

# Draft

## Overview

There are many work related to the topic on wavelet and a special type of neural network, namely, wavelet neural network, however, not much of work directly relate to trading strategy. Wavelet has been proposed to be a better way of combine both frequency domain and time domain. The idea is using dilation and translation to expand the so called mother wavelets in order to form an orthogonal basis in  $L^2$  with wavelets. The neural networks are proved to be able to approximate any continuous mapping(Funahashi (1989)). The combination of the two was proposed only several years later. Zhang and Benveniste (1992) first propose the idea of wavelet network to approximate arbitrary nonlinear functions. Zhang (1993) introduced the Wavelet as a regression selection procedure in attempt to solve the problem of random initialization of neural network.

In time series analysis domain, Pindoriya, Singh, and Singh (2008) applied wavelet neural network in energy price forecasting in electricity markets. Also in electricity market, Khoa et al. (2004) using same set of techniques to predict long-term load.

In finance, Yang Yiwen, Liu Guizhong, and Zhang Zongping (2000) propose to predict stock trend prediction with nn and multiresolution analysis. *this can be utilize to form a momentum based trading strategy.*

The most closed related work is Wang and Gupta (2013), where they use wavelet combine with neural network to predict stock prices. They, however, only use wavelet to denoise the data not using wavelet coefficients in the neural network. The trading strategy is on based on daily predictions of stock prices. The strategy earned significant return in backtest.

## the scheme

Predicting the price of next day seemly good example but lack of realistic value in stock trading. While the volatility will make the prediction hardly reliable, the friction in the real stock market will probably make the strategy less likely to be profitable. In inspired by the momentum strategy, we are looking for potential return in a relative longer period.

In the light of optimal strategy suggested by (???), we will try to use wavelet neural network to predict the return rate of the stock in next 3 months based on the data of last 12 month. Hopefully, this will outperform the benchmark which is the return of the traditional momentum strategy.

## References

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