

## Using Neural Networks to Forecast Stock Market Prices

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No.

Ramon Lawrence

Possible approaches:

- Technical analysis - detects trends based on previous trends and market psychology. Low predictive accuracy and time lag
- Fundamental analysis - tends to give superior long-term returns
- Time series forecasting - either **univariate** (e.g. **Box-Jenkins**), which involves analysing patterns in autocorrelation, or **multivariate**, to discover causal relationships
- Efficient Market Hypothesis - i.e., buy and hold
- Chaos theory - the stock market is not purely random; it is massively complex and nonlinear.

Neural networks can capture nonlinearity, and can extract rules from data.

- Input data can take many forms. Most commonly:
  - technical indicators e.g. moving averages, RSI
  - fundamentals e.g. DCF value
  - sentiment by parsing reports/media
- Performance increases with the amount of data (generally).
- However, in the case of time series, very old data may just add noise.

## Network organisation

- Backpropagation is the most common architecture
  - node weights change in proportion to their error contribution
  - different activations can be used. The sigmoid function is best at learning average behaviour; tanh is better for deviations.
  - overfitting can be dealt with by pruning, or **cross validation**



- **Supplementary learning** involves updating individual weights based on the total error. Each output node has an error threshold such that backprop only occurs if this tolerance is exceeded.
- **Moving simulation** involves constantly changing the target, learning, and prediction periods. In each iteration, the train-test-pred window moves forward in time.
- **Genetic algorithms** may be useful for large input dimensionalities, or for **hyperparameter optimisation** (e.g. deciding on NN architecture).
- **Modular NNs** are often made of smaller backprop nets which act as interchangeable subunits.
- RNNs tend to be good at dealing with temporal data.
- **Self-organising systems** require a lot of data but are often difficult to train and prone to overfitting.
- **Hybrid systems** pass NN output through an expert system rule based, designed with domain knowledge

### Comments

- Poorly organised: a lot of repetition within a few lines, and the ideas are presented in an illogical order.
- Presents many individual results without linking ideas or adding value.
- Some of the networks discussed appear to have disappeared from the literature.