Appendix D — Bibliography

The bibliography is intended to be practical rather than encyclopedic. I have listed books and websites that I have found provide background and examples that are helpful in learning about and implementing trading systems with the characteristics that have high probability of being profitable with low risk.

This list is heavy in machine learning, pattern recognition, probability, statistics, modeling, and simulation—because those topics are of primary importance in developing quantitative trading systems. It is light on traditional trading systems, indicators, and charting—because those topics are not very useful for systems that fit the trading profile most likely to be profitable with low risk.

If you can only read one

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