*Edited by* Thomas Walker · Frederick Davis Tyler Schwartz

# **Big Data in Finance** Opportunities and Challenges of Financial Digitalization

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Thomas Walker · Frederick Davis · Tyler Schwartz Editors

# Big Data in Finance

Opportunities and Challenges of Financial Digitalization

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### PREFACE

Our global society is becoming increasingly data centric. The use of large, detailed datasets has revolutionized many fields, including among others—medicine, biology, manufacturing, sports, marketing, and finance. With advances in how data can be collected and stored, a new phenomenon has emerged: big data. In simple terms, big data can be understood as a large amount of data that can be analyzed to understand past patterns better or predict future outcomes. This book examines the technical aspects of recent innovations surrounding big data in finance, as well as the benefits and risks associated with these developments. Moreover, the book sheds light on the ethical and privacy issues associated with big data, as well as the environmental footprint of collecting, storing, and analyzing big datasets.

The book features contributions from the international community of scholars and practitioners who work at the interface of artificial intelligence, big data, and finance. The authors review and critically analyze new developments at the intersection of big data and finance, and provide different perspectives on their impact on the financial sector and the way it operates. The book serves as a technical guide of these developments, exploring the theory and mechanisms behind the algorithms using big data, and exploring their use in a finance context. The contributors explain and demonstrate the predictive capabilities of big data in finance using different model types such as supervised, unsupervised, and semi-supervised learning. Moreover, because big data in finance has many applications that extend beyond financial institutions, the book features contributions that explore possible policy and sustainability-oriented solutions and implications of the use of big data in finance.

Montreal, Canada Montreal, Canada Montreal, Canada Thomas Walker Frederick Davis Tyler Schwartz

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Introduction



### Big Data in Finance: An Overview

Thomas Walker, Frederick Davis, and Tyler Schwartz

#### 1 INTRODUCTION

In the financial sector, the big data movement refers to the analysis of vast amounts of data with the goal of making better informed investment decisions, improving corporate operations, and enhancing decision-making processes on both the buy and supply sides of transactions (Hasan et al., 2020). Big data analysis frequently draws on artificial intelligence (AI) models and has created a paradigm shift in the operations of financial institutions (Nobanee, 2021; Sun et al., 2019). Big data and its impact on the financial industry now benefit from scholarly interest; many publications, reports, books, and conventions focus on dissecting and comparing different applications and techniques within the domain,

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© The Author(s), under exclusive license to Springer Nature Switzerland AG 2022 T. Walker et al. (eds.), *Big Data in Finance*, https://doi.org/10.1007/978-3-031-12240-8\_1 with researchers envisioning what the movement means for the future of finance (Nobanee, 2021). The two main subtopics of interest are big data itself as well as the AI techniques used to analyze that data.

AI is best understood as computer programming that automates tasks that have traditionally relied on human intelligence (Russell, 2010). A popular branch of AI is machine learning, which uses statistical techniques to allow machines to "learn" through experience by correcting previous errors (Alzubi et al., 2018). Spacecraft engineering, pattern recognition, entertainment, biology, finance, and medicine, among others, all rely on machine learning (El Naqa & Murphy, 2015).

More recently, deep learning, a sub-domain of machine learning, has proven to be effective in extracting patterns from big data (Chen & Lin, 2014). Deep learning uses neural networks, which are inspired by the neurological basis of brain functioning, to extract patterns from big data (LeCun et al., 2015). Primary applications of deep learning include image and voice recognition, as well as natural language processing (NLP) tasks such as text summarization and classification (LeCun et al., 2015; Najafabadi et al., 2015).

In the financial industry, AI has been successfully employed by financial institutions in such areas as automated lending, portfolio construction and management (robo-advising), risk management, fraud detection, quantitative and high-frequency trading, as well as customer support (Bahrammirzaee, 2010; Buchanan, 2019; Buchanan & Wright, 2021). As the complexity of algorithms increases with new developments in deep learning and other technological advancements, the role of AI in finance is certain to become more important as well, making the field a rapidly changing frontier in the financial technology (FinTech)/banking nexus and making space not only for opportunities, but also for substantial ethical, social, legal, and economic risks (Buchanan, 2019).

Big data, as a discipline, is what allows AI to learn and develop complex pattern-detecting algorithms used for a variety of purposes (O'Leary, 2013). The troves of information collected in finance, for example, include customer information on banking, historical prices of stocks for investing, social media activity to predict the current sentiment of the market, and past fraudulent transactions for fraud detection (Buchanan, 2019). The data can be structured (organized, classified data), unstructured (text, social media activity), or semi-structured (incorporating both structured and unstructured elements) (Hurwitz et al., 2013). However, most of today's data is unstructured and/or semi-structured, leading to

greater difficulties in finding a way to use the data efficiently (Yaqoob et al., 2016).

Gandomi and Haider (2015) review the existing literature and define big data using the traditional three Vs (volume, variety, and velocity) along with three more recent Vs developed by International Business Machines (IBM) (veracity), Statistical Analysis System (SAS) (variability), and Oracle (value):

- Volume: the size of the data,
- Variety: the diversity of the data,
- Velocity: the speed at which the data is generated and processed,
- Veracity: the reliability of the data (for example, social media data),
- Variability: the variability in the velocity of data,
- Value: the quality and usefulness of the data.

Currently, many questions dominate big data scholarship and professional practice, including how data is collected (consumer privacy and ethical concerns), stored (environmental impacts), secured, as well as analyzed and used (Jain et al., 2016; Lucivero, 2020; Martin, 2015). This edited book explores new developments surrounding big data in finance, and aims to provide meaningful answers to the questions posed above. While a general use of big data has often been the subject of discussion, this book will take a more focused look at big data applications in the financial sector to differentiate itself from current offerings and close a large gap in the literature.

### 2 Overview of Content

This edited book critically analyzes new developments at the intersection of big data and finance and provides different perspectives on their impact on the financial sector and the way it operates. As illustrated above, the predictive capability of big data is still developing as increasingly complex algorithms are devised by institutions to extract the most out of available data. The chapters in this book explore the predictive ability of big data in the context of finance, using different model types such as supervised learning and unsupervised learning. In addition, several of the chapters discuss the leveraging of recent advancements made in deep learning in combination with big data to innovate the financial sector. Lastly, because the use of big data in finance has many implications that go beyond their use in financial institutions, several chapters address the ethical, privacy, and environmental implications of these applications.

#### 2.1 Part I: Big Data in the Financial Markets

The book's first section looks at the use of big data in the financial markets, where it can be employed to optimize investment performance in various ways.

The section begins with Chapter 2, *Alternative Data*, where the authors discuss the role alternative data sources (e.g., internet activity, web traffic logs, chat boards, and social media platforms) can play in financial investment decisions. **Grégoire and Jepson** provide a critical analysis of the value gained from using such data sources against the difficulties in using that data—including the high cost associated with collecting, validating, and maintaining the data. The authors discuss the role big data can play in making alternative data more affordable to use.

Chapter 3, An Algorithmic Trading Strategy to Balance Profitability and Risk, proposes an algorithmic trading strategy (AT) based on financial indicators. The primary financial indicator this strategy is based on is the Balanced Investment Indicator (BII) which is used to find a balance between profitability and risk for an investor. **Peña** argues that using this proposed financial indicator for portfolio construction in combination with big data can provide an effective algorithmic trading strategy.

Following this, Chapter 4, *High-Frequency Trading and Market Efficiency in the Moroccan Stock Market*, explores whether high-frequency trading (HFT) impacts market efficiency. **Ferrouhi and Bouabdellaoui** employ a case study approach using data from the Moroccan Stock Exchange to evaluate this hypothesis at different time frequencies.

Lastly, Chapter 5, Ensemble Models using Symbolic Regression and Genetic Programming for Uncertainty Estimation in ESG and Alternative Investments, presents an ensemble algorithm using symbolic regressions (SR) to estimate the environmental, social, and governance (ESG) scores of publicly traded private equities (PE) and sustainable exchange-traded funds (ETFs). Venegas, Britez, and Gobet demonstrate how using the SR algorithm can reduce uncertainty in this investment space.