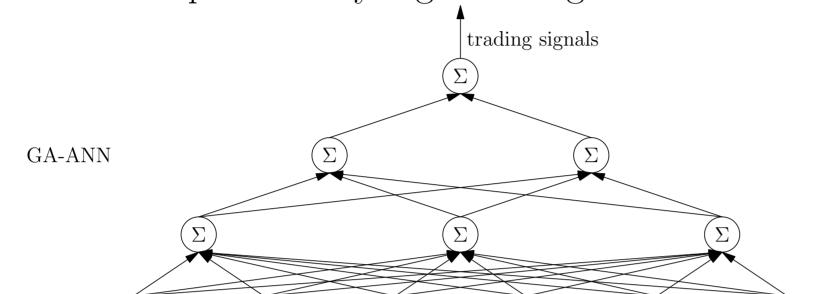
# **OpenCL**におけるNeuroWavelet ザパート クリストファー データ同化研究開発センター 特任助教

Multiresolution Analysis (フラクタルと同 1 檨)

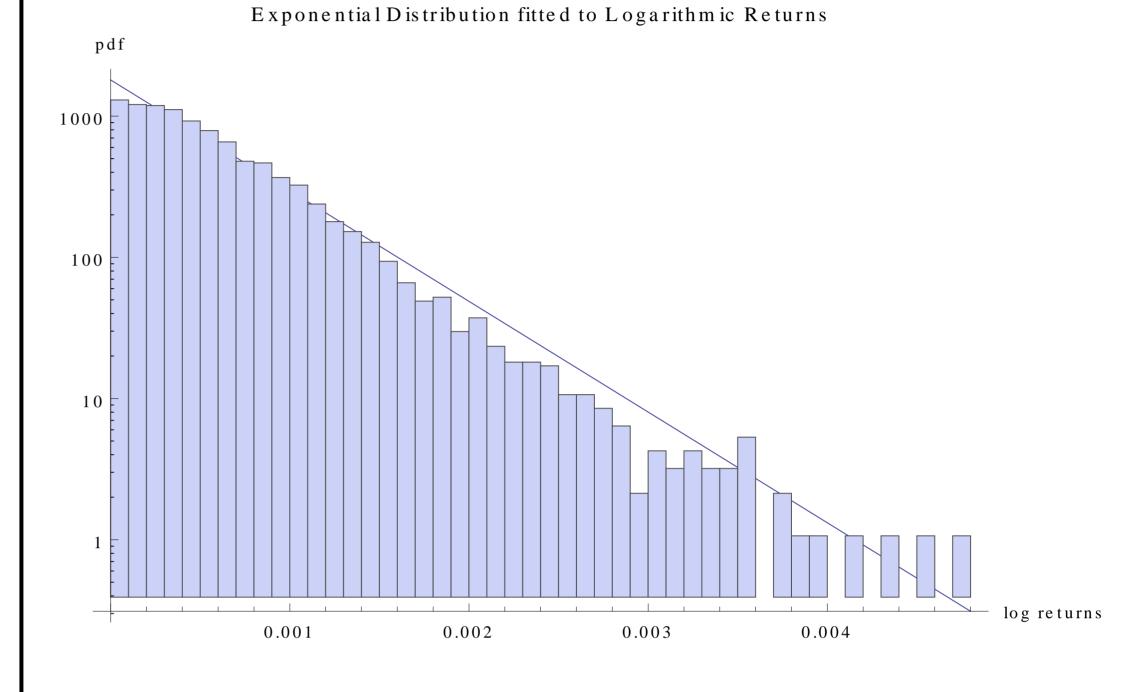
Wavelets + Neural Networks  $\rightarrow$  NeuroWavelet

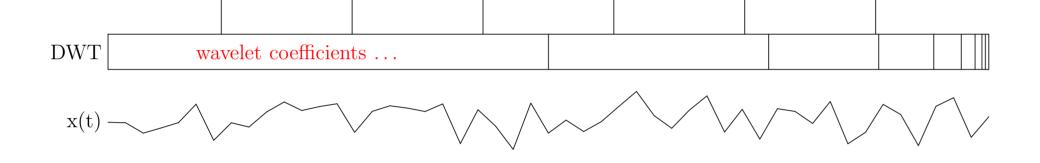
- fractals: natural basis for financial time series (Mandelbrot)
- Wavelet Transform resembles fractals in that it operates on multiple time scales
- early examples of fusing neural networks with wavelets can be found in papers by Murtagh et al.
- A NeuroWavelet architecture: artificial neural networks are optimised by a genetic algorithm



- extracted wavelet coefficients are passed as inputs to a small neural network, which is trained with genetic algorithms to output trading signals  $y \in (-1, +1)$
- need to prune weights in order to prevent excessive over-fitting (fitting past data too closely without learning how to generalise knowledge to handle new unseen cases), which is a common problem in financial time series prediction

Absolute values of logarithmic returns approximated with the Exponential Distribution





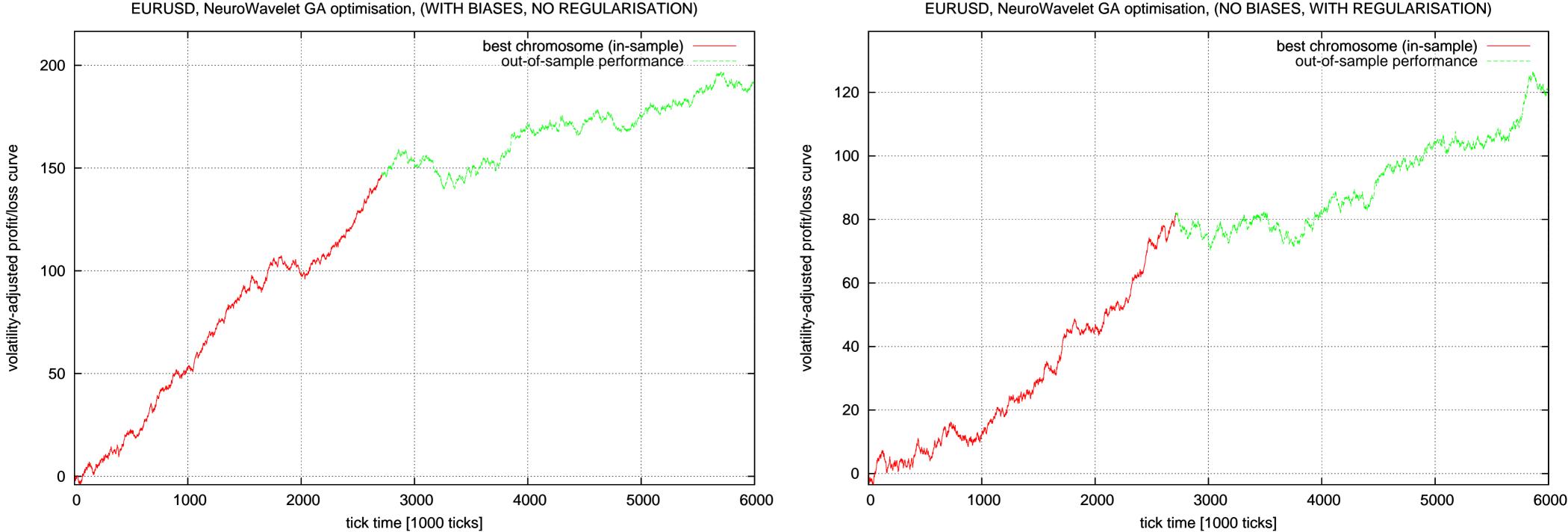
## 工程

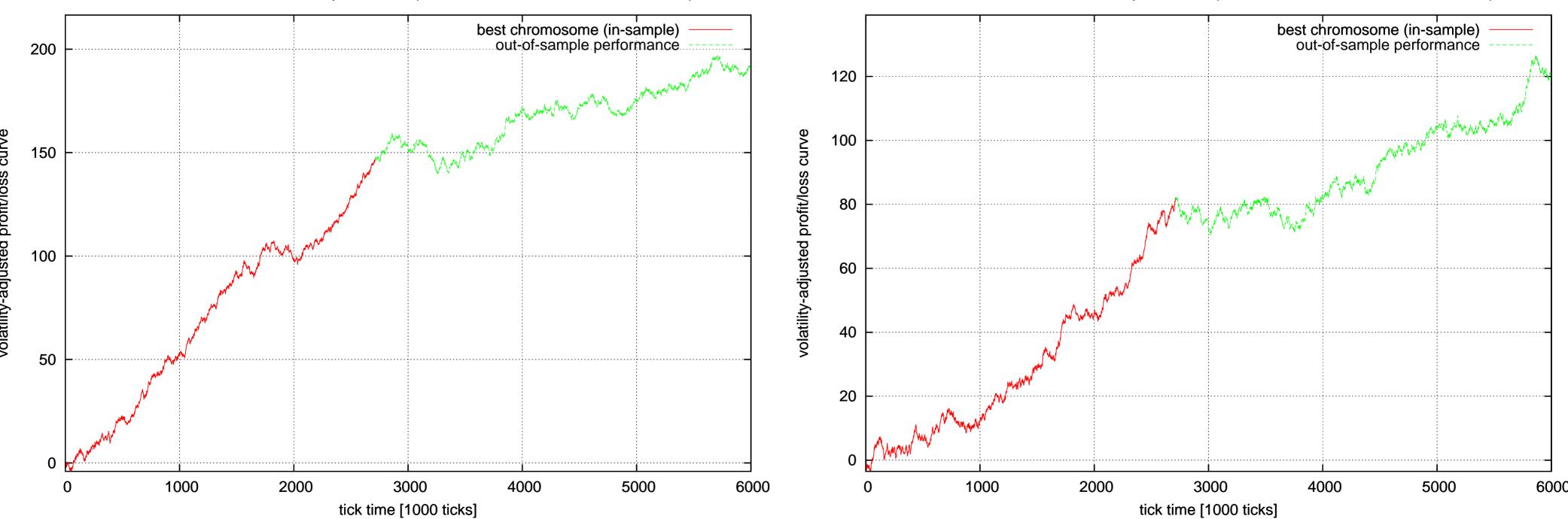
- financial time series sampled discretely once every 1,000 ticks
- need to "stabilise" the samples in order to reduce non-stationarity and deal with changing volatility
- an exponential distribution  $f(x, \lambda) = \lambda \exp^{-\lambda x}, x \ge 0$  is fitted to absolute values of logarithmic returns, parametrised by  $\lambda$  set to a reciprocal of the exponential moving average of absolute returns
- a discrete Biorthogonal Spline wavelet transform is applied to a window containing 32 samples of past volatility-adjusted financial returns x(t) for the EUR/USD exchange rate

#### weight prunning (regularisation) 3

- bias units are removed from all neurons to enforce y(-x) = -y(x) (prevent over-fitting long one-sided bull or bear market phases)
- filtering of wavelet coefficients (which form an input space for neural networks) to remove spurious noise, achieved by minimising the sum-ofsquares of input neuron weights
- maximisation of the separation of discovered input features (wavelet coefficients), achieved by imposing an orthonormalisation constraint on weights of input neurons (forcing input features to be uncorrelated)

#### Experimental results (OpenCLによるGPGPU) 4





#### EURUSD, NeuroWavelet GA optimisation, (NO BIASES, WITH REGULARISATION)

Table 1: Simulation results: no regularisation (left chart), compared with regularisation (right chart). Unseen test set performance marked in green.



### The Institute of Statistical Mathematics