

NGUIA NGUIA SINGIA FINANCE

GAINING
PREDICTIVE EDGE
IN THE MARKET

PAUL D. McNELIS

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Neural Networks in Finance

GAINING PREDICTIVE EDGE IN THE MARKET

"This book clarifies many of the mysteries of neural networks and related optimization techniques for researchers in both economics and finance. It contains many practical examples backed up with computer programs for readers to explore. I recommend it to anyone who wants to understand methods used in nonlinear forecasting."

- Blake LeBaron, Professor of Finance, Brandeis University,

"An important addition to the select collection of books on financial econometrics...

Neural Networks in Finance serves as an important reference on neural network

models of nonlinear dynamics as a practical econometric tool for better decisionmaking in financial markets."

 Roberto S. Mariano, Dean of School of Economics and Social Sciences & Vice-Provost for Research, Singapore Management University; Professor Emeritus of Economics, University of Pennsylvania

"This book represents an impressive step forward in the exposition and application of evolutionary computational tools. The author illustrates the potency of evolutionary computational tools through multiple examples, which contrast the predictive outcomes from the evolutionary approach with others of a linear and general nonlinear variety. The book will be of utmost appeal to both academics throughout the social sciences as well as practitioners, especially in the area of finance."

 Carlos Asilis, Portfolio Manager, VegaPlus Capital Partners; formerly Chief Investment Strategist, JPMorgan Chase

Neural Networks in Finance explores the intuitive appeal of neural networks and the genetic algorithm in finance. It demonstrates how neural networks used in combination with evolutionary computation outperform classical econometric methods for accuracy in forecasting, classification and dimensionality reduction. The text shows that these networks are easy to implement and interpret once the time-honored quest for closed form solutions is reconsidered.

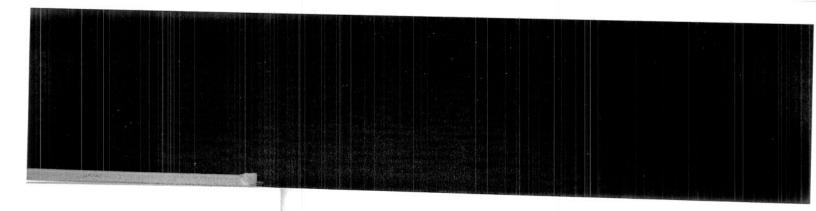
McNelis utilizes a variety of examples, from forecasting automobile production and corporate bond spread, to inflation and deflation processes in Hong Kong and Japan, to credit card default in Germany, to bank failures in Texas, to cap-floor volatilities in New York and Hong Kong. Numerical illustrations use MATLAB code and the book is accompanied by a website.

Paul McNelis is Professor of Economics, Georgetown University, Washington, DC.

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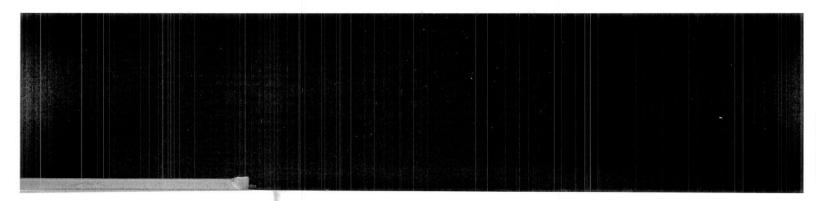
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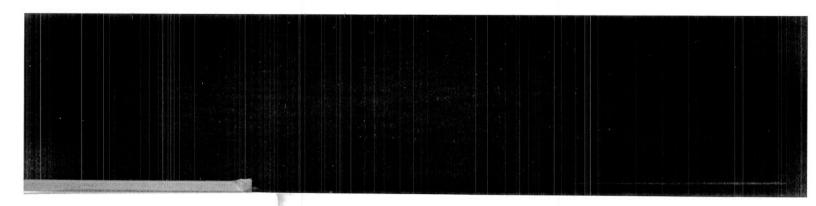
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Preface

Adjusting to the power of the a whole different mind-set for l

Thomas Friedman, The Lexus

Questions of finance and ma quantitative. Applied research in predicting the direction of return, spreads, or likelihood in economic conditions, polic behavior in domestic or foreign both the precision of the esting default rates, as well as the constitution of the estimates may be obtained. If and computational.

Peter Bernstein (1998) rer Gods, that the driving force was the precise calculation o represent the foremost "game doubt that the precise calcul game is the driving force in quand policy evaluation.

Besides precision, speed of quantitative financial analysi or in financial institutions do having to commit to buy or