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## Deep learning for finance: deep portfolios

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## Abstract

We explore the use of deep learning hierarchical models for problems in financial prediction and classification. Financial prediction problems – such as those presented in designing and pricing securities, constructing portfolios, and risk management – often involve large data sets with complex data interactions that currently are difficult or impossible to specify in a full economic model. Applying deep learning methods to these problems can produce more useful results than standard methods in finance. In particular, deep learning can detect and exploit interactions in the data that are, at least currently, invisible to any existing financial economic theory. Copyright © 2016 John Wiley & Sons, Ltd.

## References

- 1 Dean J, Corrado G, Monga R, et al. Large scale distributed deep networks. *Advances in Neural Information Processing Systems* 2012; 25: 1223–1231.

[Google Scholar](#)

- 2 Ripley BD. *Pattern Recognition and Neural Networks*. Cambridge University Press: Cambridge, 1996.

[Google Scholar](#)

- 3 Kolmogorov A. The representation of continuous functions of many variables by superposition of continuous functions of one variable and addition. *Dokl. Akad. Nauk SSSR* 1957; 114: 953–956.

[Web of Science®](#) | [Google Scholar](#)

4 Diaconis P, Shahshahani M. On non-linear functions of linear combinations. *SIAM Journal on Scientific and Statistical Computing* 1984; 5(1): 175–191.

| [Web of Science®](#) | [Google Scholar](#) |

---

5 Lorentz GG. The 13th problem of Hilbert. *Proceedings of Symposia in Pure Mathematics, American Mathematical Society* 1976; 28: 419–430.

| [Google Scholar](#) |

---

6 Gallant AR, White H. There exists a neural network that does not make avoidable mistakes. *IEEE International Conference on Neural Networks* 1988; 1: 657–664.

| [Google Scholar](#) |

---

7 Poggio T, Girosi F. Networks for approximation and learning. *Proceedings of the IEEE* 1990; 78(9): 1481–1497.

| [Web of Science®](#) | [Google Scholar](#) |

---

8 Hornik K, Stinchcombe M, White H. Multilayer feedforward networks are universal approximators. *Neural networks* 1989; 2(5): 359–366.

| [Web of Science®](#) | [Google Scholar](#) |

---

9 LeCun YA, Bottou L, Orr GB, Muller KR. Efficient backprop. *Neural Networks: Tricks of the Trade* 1998; 1524: 9–48.

| [Web of Science®](#) | [Google Scholar](#) |

---

10 Polson NG, Scott JG, Willard BT. Proximal algorithms in statistics and machine learning. *Statistical Science* 2015; 30: 559–581.

| [Web of Science®](#) | [Google Scholar](#) |

---

11 Stein C. Estimation of the mean of a multivariate normal distribution. *Journal of the American Statistical Association* 1981; 97: 210–221.

| [Google Scholar](#) |

---

12 Hinton GE, Salakhutdinov RR. Reducing the dimensionality of data with neural networks. *Science* 2006; 313(5786): 504–507.

| [CAS](#) | [PubMed](#) | [Web of Science®](#) | [Google Scholar](#) |

13 Srivastava. et al. Dropout: a simple way to prevent neural networks from overfitting. *Journal of Machine Learning Research* 2014; 15: 1929–1958.

| [Web of Science®](#) | [Google Scholar](#) |

14 Lake BM, Salakhutdinov R, Tenenbaum JB. Human-level concept learning through probabilistic program induction. *Science* 2015; 3560: 1332–1338.

| [CAS](#) | [Web of Science®](#) | [Google Scholar](#) |

15 Cook RD. Fisher lecture: dimension reduction in regression. *Statistical Science* 2007: 1–26.

| [Web of Science®](#) | [Google Scholar](#) |

16 Hutchinson JM, Lo AW, Poggio T. A nonparametric approach to pricing and hedging derivative securities via learning networks. *Journal of Finance* 1994; 48(3): 851–889.

| [Google Scholar](#) |

17 Wold H. Causal inference from observational data: a review of end and means. *Journal of the Royal Statistical Society* 1956; Series A(General): 28–61.

| [Web of Science®](#) | [Google Scholar](#) |

18 Hastie T, Tibshirani R, Friedman J. *The Elements of Statistical Learning*, vol. 2, 2009.

| [Google Scholar](#) |

19 Sirignano J. Deep learning for limit order books, 2016. arXiv 1601.01987v7.

| [Google Scholar](#) |

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